

00:00:01.230 --> 00:00:03.780
<v ->And I'm gonna turn everything over to Dr. Wright.</v>

2
00:00:07.730 --> 00:00:09.770
<v ->Good morning and welcome to the third session</v>

3
00:00:09.770 --> 00:00:12.940
of our four-part Data Science Career Seminar Series:

4
00:00:12.940 --> 00:00:15.350
Bringing Data Science to Addiction Research.

5
00:00:15.350 --> 00:00:17.040
My name is Susan Wright and I'm

6
00:00:17.040 --> 00:00:19.940
from the Division of Neuroscience and Behavior or D&B

7
00:00:19.940 --> 00:00:20.920
and I'm the Program Director

8
00:00:20.920 --> 00:00:22.930
for Big Data and Computational Science.

9
00:00:22.930 --> 00:00:25.750
I'm leading our data science efforts here at NIDA.

10
00:00:25.750 --> 00:00:28.180
Training in data science is also a priority for NIDA

11
00:00:28.180 --> 00:00:29.240
and it's supported by our new

12
00:00:29.240 --> 00:00:32.320
Office of Research Training, Diversity, and Disparities

13
00:00:32.320 --> 00:00:34.550

or ORTDD.

14

00:00:34.550 --> 00:00:36.130

We've organized a seminar series

15

00:00:36.130 --> 00:00:38.760

with the full support of our NIDA Director, Dr. Nora Volkow,

16

00:00:38.760 --> 00:00:40.420

and the organizers include members

17

00:00:40.420 --> 00:00:42.720

of the Division of Neuroscience and Behavior and the

18

00:00:42.720 --> 00:00:45.543

Office of Research Training, Diversity, and Disparities.

19

00:00:47.400 --> 00:00:50.460

The organizers include myself, Roger Little,

20

00:00:50.460 --> 00:00:51.540

who is the Deputy Director

21

00:00:51.540 --> 00:00:54.090

of the Division of Neuroscience and Behavior.

22

00:00:54.090 --> 00:00:56.810

Dr. Wilson Compton, NIDA Deputy Director

23

00:00:56.810 --> 00:00:58.080

and Acting Director of the

24

00:00:58.080 --> 00:01:02.140

Office of Research Training, Diversity, and Disparities.

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00:01:02.140 --> 00:01:04.620

Dr. Albert Avila, the deputy director of the

26

00:01:04.620 --> 00:01:07.430

Office of Research Training, Diversity, and Disparities,

27

00:01:07.430 --> 00:01:08.263

and the Director

28

00:01:08.263 --> 00:01:10.720

of the Office of Disparities and Health Disparities.

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00:01:10.720 --> 00:01:11.970

And Dr. Lindsey Friend,

30

00:01:11.970 --> 00:01:14.600

the Research and Career Development Program Officer in the

31

00:01:14.600 --> 00:01:18.120

Office of Research Training, Diversity, and Disparities.

32

00:01:18.120 --> 00:01:20.870

I want to thank Roger, Wilson, Albert, and Lindsey

33

00:01:20.870 --> 00:01:23.200

for their help with organizing the seminar series.

34

00:01:23.200 --> 00:01:24.341

And I also want to thank the team

35

00:01:24.341 --> 00:01:26.370

that has been helping with the technical details

36

00:01:26.370 --> 00:01:29.410

and that includes Usha Charya, Susan Holbrook,

37

00:01:29.410 --> 00:01:31.790

Caitlin Dudevoir, and David Metzger.

38

00:01:32.650 --> 00:01:35.860

For this session, we're highlighting women in data science.

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00:01:35.860 --> 00:01:39.200

First, we'll have an interview with Dr. Kristian Lum,

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00:01:39.200 --> 00:01:41.350

then a presentation by Dr. Brenda Curtis,

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00:01:41.350 --> 00:01:43.260

followed by a joint Q&A session

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00:01:43.260 --> 00:01:45.210

where we'll take questions from the audience.

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00:01:45.210 --> 00:01:47.580

Please use the chat box to submit your questions

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00:01:47.580 --> 00:01:49.730

and we'll get to as many of them as we can.

45

00:01:51.520 --> 00:01:54.610

Dr. Kristian Lum is an Assistant Research Professor

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00:01:54.610 --> 00:01:56.900

in the Department of Computer and Information Science

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00:01:56.900 --> 00:01:58.920

at the University of Pennsylvania.

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00:01:58.920 --> 00:02:00.490

She studies and develops statistical

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00:02:00.490 --> 00:02:01.520

and machine learning models

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00:02:01.520 --> 00:02:03.770

to tackle problems with social impact.

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00:02:03.770 --> 00:02:06.580

This includes statistical population estimation models

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00:02:06.580 --> 00:02:08.610

to estimate the number of undocumented victims

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00:02:08.610 --> 00:02:11.460

of human rights violations, fair algorithms for use

54

00:02:11.460 --> 00:02:14.590

in high stakes decision making and epidemiological models

55

00:02:14.590 --> 00:02:16.170

to study disease spread among

56

00:02:16.170 --> 00:02:20.050

and between marginalized populations in a better community.

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00:02:20.050 --> 00:02:21.610

Dr. Lum is particularly interested

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00:02:21.610 --> 00:02:23.780

in applications in criminal justice.

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00:02:23.780 --> 00:02:25.770

She enjoys using the tools of statistics

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00:02:25.770 --> 00:02:27.040

and machine learning to shine a light

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00:02:27.040 --> 00:02:29.370

on alternative interpretations of data.

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00:02:29.370 --> 00:02:32.340

She is often just as interested

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00:02:32.340 --> 00:02:34.981

in what is missing from a dataset and what is in it.

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00:02:34.981 --> 00:02:38.110

So please join me in welcoming Dr. Kristian Lum

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00:02:38.110 --> 00:02:39.660

for the interview this morning.

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00:02:40.500 --> 00:02:42.123

Thanks for joining us, Kristian.

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00:02:43.040 --> 00:02:44.190

<v ->Thanks for having me.</v>

68

00:02:45.250 --> 00:02:48.480

<v ->So you're currently an Assistant Research Professor</v>

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00:02:48.480 --> 00:02:50.340

in the Department of Computer and Information Science

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00:02:50.340 --> 00:02:52.240

at the University of Pennsylvania,

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00:02:52.240 --> 00:02:53.577

can you tell us about your career path

72

00:02:53.577 --> 00:02:55.990

and how it led you to your current position?

73

00:02:55.990 --> 00:02:56.823

<v ->Yeah, sure.</v>

74

00:02:56.823 --> 00:03:00.430

So it definitely has not been a linear path.

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00:03:00.430 --> 00:03:03.300

So I started out in statistics.

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00:03:03.300 --> 00:03:04.900

This is about,

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00:03:04.900 --> 00:03:06.077

Gosh, like 10 years ago or so.

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00:03:06.077 --> 00:03:09.077

It was when I finished my PhD in that, so it's been a while.

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00:03:10.200 --> 00:03:12.290

After that, I actually took some time off,

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00:03:12.290 --> 00:03:15.160

which I, by the way, I think is a great thing to do

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00:03:15.160 --> 00:03:17.010

if you need a little bit of a refresh,

82

00:03:17.010 --> 00:03:18.470

and I realize that not everybody can,

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00:03:18.470 --> 00:03:20.200

but for me, it was the right call.

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00:03:20.200 --> 00:03:22.370

And when I came back, I started

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00:03:22.370 --> 00:03:25.383

as an Assistant Research Professor at Virginia Tech,

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00:03:26.250 --> 00:03:27.083

where I was working

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00:03:27.083 --> 00:03:29.630

in the Network Dynamics and Simulation Science Laboratory.

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00:03:29.630 --> 00:03:32.440

It's been renamed since, but it was essentially working

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00:03:32.440 --> 00:03:34.450
on things like age based models,

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00:03:34.450 --> 00:03:37.770
some computational epidemiology on these age based models.

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00:03:37.770 --> 00:03:38.820
And mostly what I focused

92

00:03:38.820 --> 00:03:43.010
on there was developing synthetic populations

93

00:03:44.170 --> 00:03:46.270
of the the United States and other countries.

94

00:03:46.270 --> 00:03:49.350
So basically simulated representations of individual people

95

00:03:49.350 --> 00:03:51.630
that were demographically representative and like layering

96

00:03:51.630 --> 00:03:53.850
on top all sorts of other information in a way

97

00:03:53.850 --> 00:03:57.130
that was consistent with data sources that were available.

98

00:03:57.130 --> 00:04:01.440
From there, I ended up taking a different job.

99

00:04:01.440 --> 00:04:04.210
So I worked briefly in at Datapad,

100

00:04:04.210 --> 00:04:06.770
which was sort of Silicon Valley

101

00:04:06.770 --> 00:04:08.510

like data science nonprofit,

102

00:04:08.510 --> 00:04:09.820

sorry, start up, not nonprofit,

103

00:04:09.820 --> 00:04:12.000

I'll get to that in a second, it was a startup

104

00:04:12.000 --> 00:04:14.263

and that was acquired soon after I started.

105

00:04:15.650 --> 00:04:17.930

And so I will take a quick step back,

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00:04:17.930 --> 00:04:20.010

which is gonna explain the next six years

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00:04:20.010 --> 00:04:23.900

of my career after this is that when I was in grad school,

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00:04:23.900 --> 00:04:27.760

I sort of like cold emailed a person,

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00:04:27.760 --> 00:04:28.730

sort of an organization

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00:04:28.730 --> 00:04:30.280

whose work I was really interested in.

111

00:04:30.280 --> 00:04:33.730

So Patrick Ball was the leader of the organization

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00:04:33.730 --> 00:04:36.300

and he sort of was really involved

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00:04:36.300 --> 00:04:37.870

in this casualty estimation stuff.

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00:04:37.870 --> 00:04:39.900

So estimating, like I think you said in my intro,

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00:04:39.900 --> 00:04:43.160

the number of people who've been killed in various conflicts

116

00:04:43.160 --> 00:04:44.380

around the world when you don't believe

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00:04:44.380 --> 00:04:46.540

that everybody has been recorded.

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00:04:46.540 --> 00:04:48.320

And so I reached out to him

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00:04:48.320 --> 00:04:50.370

when I was at midway through grad school

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00:04:50.370 --> 00:04:52.730

and I ended up interning there for the summer.

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00:04:52.730 --> 00:04:54.390

And in all of these sort of intervening years,

122

00:04:54.390 --> 00:04:56.730

I had sort of stayed in the periphery of the organization.

123

00:04:56.730 --> 00:04:58.800

It was called the Human Rights Data Analysis Group,

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00:04:58.800 --> 00:05:02.400

still is called that in fact, doing some consulting,

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00:05:02.400 --> 00:05:04.030

just sort of staying involved and interested

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00:05:04.030 --> 00:05:06.740

'cause I was really interested in what they did,

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00:05:06.740 --> 00:05:07.990

thought they do some really high impact

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00:05:07.990 --> 00:05:09.520

and interesting projects.

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00:05:09.520 --> 00:05:13.430

And so after Datapad was acquired,

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00:05:13.430 --> 00:05:16.820

I ended up doing part-time consulting for them again

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00:05:16.820 --> 00:05:19.010

and doing part-time data science consulting.

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00:05:19.010 --> 00:05:20.040

So through Datapad

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00:05:20.040 --> 00:05:22.250

I'd met quite a few people who were

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00:05:22.250 --> 00:05:25.230

sort of in the startup world and in sort of tech basically

135

00:05:25.230 --> 00:05:27.800

and did data science consulting for them.

136

00:05:27.800 --> 00:05:31.510

And I think it was around 2015 or so

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00:05:31.510 --> 00:05:33.340

and the years are sort of blending together

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00:05:33.340 --> 00:05:37.350

at this point for me, but they had a budget

139

00:05:37.350 --> 00:05:38.677
to hire me full-time at HRDAG,

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00:05:38.677 --> 00:05:41.250
and so I jumped on board and that was where I spent

141

00:05:42.360 --> 00:05:44.040
from whenever that point was again,

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00:05:44.040 --> 00:05:45.620
'cause I was sort of consulting

143

00:05:45.620 --> 00:05:46.880
at an increasing rate over time,

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00:05:46.880 --> 00:05:48.320
so it's hard to remember the exact date

145

00:05:48.320 --> 00:05:50.250
where I jumped on full-time.

146

00:05:50.250 --> 00:05:52.700
But from whenever that was roughly 2015

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00:05:52.700 --> 00:05:55.320
up until March of last year,

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00:05:55.320 --> 00:05:56.810
I was at HRDAG and there,

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00:05:56.810 --> 00:06:00.160
I led project on the United States.

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00:06:00.160 --> 00:06:02.640
So in the human rights world, there's often this criticism

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00:06:02.640 --> 00:06:06.460

that sort of human rights organizations are mostly based

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00:06:06.460 --> 00:06:08.160

in like the United States and Western Europe

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00:06:08.160 --> 00:06:10.260

and have a tendency to be looking outward

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00:06:10.260 --> 00:06:11.680

and sort of pointing all over the world

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00:06:11.680 --> 00:06:13.930

and saying, you people in developing countries

156

00:06:13.930 --> 00:06:17.093

like these are the things that, sorry, my dog,

157

00:06:18.700 --> 00:06:21.810

these are the abuses that you that you're all perpetrating.

158

00:06:21.810 --> 00:06:23.020

And we were thinking,

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00:06:23.020 --> 00:06:24.550

it was time for us to look inwards,

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00:06:24.550 --> 00:06:25.847

the United States has its own problems.

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00:06:25.847 --> 00:06:28.970

And so we thought, what projects could we do

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00:06:29.850 --> 00:06:31.970

that would speak to issues in the United States?

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00:06:31.970 --> 00:06:34.960

So usually, we look for projects where our skills

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00:06:34.960 --> 00:06:38.130

as statisticians and data scientists

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00:06:38.130 --> 00:06:40.380

actually give us sort of like unique,

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00:06:40.380 --> 00:06:41.213

I don't want to say advantage

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00:06:41.213 --> 00:06:43.890

like a unique perspective on a problem, right?

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00:06:43.890 --> 00:06:45.560

And so we ended up working

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00:06:45.560 --> 00:06:47.890

on the criminal justice system in the United States.

170

00:06:47.890 --> 00:06:49.350

So there, I started studying things

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00:06:49.350 --> 00:06:53.160

like predictive policing systems, risk assessment models

172

00:06:53.160 --> 00:06:56.210

that evaluate someone's likelihood of rearrest

173

00:06:56.210 --> 00:06:59.820

if that'll you use to then say, make recommendations

174

00:06:59.820 --> 00:07:02.830

about pretrial release, things like that.

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00:07:02.830 --> 00:07:06.330

And so I did that for about five, maybe six years,

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00:07:06.330 --> 00:07:08.830
somewhere in that range.

177

00:07:08.830 --> 00:07:11.850
I also became very involved in the algorithmic fairness,

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00:07:11.850 --> 00:07:14.093
accountability and transparency community,

179

00:07:15.380 --> 00:07:17.570
which I think we might talk about a little bit later.

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00:07:17.570 --> 00:07:19.960
And then last year on March 2nd,

181

00:07:19.960 --> 00:07:22.910
I began my new job at the University of Pennsylvania.

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00:07:22.910 --> 00:07:25.930
And I'll note that's an interesting time to start a new job.

183

00:07:25.930 --> 00:07:28.880
It was I think maybe one week exactly before the university

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00:07:28.880 --> 00:07:29.713
sort of shut down due to COVID.

185

00:07:29.713 --> 00:07:33.310
So yeah, interesting times, but sorry,

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00:07:33.310 --> 00:07:36.150
maybe that was longer than you expected, but my career path

187

00:07:36.150 --> 00:07:40.310
to get here has been a little bit of an adventure.

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00:07:40.310 --> 00:07:41.210

It's definitely non-linear

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00:07:41.210 --> 00:07:43.987

and so it was kind of thought I'd go through it all.

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00:07:45.816 --> 00:07:47.630

<v ->That was great to hear about, thank you.</v>

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00:07:47.630 --> 00:07:49.430

So we always hear about how women make

192

00:07:49.430 --> 00:07:51.860

up less than the workforce in STEM careers.

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00:07:51.860 --> 00:07:54.070

Were you always interested in pursuing a career

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00:07:54.070 --> 00:07:56.140

in STEM and if so, were there any points

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00:07:56.140 --> 00:07:58.140

in either your education or your career

196

00:07:58.140 --> 00:08:01.480

where you felt especially supported or not supported?

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00:08:01.480 --> 00:08:04.190

<v ->Yeah, I think I was always interested in STEM</v>

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00:08:04.190 --> 00:08:06.790

'cause as long as I can remember, even going back

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00:08:06.790 --> 00:08:09.967

to being a little kid, I always really liked math,

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00:08:09.967 --> 00:08:14.530

and I can remember being in high school and taking calculus

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00:08:14.530 --> 00:08:16.530

at the local community college over the summer

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00:08:16.530 --> 00:08:17.880

just cause I kind of wanted to get ahead.

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00:08:17.880 --> 00:08:19.900

I thought it'd be fun, I guess I was a big nerd,

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00:08:19.900 --> 00:08:20.733

I don't know.

205

00:08:21.760 --> 00:08:23.187

And it was the first time where I was like,

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00:08:23.187 --> 00:08:25.940

"Wow, this is awesome."

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00:08:25.940 --> 00:08:28.540

Because it wasn't just here's the algorithm,

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00:08:28.540 --> 00:08:31.690

you execute as a human to say like do long division.

209

00:08:31.690 --> 00:08:34.340

I felt like it was actually explaining why things happened

210

00:08:34.340 --> 00:08:36.380

and helping you to set up problems,

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00:08:36.380 --> 00:08:38.490

so you could address, you have set up the math

212

00:08:38.490 --> 00:08:40.360

so you could address a real world problem.

213

00:08:40.360 --> 00:08:41.837

And I remember just thinking like,

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00:08:41.837 --> 00:08:43.770

"Wow, this is what I wanna do."

215

00:08:43.770 --> 00:08:46.140

And so that was, I think maybe my junior year of high school

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00:08:46.140 --> 00:08:49.160

and basically from there, it was a pretty straight shot.

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00:08:49.160 --> 00:08:51.940

I went into college knowing I wanted to do math

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00:08:51.940 --> 00:08:53.370

about halfway through.

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00:08:53.370 --> 00:08:55.800

I took a statistics course and I sort of had another one

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00:08:55.800 --> 00:08:57.217

of those moments where I was like,

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00:08:57.217 --> 00:08:59.860

"Yup, this is the thing, I really like this."

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00:08:59.860 --> 00:09:01.750

I liked math as well.

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00:09:01.750 --> 00:09:04.340

I really was drawn to sort of statistics

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00:09:04.340 --> 00:09:07.930

because of the grounding in real world problems a little bit

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00:09:07.930 --> 00:09:11.450

more than math when you get up to some of the higher levels.

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00:09:11.450 --> 00:09:13.240

And of course, people do applied math, it's very grounded,

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00:09:13.240 --> 00:09:16.380

but in my experience, a statistics was sort of the path

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00:09:16.380 --> 00:09:17.700

that made sense for me.

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00:09:17.700 --> 00:09:19.323

So yeah, it's always been,

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00:09:20.210 --> 00:09:22.930

I think that was definitely always something I knew

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00:09:22.930 --> 00:09:24.270

I wanted to do.

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00:09:24.270 --> 00:09:25.803

In terms of feeling supported,

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00:09:27.843 --> 00:09:31.820

I think it's a little bit of both, to be honest.

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00:09:31.820 --> 00:09:35.140

I think I've always felt quite a bit of support from say

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00:09:35.140 --> 00:09:38.230

like my family and most of my professors,

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00:09:38.230 --> 00:09:41.070

and I felt that there are paths

237

00:09:41.070 --> 00:09:43.600

to having a career in this area.

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00:09:43.600 --> 00:09:46.000

But there have also been times where I felt like

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00:09:47.990 --> 00:09:49.280

I don't even know how to say this exactly,

240

00:09:49.280 --> 00:09:53.000

but like very unwelcome in the sense of

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00:09:54.920 --> 00:09:56.233

feeling like,

242

00:09:58.510 --> 00:09:59.880

I'm trying to word this delicately.

243

00:09:59.880 --> 00:10:01.840

So I guess I'll be blunt and say that

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00:10:01.840 --> 00:10:03.727

in 2017 I wrote an article, it was called,

245

00:10:03.727 --> 00:10:05.460

"Statistics We have a Problem."

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00:10:05.460 --> 00:10:06.970

And it was about my experience

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00:10:06.970 --> 00:10:09.390

with sexual harassment in statistics.

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00:10:09.390 --> 00:10:11.900

And if anyone wants to check that out,

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00:10:11.900 --> 00:10:14.010

you should feel more than welcome to do so,

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00:10:14.010 --> 00:10:15.430

it's still on the Internet.

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00:10:15.430 --> 00:10:17.530

It made somewhat of a splash in the field,

252

00:10:17.530 --> 00:10:22.050

I don't want to rehash all of the details of that here,

253

00:10:22.050 --> 00:10:24.863

I don't think this is exactly the time and place for that,

254

00:10:25.700 --> 00:10:27.920

but it's not because I'm shy about talking about it.

255

00:10:27.920 --> 00:10:31.010

But I guess what I would say, I think like big picture,

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00:10:31.010 --> 00:10:32.340

I felt like supported,

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00:10:32.340 --> 00:10:34.980

but then there are all these things that happen

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00:10:34.980 --> 00:10:37.700

along the way that make you maybe feel

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00:10:37.700 --> 00:10:41.590

like you might have to compromise some of your values

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00:10:41.590 --> 00:10:43.630

or integrity or put up with things

261

00:10:43.630 --> 00:10:47.970

that you just don't wanna put up with to make it

262

00:10:48.925 --> 00:10:49.870

in this career.

263

00:10:49.870 --> 00:10:52.420

And so in the article,

264

00:10:52.420 --> 00:10:53.450

I do talk about this a little bit,

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00:10:53.450 --> 00:10:54.630

that was one of the precipitate,

266

00:10:54.630 --> 00:10:55.980

there were some precipitating events

267

00:10:55.980 --> 00:10:58.270

involving sexual harassment where I did decide,

268

00:10:58.270 --> 00:11:02.510

okay, statistics or academic statistics is not going

269

00:11:02.510 --> 00:11:05.420

to be a place where I can thrive as a researcher

270

00:11:05.420 --> 00:11:09.070

and as a human who has to live among these colleagues,

271

00:11:09.070 --> 00:11:12.303

and so that was part of the reason that I left at one point.

272

00:11:17.300 --> 00:11:18.580

<v ->So what do you think can be done</v>

273

00:11:18.580 --> 00:11:20.710

to help women feel more supported in STEM careers

274

00:11:20.710 --> 00:11:23.760

in general, but particularly data science?

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00:11:23.760 --> 00:11:25.540

<v ->What do I think can help?</v>

276

00:11:25.540 --> 00:11:30.540

I think if you see that people are being treated poorly,

277

00:11:30.800 --> 00:11:32.230

even sort of minor things,

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00:11:32.230 --> 00:11:34.730

I think it helps to say something.

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00:11:34.730 --> 00:11:37.170

I mean, I know it can be embarrassing for the person

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00:11:37.170 --> 00:11:39.240

or for you to to step in and say something,

281

00:11:39.240 --> 00:11:42.300

but I think, simple things like, "Hey, I saw that,"

282

00:11:42.300 --> 00:11:44.270

or "You doing okay?"

283

00:11:44.270 --> 00:11:45.390

And these are sort of the minor things,

284

00:11:45.390 --> 00:11:47.790

I'm talking about the sort of harassment type of things

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00:11:47.790 --> 00:11:51.610

or even talking to folks who are behaving inappropriately

286

00:11:51.610 --> 00:11:55.390

and being like, "What I saw wasn't okay."

287

00:11:55.390 --> 00:11:57.170

I actually think that makes a big difference

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00:11:57.170 --> 00:12:00.470

because I think what people who maybe don't experience

289

00:12:00.470 --> 00:12:02.600

these things themselves don't understand,

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00:12:02.600 --> 00:12:04.540

it's not just the one thing that you saw,

291

00:12:04.540 --> 00:12:05.940

it's like all of these other things

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00:12:05.940 --> 00:12:07.950

that sort of accumulate over time

293

00:12:07.950 --> 00:12:11.390

and really just wear someone down and make them feel...

294

00:12:11.390 --> 00:12:12.480

And it's worse than unwelcome,

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00:12:12.480 --> 00:12:15.420

make them feel like this isn't a place

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00:12:15.420 --> 00:12:17.330

where they can thrive.

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00:12:17.330 --> 00:12:18.910

And so I think that's one thing

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00:12:18.910 --> 00:12:21.100

and I think there are bigger picture things too.

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00:12:21.100 --> 00:12:24.570

Like the things where I feel like we're doing pretty well,

300

00:12:24.570 --> 00:12:27.910

just panels and things like to encourage women

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00:12:27.910 --> 00:12:30.130

that this is a place where you can use your skills,

302

00:12:30.130 --> 00:12:34.476

you will be valued, having opportunities

303

00:12:34.476 --> 00:12:36.880

for people to learn more about the field.

304

00:12:36.880 --> 00:12:39.590

I think it's a little bit of everything, to be honest

305

00:12:41.870 --> 00:12:44.290

<v ->Definitely will be checking out your article.</v>

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00:12:44.290 --> 00:12:46.030

So your work on the development

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00:12:46.030 --> 00:12:47.740

of statistical and machine learning models

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00:12:47.740 --> 00:12:49.680

to tackle problems with social impact,

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00:12:49.680 --> 00:12:51.260

and I know that you're particularly interested

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00:12:51.260 --> 00:12:53.270

in applications to criminal justice.

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00:12:53.270 --> 00:12:55.340

Can you tell us about some of the specific projects

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00:12:55.340 --> 00:12:56.380

you've worked on where you felt

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00:12:56.380 --> 00:12:58.470

like you've made the biggest difference?

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00:12:58.470 --> 00:13:00.260

<v ->Yeah, so I think probably the project</v>

315

00:13:00.260 --> 00:13:02.010

that made the biggest difference came out

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00:13:02.010 --> 00:13:03.370

quite a few years ago at this point.

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00:13:03.370 --> 00:13:05.020

So it was in, I think 2016

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00:13:06.162 --> 00:13:08.450

and this is the project on predictive policing.

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00:13:08.450 --> 00:13:11.170

So predictive policing is essentially the idea

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00:13:11.170 --> 00:13:13.350

that you could use data

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00:13:13.350 --> 00:13:15.750

and it's usually things like police records of crime

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00:13:15.750 --> 00:13:19.410

to make predictions about who will commit a crime

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00:13:19.410 --> 00:13:21.350

in the future, sometimes who will be a victim

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00:13:21.350 --> 00:13:23.230

of a crime in the future, lumped together.

325

00:13:23.230 --> 00:13:25.680

It's kind of a strange thing to do in my opinion

326

00:13:25.680 --> 00:13:28.360

or where crime will occur in the future.

327

00:13:28.360 --> 00:13:29.850

And these things are actually pretty popular,

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00:13:29.850 --> 00:13:32.370

these sorts of models, some police departments make

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00:13:32.370 --> 00:13:33.880

their own if they're larger departments

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00:13:33.880 --> 00:13:36.750

but a lot of them buy tools

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00:13:36.750 --> 00:13:40.440

or software from companies that sell these sorts of things.

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00:13:40.440 --> 00:13:42.440

And so I wrote a paper

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00:13:42.440 --> 00:13:45.730

where we reproduced a predictive policing algorithm

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00:13:45.730 --> 00:13:47.280

that was published in Jaza,

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00:13:47.280 --> 00:13:49.450

so one of the big statistics journals,

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00:13:49.450 --> 00:13:53.980

and we applied it to data in from Oakland, California

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00:13:53.980 --> 00:13:57.580

to see what would happen if this algorithm had been

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00:13:57.580 --> 00:13:59.450

in the past applied there.

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00:13:59.450 --> 00:14:02.540

And so we actually, we did a few comparison.

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00:14:02.540 --> 00:14:04.890

So the idea is if police records

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00:14:04.890 --> 00:14:07.370

are not a representative sample of crime and in particular,

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00:14:07.370 --> 00:14:10.400

if they over represent say communities of color,

343

00:14:10.400 --> 00:14:13.130

then the algorithms will learn those disparate patterns

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00:14:13.130 --> 00:14:17.340

and be used to concentrate and reallocate policing back

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00:14:17.340 --> 00:14:19.650

to the locations where there were overrepresented

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00:14:19.650 --> 00:14:22.160

in the past and in particular, they could amplify,

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00:14:22.160 --> 00:14:25.020

or reproduce, or perpetuate historical racial bias

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00:14:25.020 --> 00:14:27.670

in policing and that was sort of the idea.

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00:14:27.670 --> 00:14:30.980

And so we needed to compare to like, where do we think?

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00:14:30.980 --> 00:14:33.920

So we used a dataset on drug crimes in Oakland,

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00:14:33.920 --> 00:14:35.360

I needed a data set to say like, okay,

352

00:14:35.360 --> 00:14:37.930

where do we think drug use is actually happening,

353

00:14:37.930 --> 00:14:38.890

seems like a reasonable thing to think about.

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00:14:38.890 --> 00:14:41.930

And so we used some public health data compared

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00:14:41.930 --> 00:14:46.030

it to census data, just to get a sense that if you look

356

00:14:46.030 --> 00:14:47.380

at this public health data,

357

00:14:48.720 --> 00:14:52.210

it seems like drug use is probably all over the place.

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00:14:52.210 --> 00:14:54.670

But then when you look at the historical police records

359

00:14:54.670 --> 00:14:57.200

of drug crime, it's really highly concentrated

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00:14:57.200 --> 00:14:59.880

in black communities and Hispanic communities.

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00:14:59.880 --> 00:15:02.500

And so when we ran the algorithm on that historical data

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00:15:02.500 --> 00:15:05.340

just like would be done if the algorithm had been applied

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00:15:05.340 --> 00:15:08.340

to this data and I'll give a caveat in a second there,

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00:15:08.340 --> 00:15:11.330

we saw that in fact, it would reproduce the historical bias

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00:15:11.330 --> 00:15:14.180

in policing and could be used to sort of perpetuate that.

366

00:15:15.070 --> 00:15:18.640

In practice, the algorithm that we applied,

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00:15:18.640 --> 00:15:19.550

they say they don't apply it

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00:15:19.550 --> 00:15:20.590

to drug crime data,

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00:15:20.590 --> 00:15:24.740

but I think the sort of broad overarching point remains.

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00:15:24.740 --> 00:15:28.860

And so I think this particular project had a

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00:15:28.860 --> 00:15:29.693

pretty big impact.

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00:15:29.693 --> 00:15:33.440

This was something that advocacy groups,

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00:15:33.440 --> 00:15:35.100

people who work in this space

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00:15:35.100 --> 00:15:38.510

in like less quantitative ways had been saying for awhile.

375

00:15:38.510 --> 00:15:39.970

And this is one of the things that I think is

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00:15:39.970 --> 00:15:43.450

really important if you want to work in, I don't know,

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00:15:43.450 --> 00:15:44.380

I don't actually love this term,

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00:15:44.380 --> 00:15:45.430

but like data for social good

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00:15:45.430 --> 00:15:47.520

or whatever that sort

380

00:15:47.520 --> 00:15:52.520

of field is called is that people who work in those areas

381

00:15:52.520 --> 00:15:54.620

in non-quantitative ways often, already kind of know

382

00:15:54.620 --> 00:15:56.300

what's going on, and it can be useful

383

00:15:56.300 --> 00:16:00.410

to give a sort of quantitative voice or rigor

384

00:16:00.410 --> 00:16:01.840

to some of those same concerns.

385

00:16:01.840 --> 00:16:02.917

So like look into it and be like,

386

00:16:02.917 --> 00:16:04.920

"Okay, this is what you think is going on."

387

00:16:04.920 --> 00:16:06.770

Let's investigate this using data,

388

00:16:06.770 --> 00:16:08.880

using the tools of data science, because this is

389

00:16:08.880 --> 00:16:12.120

what people in power, people who could make decisions listen

390

00:16:12.120 --> 00:16:13.533

to in a lot of ways.

391

00:16:15.226 --> 00:16:19.220

Let's see if what anecdotally you think is happening

392

00:16:19.220 --> 00:16:21.170

or might happen is born out by the data

393

00:16:25.930 --> 00:16:27.760

<v ->Just to follow up on that last part,</v>

394

00:16:27.760 --> 00:16:29.830

you mentioned that term, you said you don't like,

395

00:16:29.830 --> 00:16:33.110

could you clarify why or what should be used instead

396

00:16:33.110 --> 00:16:35.630

<v ->Just because I feel like it's too broad.</v>

397

00:16:35.630 --> 00:16:37.800

I feel like when you're saying social good,

398

00:16:37.800 --> 00:16:38.803

Who's social good?

399

00:16:39.895 --> 00:16:40.880

One thing that I've learned working

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00:16:40.880 --> 00:16:45.120

in the criminal justice area is that there are

401

00:16:45.120 --> 00:16:46.930

many people who work in this area

402

00:16:46.930 --> 00:16:48.830

who think what they're doing is for social good

403

00:16:48.830 --> 00:16:53.050

and it sort of a very broad spectrum of approaches

404

00:16:53.050 --> 00:16:55.210

to dealing with social issues.

405

00:16:55.210 --> 00:16:57.750

And in a lot of cases, people on both sides

406

00:16:57.750 --> 00:17:00.580

of the spectrum think they're basically doing data science

407

00:17:00.580 --> 00:17:02.360

for social good and have very different opinions

408

00:17:02.360 --> 00:17:04.520

about the world they'd like to create

409

00:17:04.520 --> 00:17:07.440

or how they'd like to get there, right?

410

00:17:07.440 --> 00:17:10.110

And so I just like to be kind of specific

411

00:17:10.110 --> 00:17:13.150

about the sorts of projects that I do

412

00:17:13.150 --> 00:17:16.140

and the types of lenses I take to looking at data

413

00:17:17.870 --> 00:17:20.860

just because when it doesn't say a whole lot,

414

00:17:20.860 --> 00:17:22.710

I guess when you say for social good.

415

00:17:25.370 --> 00:17:28.660

<v ->Makes sense, so in machine learning algorithms are said</v>

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00:17:28.660 --> 00:17:30.790

to be fair if the results are independent

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00:17:30.790 --> 00:17:34.080

of given variables, typically those considered sensitive,

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00:17:34.080 --> 00:17:36.240

can you talk a little bit about algorithmic bias

419

00:17:36.240 --> 00:17:38.680

and the impacts it can have on society?

420

00:17:38.680 --> 00:17:41.770

<v ->Yeah, sure, so what you listed is is one example</v>

421

00:17:41.770 --> 00:17:42.870

of a definition of fairness

422

00:17:42.870 --> 00:17:45.410

and that's one that I've written about in the past.

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00:17:45.410 --> 00:17:47.960

But it's one of many definitions.

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00:17:47.960 --> 00:17:50.460

So there's independence, there also equality

425

00:17:50.460 --> 00:17:53.670

of say false positive rates, equality of say accuracy,

426
00:17:53.670 --> 00:17:56.350
there's any number of ways you can dice it.

427
00:17:56.350 --> 00:17:58.530
And so when I was writing about this,

428
00:17:58.530 --> 00:18:00.510
this is back probably 2016, 2017.

429
00:18:00.510 --> 00:18:03.190
Also, this is really what sort of like the state

430
00:18:03.190 --> 00:18:05.500
of the field was, and I think it's useful to be thinking

431
00:18:05.500 --> 00:18:07.660
about different ways to measure inequality

432
00:18:07.660 --> 00:18:09.470
and what fairness might look like.

433
00:18:09.470 --> 00:18:10.630
But those really only take

434
00:18:10.630 --> 00:18:12.900
into account the model in isolation.

435
00:18:12.900 --> 00:18:17.010
And so I think since then, this field has expanded the scope

436
00:18:17.010 --> 00:18:21.870
of what it means to talk about a model behaving fairly,

437
00:18:21.870 --> 00:18:24.110
to think about sort of like upstream

438
00:18:24.110 --> 00:18:25.850

and downstream consequences as well.

439

00:18:25.850 --> 00:18:28.330

So will the model be used fairly, right?

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00:18:28.330 --> 00:18:29.500

If say it's a model that's used

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00:18:29.500 --> 00:18:33.250

to inform human decision makers, does that actually lead

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00:18:33.250 --> 00:18:37.350

to humans making decisions that are better or more fair

443

00:18:37.350 --> 00:18:40.960

or along any dimension that you might want like,

444

00:18:40.960 --> 00:18:43.250

is this moving us towards a world that we might like,

445

00:18:43.250 --> 00:18:47.280

is it moving us towards say in the criminal justice context,

446

00:18:47.280 --> 00:18:50.320

often people want to use predictive models

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00:18:50.320 --> 00:18:51.920

as part of a bail reform,

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00:18:51.920 --> 00:18:54.520

type of movement to reduce pretrial detention.

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00:18:54.520 --> 00:18:56.620

So is the introduction of model actually moving

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00:18:56.620 --> 00:18:58.030

us towards that goal of having

451

00:18:58.030 --> 00:19:00.150
fewer people incarcerated pre-trial?

452

00:19:00.150 --> 00:19:02.770
So sort of expanding the scope is I think one

453

00:19:02.770 --> 00:19:05.370
of the important things when we think about talking

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00:19:05.370 --> 00:19:06.450
about fair models, right?

455

00:19:06.450 --> 00:19:09.380
So I think doing these sorts of evaluations of the model

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00:19:09.380 --> 00:19:12.140
in isolation, it's an important component

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00:19:12.140 --> 00:19:16.830
because I think it's very unlikely that as you start

458

00:19:16.830 --> 00:19:18.430
sort of telescoping out if you have

459

00:19:18.430 --> 00:19:20.950
some sort of extreme unfairness say at one step

460

00:19:21.970 --> 00:19:23.730
that it's not going to get worse as time goes on,

461

00:19:23.730 --> 00:19:28.450
but it doesn't guarantee that the system is going

462

00:19:28.450 --> 00:19:29.750
to be working in a way that's moving

463

00:19:29.750 --> 00:19:31.440

you towards say larger social goals.

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00:19:31.440 --> 00:19:35.530

Like for example, reduced number of people in jail.

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00:19:35.530 --> 00:19:39.020

And also often when people think about the model

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00:19:39.020 --> 00:19:40.650

in isolation, don't think a whole lot

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00:19:40.650 --> 00:19:42.820

about the data generating process.

468

00:19:42.820 --> 00:19:45.320

So for example, what are the inputs to the model?

469

00:19:45.320 --> 00:19:46.180

Where did those come from?

470

00:19:46.180 --> 00:19:50.350

Are those measured in a way that is fair, right?

471

00:19:50.350 --> 00:19:52.720

So one of the examples I can think of is a project

472

00:19:52.720 --> 00:19:55.380

I did looking at the role of overbooking

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00:19:55.380 --> 00:19:57.950

on these recidivism prediction models.

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00:19:57.950 --> 00:19:59.880

So basically saying, okay, they take in a bunch

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00:19:59.880 --> 00:20:02.820

of inputs including person's criminal history

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00:20:02.820 --> 00:20:05.700

and what the police say, like the booking charges,

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00:20:05.700 --> 00:20:06.930

what the police say the person did

478

00:20:06.930 --> 00:20:08.610

before they've ever been convicted,

479

00:20:08.610 --> 00:20:09.980

before it's ever been tried,

480

00:20:09.980 --> 00:20:12.960

even for prosecutors really looked at those charges, right?

481

00:20:12.960 --> 00:20:16.430

How often are charges that are ultimately unsubstantiated

482

00:20:16.430 --> 00:20:19.023

by the court system used to,

483

00:20:20.420 --> 00:20:23.120

they are pushed through a model to result

484

00:20:23.120 --> 00:20:25.630

in a higher recommended level of supervision.

485

00:20:25.630 --> 00:20:27.700

So a higher level of being safe supervised

486

00:20:27.700 --> 00:20:31.370

by pretrial services, may be detention based only on charges

487

00:20:31.370 --> 00:20:32.780

that are ultimately unsubstantiated.

488

00:20:32.780 --> 00:20:34.840

So how often the charges the person isn't convicted

489

00:20:34.840 --> 00:20:37.770

or caused them to be recommended for higher,

490

00:20:37.770 --> 00:20:42.710

more sort of serious or punitive conditions.

491

00:20:42.710 --> 00:20:44.170

And the answer was actually pretty high,

492

00:20:44.170 --> 00:20:45.300

it was close to 30%.

493

00:20:45.300 --> 00:20:47.910

And so thinking about all these sorts of things.

494

00:20:47.910 --> 00:20:49.830

but where does the data come from?

495

00:20:49.830 --> 00:20:50.920

Could it be gamed, right?

496

00:20:50.920 --> 00:20:51.753

Could somebody be like,

497

00:20:51.753 --> 00:20:55.170

"Hey, we think this person's needs to be off the street."

498

00:20:55.170 --> 00:20:57.060

We could just sort of tag on some extra charges

499

00:20:57.060 --> 00:20:59.610

and that will then induce them to be recommended

500

00:20:59.610 --> 00:21:02.430

to say not be released, all these sorts of things,

501
00:21:02.430 --> 00:21:06.200
sort of considering both the human component and the model

502
00:21:06.200 --> 00:21:10.440
and sort of effects you wanna see in the real world,

503
00:21:10.440 --> 00:21:12.380
all together, it's complicated.

504
00:21:12.380 --> 00:21:13.760
There are a lot of moving pieces.

505
00:21:13.760 --> 00:21:16.910
I think each component it's important

506
00:21:16.910 --> 00:21:18.730
to evaluate say the model in isolation,

507
00:21:18.730 --> 00:21:20.923
but that's not the end of the story.

508
00:21:24.827 --> 00:21:26.940
<v ->And that's very interesting.</v>

509
00:21:26.940 --> 00:21:29.800
So many commercial companies prefer to keep the details

510
00:21:29.800 --> 00:21:32.280
of the algorithms they use confidential,

511
00:21:32.280 --> 00:21:33.510
there only seem to be a lot of standards

512
00:21:33.510 --> 00:21:35.320
for their construction and operation,

513
00:21:35.320 --> 00:21:37.220

do you see this changing anytime soon?

514

00:21:38.720 --> 00:21:39.885

<v ->No.</v>

515

00:21:39.885 --> 00:21:42.552

(laughs loudly)

516

00:21:43.495 --> 00:21:46.870

I think that's accurate and likely to continue.

517

00:21:52.570 --> 00:21:53.960

<v ->So I know that you're also interested</v>

518

00:21:53.960 --> 00:21:56.350

in some alternative interpretations of data.

519

00:21:56.350 --> 00:21:57.690

Can you tell us a bit about your work

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00:21:57.690 --> 00:21:59.470

in this area and some of the guidelines

521

00:21:59.470 --> 00:22:01.793

around exploring alternative interpretations?

522

00:22:04.200 --> 00:22:05.130

<v ->'Cause I can sort of take this back</v>

523

00:22:05.130 --> 00:22:06.880

to the predictive policing context, right?

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00:22:06.880 --> 00:22:09.760

People usually think of police records of crime is sort

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00:22:09.760 --> 00:22:11.520

of a ground truth measure

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00:22:11.520 --> 00:22:13.700

of where crime is occurring, right.

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00:22:13.700 --> 00:22:15.160

And that's sort of the premise

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00:22:15.160 --> 00:22:17.850

of using them to make predictions about future crime, right?

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00:22:17.850 --> 00:22:19.690

And so I think one of the alternative lenses is

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00:22:19.690 --> 00:22:21.710

that these are records

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00:22:21.710 --> 00:22:23.640

of where police are enforcing crimes

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00:22:23.640 --> 00:22:25.100

and sort of leaving it at that, right?

533

00:22:25.100 --> 00:22:28.110

Like this doesn't necessarily tell us a whole lot

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00:22:28.110 --> 00:22:31.780

about where crimes are occurring or where the problems are.

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00:22:31.780 --> 00:22:34.240

And so I think that's one example.

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00:22:34.240 --> 00:22:38.380

I think often the approach I take is to use

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00:22:38.380 --> 00:22:41.030

some of the same tools, say of data science

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00:22:41.030 --> 00:22:43.440

or sort of the type of tools that are used

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00:22:43.440 --> 00:22:44.890

in these areas and sort of trying to turn

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00:22:44.890 --> 00:22:46.040

it around in that way, like, okay,

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00:22:46.040 --> 00:22:49.870

say what would happen if we interpret this data differently?

542

00:22:49.870 --> 00:22:53.010

Or how can we use the same sort of mathematical tools

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00:22:53.010 --> 00:22:56.440

to like highlight issues with some population

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00:22:56.440 --> 00:22:58.690

who's often not represented?

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00:22:58.690 --> 00:23:01.920

So another example of this is a recent project I did

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00:23:01.920 --> 00:23:04.740

in collaboration with some researchers at the ACLU

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00:23:04.740 --> 00:23:09.740

and Eric Lofgren at Washington State University

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00:23:09.880 --> 00:23:11.530

and Nina Fefferman at University of Tennessee,

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00:23:11.530 --> 00:23:14.850

both epidemiologists to model the spread of COVID

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00:23:14.850 --> 00:23:17.350

within in between jails and communities.

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00:23:17.350 --> 00:23:18.750

The idea being that there are all these sorts

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00:23:18.750 --> 00:23:20.900

of epidemiological models out there going

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00:23:20.900 --> 00:23:23.990

on to make predictions and drive policy

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00:23:23.990 --> 00:23:25.036

about what we should be doing

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00:23:25.036 --> 00:23:29.930

to deal with what's going on with COVID

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00:23:29.930 --> 00:23:32.230

or this is especially true

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00:23:32.230 --> 00:23:34.550

during the beginning of the pandemic.

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00:23:34.550 --> 00:23:38.180

And although epidemiologists and public health folks were

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00:23:38.180 --> 00:23:39.718

sort of already ringing the alarm bells

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00:23:39.718 --> 00:23:43.300

that jails are often hotbeds of infection.

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00:23:43.300 --> 00:23:45.540

In fact, because of very poorest places,

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00:23:45.540 --> 00:23:48.980

people will travel back and forth between them quite a bit,

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00:23:48.980 --> 00:23:50.660

that can drive infection in the community.

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00:23:50.660 --> 00:23:53.200

We wanted to be able to use some of those tools

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00:23:53.200 --> 00:23:57.840

of computational epi to sort of highlight how big

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00:23:57.840 --> 00:23:59.690

of a problem this could be.

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00:23:59.690 --> 00:24:01.160

And so that sort of another,

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00:24:01.160 --> 00:24:03.120

I don't know if it's never another interpretation on data,

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00:24:03.120 --> 00:24:06.620

but it's sort of another, using those types of tools

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00:24:06.620 --> 00:24:08.780

like data-based or computational tools

571

00:24:08.780 --> 00:24:11.333

to highlight different problems in the world.

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00:24:14.854 --> 00:24:17.510

<v ->Great, so data missing from a data set is</v>

573

00:24:17.510 --> 00:24:19.610

something that data scientists frequently do

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00:24:19.610 --> 00:24:20.990

and something that you've noted

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00:24:20.990 --> 00:24:24.200

that you were often just as interested in if not,

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00:24:24.200 --> 00:24:25.060
can you tell us a little bit

577

00:24:25.060 --> 00:24:27.120
about how you approach this challenge?

578

00:24:27.120 --> 00:24:28.260
<v ->Yeah, so I can tell you a little bit</v>

579

00:24:28.260 --> 00:24:31.000
about the organization I worked for,

580

00:24:31.000 --> 00:24:32.890
the Human Rights Data Analysis Group,

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00:24:32.890 --> 00:24:34.740
usually abbreviated as HRDAG.

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00:24:34.740 --> 00:24:38.040
And while I was there how I approached this

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00:24:38.040 --> 00:24:40.160
because this has been like a large part of my career

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00:24:40.160 --> 00:24:43.160
at this point, I think it's actually pretty interesting.

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00:24:43.160 --> 00:24:47.080
So the idea there was that people in times

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00:24:47.080 --> 00:24:50.010
of say civil conflicts often it's really hard

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00:24:50.010 --> 00:24:52.010
to get good data on the number

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00:24:52.010 --> 00:24:54.573

of people who've been killed or disappeared.

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00:24:55.910 --> 00:24:56.980

The reasons are numerous.

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00:24:56.980 --> 00:24:58.860

It could be things like perpetrators

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00:24:58.860 --> 00:25:02.390

intentionally hiding bodies, just lack of infrastructure

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00:25:02.390 --> 00:25:06.530

to record and just things can be kind of hectic

593

00:25:07.840 --> 00:25:09.163

during a conflict, right.

594

00:25:10.240 --> 00:25:11.550

And the reason I think

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00:25:11.550 --> 00:25:14.490

that it's important to have a good sense on what's missing

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00:25:14.490 --> 00:25:17.680

from the data is that if you don't account for that,

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00:25:17.680 --> 00:25:20.170

you can tell very different stories.

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00:25:20.170 --> 00:25:22.060

The data can tell you very different

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00:25:22.060 --> 00:25:23.220

and misleading stories, right?

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00:25:23.220 --> 00:25:24.053

So for example,

601
00:25:24.053 --> 00:25:28.870
if we have say one perpetrator organization hiding bodies,

602
00:25:28.870 --> 00:25:31.150
for example, then you were to do some analysis

603
00:25:31.150 --> 00:25:33.430
on how many people were killed and by whom,

604
00:25:33.430 --> 00:25:34.740
you might find that the people who are

605
00:25:34.740 --> 00:25:36.560
actually killing way more people,

606
00:25:36.560 --> 00:25:39.240
so committing many more human rights abuses,

607
00:25:39.240 --> 00:25:42.160
had fewer recorded killings.

608
00:25:42.160 --> 00:25:43.370
And then you might conclude that they were

609
00:25:43.370 --> 00:25:44.963
actually responsible for fewer.

610
00:25:46.020 --> 00:25:47.440
It's also, I think important

611
00:25:47.440 --> 00:25:49.610
for say retrospective policy analysis.

612
00:25:49.610 --> 00:25:51.870
So if you have, for example,

613
00:25:51.870 --> 00:25:54.930

one region where there just hasn't been a lot

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00:25:54.930 --> 00:25:56.790

of documentation, perhaps it's more rural,

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00:25:56.790 --> 00:25:57.970

that's something that we see a lot,

616

00:25:57.970 --> 00:25:59.983

this sort of urban, rural bias.

617

00:26:01.000 --> 00:26:02.477

Then if you wanted to look at,

618

00:26:02.477 --> 00:26:05.100

"Hey, how did this policy change if it was say implemented

619

00:26:05.100 --> 00:26:06.910

across time and space differently,

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00:26:06.910 --> 00:26:08.840

the number of killings that were occurring."

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00:26:08.840 --> 00:26:13.170

You might also make untrue or erroneous conclusions

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00:26:13.170 --> 00:26:15.920

about what was effective because what you might be picking

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00:26:15.920 --> 00:26:18.760

up on is some combination

624

00:26:18.760 --> 00:26:21.440

of people just aren't being recorded as much over here.

625

00:26:21.440 --> 00:26:24.240

And maybe the impact was just different over time.

626

00:26:24.240 --> 00:26:25.202

And these are all things we've seen.

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00:26:25.202 --> 00:26:28.720

We've also seen things like, for example,

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00:26:28.720 --> 00:26:31.360

a project in Columbia that we were working on

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00:26:31.360 --> 00:26:33.400

and we got one data set on the number of killings,

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00:26:33.400 --> 00:26:35.720

and it was really pretty flat

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00:26:35.720 --> 00:26:37.800

over a very long stretch of time.

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00:26:37.800 --> 00:26:39.520

And we were like, "That doesn't make a lot of sense."

633

00:26:39.520 --> 00:26:42.090

from what we're hearing from and what we understand

634

00:26:42.090 --> 00:26:44.870

of the conflict and what our partners who live

635

00:26:44.870 --> 00:26:47.070

there have said it, that's not exactly true.

636

00:26:47.070 --> 00:26:49.320

There have been sort of spikes

637

00:26:49.320 --> 00:26:50.840

in violence during this conflict.

638

00:26:50.840 --> 00:26:51.673

What is this?

639

00:26:51.673 --> 00:26:53.010

And so as we dug into it more

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00:26:53.010 --> 00:26:55.780

and we spoke to the people in charge of the data,

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00:26:55.780 --> 00:26:57.077

they were basically like,

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00:26:57.077 --> 00:26:58.620

"Yeah, I mean, that makes perfect sense."

643

00:26:58.620 --> 00:26:59.810

This is an organization

644

00:26:59.810 --> 00:27:02.840

that actually investigates the killings.

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00:27:02.840 --> 00:27:06.060

And so they just have an organizational capacity

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00:27:06.060 --> 00:27:09.180

of X and not exactly X, right.

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00:27:09.180 --> 00:27:12.320

But you know about X and so if it exceeds that capacity,

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00:27:12.320 --> 00:27:14.180

well, those things don't make it into the database.

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00:27:14.180 --> 00:27:16.677

And so you look at it across time and you could conclude,

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00:27:16.677 --> 00:27:20.040

"Okay, this is like really a pretty steady conflict.

651
00:27:20.040 --> 00:27:21.460
There weren't really spikes and violence."

652
00:27:21.460 --> 00:27:22.810
But then if you try to correct for that

653
00:27:22.810 --> 00:27:25.690
and understand what's missing,

654
00:27:25.690 --> 00:27:29.820
you would instead find the sort of dynamics

655
00:27:29.820 --> 00:27:31.890
of the conflict were very different then

656
00:27:31.890 --> 00:27:32.723
than what you thought.

657
00:27:32.723 --> 00:27:35.507
And so the approach that we took at HRDAG

658
00:27:35.507 --> 00:27:36.950
and they still take

659
00:27:36.950 --> 00:27:38.300
and I still interested in this area though

660
00:27:38.300 --> 00:27:41.470
I haven't worked on it for about the past year is

661
00:27:41.470 --> 00:27:43.600
to basically do population estimation.

662
00:27:43.600 --> 00:27:45.820
Sometimes this is called capture recapture,

663
00:27:45.820 --> 00:27:48.080

though because of the context area,

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00:27:48.080 --> 00:27:50.640

that term can be a little bit misleading

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00:27:50.640 --> 00:27:53.580

because people are actually captured and we're not

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00:27:53.580 --> 00:27:56.700

we aren't like capturing humans and releasing them.

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00:27:56.700 --> 00:27:59.040

But if you're familiar with that terminology,

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00:27:59.040 --> 00:28:00.160

that's essentially the same,

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00:28:00.160 --> 00:28:02.640

it's the same statistical methodology.

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00:28:02.640 --> 00:28:05.570

And essentially, the idea is if you get multiple lists

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00:28:05.570 --> 00:28:09.050

of people who've been killed, you can look at the overlaps

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00:28:09.050 --> 00:28:13.010

among them to infer the number that were not recorded

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00:28:13.010 --> 00:28:14.020

on any of those lists.

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00:28:14.020 --> 00:28:15.640

So just to give some intuition for that.

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00:28:15.640 --> 00:28:16.800

So you had two lists

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00:28:16.800 --> 00:28:19.180
and both of them overlapped a whole lot.

677
00:28:19.180 --> 00:28:20.260
Well, then you might conclude,

678
00:28:20.260 --> 00:28:22.310
that's probably close to the whole universe

679
00:28:22.310 --> 00:28:23.700
of people who've been killed.

680
00:28:23.700 --> 00:28:26.040
Whereas if you have two lists and they barely overlap,

681
00:28:26.040 --> 00:28:27.030
you'd probably conclude

682
00:28:27.030 --> 00:28:29.070
that there are a whole lot more people

683
00:28:29.070 --> 00:28:31.100
in that population that you haven't recorded.

684
00:28:31.100 --> 00:28:34.570
And of course, that sort of logic is formalized

685
00:28:34.570 --> 00:28:35.600
in statistical models.

686
00:28:35.600 --> 00:28:37.010
It's not just us looking at overlaps

687
00:28:37.010 --> 00:28:39.450
and being like, "1,400," right?

688
00:28:39.450 --> 00:28:42.020

Like it's actually comes out of statistical models

689

00:28:42.020 --> 00:28:45.020

and those of course have their own assumptions

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00:28:46.923 --> 00:28:48.920

that you need to deal with.

691

00:28:48.920 --> 00:28:52.460

But in any case that's the general idea

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00:28:52.460 --> 00:28:54.460

to get a better sort of statistical handle

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00:28:54.460 --> 00:28:56.593

on what's missing from your data.

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00:28:59.548 --> 00:29:01.430

<v ->So you have worked on a number of very important</v>

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00:29:01.430 --> 00:29:03.180

and interesting projects.

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00:29:03.180 --> 00:29:05.190

So what are some of the most important lessons

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00:29:05.190 --> 00:29:07.340

you've learned so far in your career?

698

00:29:07.340 --> 00:29:08.610

And if you had to do things over,

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00:29:08.610 --> 00:29:10.110

what would you do differently?

700

00:29:11.372 --> 00:29:13.660

<v ->What are some of the most, let me see,</v>

701

00:29:13.660 --> 00:29:15.560

what are the most important lessons

702

00:29:15.560 --> 00:29:17.270

I've learned in my career?

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00:29:17.270 --> 00:29:19.500

I think, especially in the areas that I work

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00:29:19.500 --> 00:29:20.650

on when we're talking about data

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00:29:20.650 --> 00:29:22.453

that represents real people,

706

00:29:23.580 --> 00:29:26.880

I think treating the data with the respect it deserves

707

00:29:26.880 --> 00:29:29.750

like not sort of, I think it's often easy

708

00:29:29.750 --> 00:29:32.410

as data scientists to just get lost in like the X's

709

00:29:32.410 --> 00:29:36.630

and Y's in sort of it's just a database of numbers

710

00:29:36.630 --> 00:29:39.750

sort of abstracting away because to some extent,

711

00:29:39.750 --> 00:29:42.100

I think to do that sort work day in and day out,

712

00:29:42.100 --> 00:29:44.300

you kind of have to, but at the end of the day,

713

00:29:44.300 --> 00:29:46.390

I think remembering that each,

714

00:29:46.390 --> 00:29:48.337

say row in your dataset represents a real person

715

00:29:48.337 --> 00:29:50.600

and a real person's story

716

00:29:50.600 --> 00:29:53.560

and sort of treating it with the care it deserves

717

00:29:53.560 --> 00:29:57.640

when you say describe the data, when you talk about results,

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00:29:57.640 --> 00:30:01.110

when you make recommendations for what say to do

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00:30:01.110 --> 00:30:02.033

in the real world.

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00:30:03.126 --> 00:30:05.550

I think that really goes a long way

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00:30:05.550 --> 00:30:07.930

and that's something I've come to appreciate

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00:30:07.930 --> 00:30:10.593

more and more over time is that,

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00:30:12.050 --> 00:30:14.590

yes, when you say like write down a regression model

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00:30:14.590 --> 00:30:16.040

or something, there's something very abstract

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00:30:16.040 --> 00:30:19.370

about that, and that can help you to understand the world,

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00:30:19.370 --> 00:30:20.850

but at the end of the day, like the data is

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00:30:20.850 --> 00:30:22.140

representing real people.

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00:30:22.140 --> 00:30:24.030

And so having some care about that,

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00:30:24.030 --> 00:30:25.530

I think goes a really long way

730

00:30:28.510 --> 00:30:30.620

<v ->That's very important to keep in mind.</v>

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00:30:30.620 --> 00:30:32.780

So what is the best and worst career advice

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00:30:32.780 --> 00:30:33.830

you've ever received?

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00:30:37.269 --> 00:30:38.943

<v ->I don't think I've received a lot of career advice.</v>

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00:30:38.943 --> 00:30:40.530

And I'm gonna be honest,

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00:30:40.530 --> 00:30:43.080

maybe I should be asking for it more, I don't know.

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00:30:47.780 --> 00:30:49.700

I guess I'll say I can tell you what's been good

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00:30:49.700 --> 00:30:52.520

and bad about my own career strategy.

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00:30:52.520 --> 00:30:53.620

And I don't know if anyone's advice,

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00:30:53.620 --> 00:30:55.630

maybe this is extremely unadvisable.

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00:30:55.630 --> 00:30:56.463

I don't know,

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00:30:56.463 --> 00:31:01.463

but I don't think anyone's really told me not to do this.

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00:31:02.010 --> 00:31:04.360

And it's basically that I've kind of just followed

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00:31:04.360 --> 00:31:05.290

what I think is interesting.

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00:31:05.290 --> 00:31:06.123

Like you said, I've worked

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00:31:06.123 --> 00:31:07.970

on some pretty interesting projects.

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00:31:07.970 --> 00:31:11.050

I've been kind of less concerned

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00:31:11.050 --> 00:31:14.280

about say like career advancement

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00:31:14.280 --> 00:31:18.133

or clawing my way to the top of something.

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00:31:20.030 --> 00:31:21.780

I honestly think that served me really well.

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00:31:21.780 --> 00:31:26.650

Like I have really enjoyed the journey.

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00:31:26.650 --> 00:31:29.000

I really enjoy all the projects that I work on.

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00:31:30.450 --> 00:31:31.770

And so I think I have,

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00:31:31.770 --> 00:31:33.520

actually I have been encouraged

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00:31:33.520 --> 00:31:36.033

to do that in the past, maybe not explicitly.

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00:31:37.407 --> 00:31:38.280

Is that good advice?

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00:31:38.280 --> 00:31:39.480

Yeah, I don't know, right?

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00:31:39.480 --> 00:31:42.550

Like I think also as you start looking around it,

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00:31:42.550 --> 00:31:45.050

your peers, you might find if you take that approach

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00:31:45.050 --> 00:31:46.850

that people are say getting tenure

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00:31:46.850 --> 00:31:49.600

and you're kind of still drifting around.

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00:31:49.600 --> 00:31:51.350

And so if that's something that's important,

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00:31:51.350 --> 00:31:54.670

maybe a more linear path is a better choice,

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00:31:54.670 --> 00:31:56.050

but what I can say is I'm pretty happy

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00:31:56.050 --> 00:31:57.650

with how I've been doing things.

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00:32:00.990 --> 00:32:04.103

<v ->Well, it sounds like you've had a great career so far.</v>

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00:32:04.103 --> 00:32:05.030

<v ->Thanks.</v>

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00:32:05.030 --> 00:32:07.990

<v ->Based on your past projects and experiences,</v>

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00:32:07.990 --> 00:32:09.420

where do you see data science

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00:32:09.420 --> 00:32:11.920

and addiction research intersecting in the future?

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00:32:18.530 --> 00:32:20.510

<v ->I think data science really has the power</v>

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00:32:20.510 --> 00:32:25.255

to sort of touch all facets of any sort of science, right?

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00:32:25.255 --> 00:32:28.817

Any sort of social impact sort of areas where there's data

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00:32:28.817 --> 00:32:31.760

and there's clearly data in this area,

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00:32:31.760 --> 00:32:32.970

I guess just sort of going back to one

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00:32:32.970 --> 00:32:35.090

of the things I was saying earlier,

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00:32:35.090 --> 00:32:37.330
the way that drug addiction

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00:32:37.330 --> 00:32:38.780
and drug use has been criminalized.

778

00:32:38.780 --> 00:32:41.900
I think it's reasonable to think that the data

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00:32:41.900 --> 00:32:43.520
that sort of traditional perspectives

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00:32:43.520 --> 00:32:45.340
on the data maybe aren't the full story.

781

00:32:45.340 --> 00:32:49.680
And so using that data or using other perspectives

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00:32:49.680 --> 00:32:52.530
to look at it will be an important direction to go.

783

00:32:52.530 --> 00:32:54.920
It's also very likely missing a whole lot of things.

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00:32:54.920 --> 00:32:56.750
So thinking about what is missing,

785

00:32:56.750 --> 00:32:58.100
what's systematically missing,

786

00:32:58.100 --> 00:32:59.430
how might that change the story

787

00:32:59.430 --> 00:33:03.350
that you're pulling out of the data will be

788

00:33:03.350 --> 00:33:05.483

an important avenue going forward.

789

00:33:08.790 --> 00:33:09.733

<v ->Are there any areas you would</v>

790

00:33:09.733 --> 00:33:11.773

like to see this research expanded?

791

00:33:12.870 --> 00:33:15.660

<v ->Data science or addiction research?</v>

792

00:33:15.660 --> 00:33:17.560

<v ->Data science for addiction research.</v>

793

00:33:18.970 --> 00:33:21.310

<v ->Honestly, this is not really my area of expertise,</v>

794

00:33:21.310 --> 00:33:22.143

so I will leave

795

00:33:22.143 --> 00:33:24.430

that to the folks who know what they're talking about.

796

00:33:24.430 --> 00:33:28.930

I think that's always a good idea to not talk

797

00:33:28.930 --> 00:33:30.620

when you don't know what you're talking about.

798

00:33:30.620 --> 00:33:32.400

So I'll exercise that

799

00:33:33.387 --> 00:33:36.100

<v ->Are there any areas though for, I guess, data science</v>

800

00:33:36.100 --> 00:33:38.420

in general, that you'd like to see expanded then?

801

00:33:39.850 --> 00:33:41.610

<v ->I mean, I really like this area</v>

802

00:33:41.610 --> 00:33:44.930

of data science where you kind of...

803

00:33:44.930 --> 00:33:49.930

Data science and the advocacy world, data science to sort of

804

00:33:50.180 --> 00:33:53.170

like more of more, this is a noxious word,

805

00:33:53.170 --> 00:33:55.630

so be prepared but like bespoke type of analysis

806

00:33:55.630 --> 00:33:57.230

where you're like, "Hey, let's take

807

00:33:57.230 --> 00:33:58.440

on this like very kind of niche

808

00:33:58.440 --> 00:34:00.780

and small question and like throw like the

809

00:34:00.780 --> 00:34:02.770

most powerful tools at it to see what we can get."

810

00:34:02.770 --> 00:34:05.110

I think for a while we've been sort of focused

811

00:34:05.110 --> 00:34:10.110

on scaling things and small batch, artisanal data science,

812

00:34:10.500 --> 00:34:12.003

I think is pretty cool.

813

00:34:15.930 --> 00:34:17.800

<v ->And lastly, are there any areas</v>

814

00:34:17.800 --> 00:34:20.283

you'd like to see NIH doing in data science?

815

00:34:23.740 --> 00:34:25.280

<v ->I mean, everything, right?</v>

816

00:34:25.280 --> 00:34:27.480

I don't know if I can narrow it down really.

817

00:34:30.700 --> 00:34:32.990

<v ->That's fine, it's been great talking to you this morning.</v>

818

00:34:32.990 --> 00:34:34.287

We really enjoyed hearing about your career

819

00:34:34.287 --> 00:34:37.070

and your experiences and your advice and everything.

820

00:34:37.070 --> 00:34:37.903

So thank you so much,

821

00:34:37.903 --> 00:34:39.600

I wanna have everyone clap for you,

822

00:34:39.600 --> 00:34:41.210

I know it's hard to do in this virtual environment.

823

00:34:41.210 --> 00:34:42.664

<v ->Thank you.</v>

824

00:34:42.664 --> 00:34:45.000

<v ->And next, we're gonna be having a presentation</v>

825

00:34:45.000 --> 00:34:48.250

from Dr. Brenda Curtis from the NIDA IRP.

826

00:34:48.250 --> 00:34:50.130

So Dr. Brenda Curtis is the chief

827

00:34:50.130 --> 00:34:52.620

of technology and translational research unit

828

00:34:52.620 --> 00:34:54.950

of the NIDA intramural research program.

829

00:34:54.950 --> 00:34:57.030

She earned both a bachelor's degree in biology

830

00:34:57.030 --> 00:34:58.233

and a master's degree in public health

831

00:34:58.233 --> 00:35:00.050

from the University of Illinois

832

00:35:00.050 --> 00:35:01.150

and then obtained her doctorate

833

00:35:01.150 --> 00:35:03.560

in communications from the University of Pennsylvania

834

00:35:03.560 --> 00:35:05.230

where she most recently held the appointment

835

00:35:05.230 --> 00:35:07.520

of assistant professor of psychology

836

00:35:07.520 --> 00:35:08.810

and psychiatry addictions

837

00:35:08.810 --> 00:35:10.730

at the Perelman School of Medicine.

838

00:35:10.730 --> 00:35:12.750

Dr. Curtis also completed a fellowship

839

00:35:12.750 --> 00:35:13.870

in the Fordham University,

840

00:35:13.870 --> 00:35:16.120

HIV and Drug Abuse Prevention Research

841

00:35:16.120 --> 00:35:17.740

Ethics Training Institute.

842

00:35:17.740 --> 00:35:18.990

Her training in public health

843

00:35:18.990 --> 00:35:20.790

and health communication allows her

844

00:35:20.790 --> 00:35:22.470

to employ a public health approach

845

00:35:22.470 --> 00:35:24.220

by using effective communication techniques

846

00:35:24.220 --> 00:35:26.650

to ensure recruitment and retention rates are achieved.

847

00:35:26.650 --> 00:35:29.900

Her research focuses translational leveraging social media

848

00:35:29.900 --> 00:35:32.390

and big data methodology to form the development,

849

00:35:32.390 --> 00:35:34.300

evaluation, and implementation

850

00:35:34.300 --> 00:35:36.750

of technology-based tools that address substance use

851
00:35:36.750 --> 00:35:39.530
and related conditions such as HIV&AIDS.

852
00:35:39.530 --> 00:35:41.580
Dr. Curtis employees multiple methodologies

853
00:35:41.580 --> 00:35:43.970
to facilitate the flow of scientific discovery

854
00:35:43.970 --> 00:35:45.780
to practical application allowing

855
00:35:45.780 --> 00:35:48.200
her to not only reach underserved populations,

856
00:35:48.200 --> 00:35:49.590
but to design health monitoring

857
00:35:49.590 --> 00:35:51.800
and behavioral change interventions

858
00:35:51.800 --> 00:35:54.610
that are user-centered, inclusive, and evidence-based.

859
00:35:54.610 --> 00:35:57.860
So please join me in welcoming Dr. Brenda Curtis,

860
00:35:57.860 --> 00:35:59.090
virtual applause here,

861
00:35:59.090 --> 00:36:01.480
and you can go ahead and share your sides now.

862
00:36:01.480 --> 00:36:02.313
<v ->0kay.</v>

863
00:36:08.320 --> 00:36:10.640

So thank you for inviting me

864

00:36:10.640 --> 00:36:14.330

and I'm gonna walk you through kind of start off

865

00:36:14.330 --> 00:36:18.223

with the why I do this and kind of how I got here.

866

00:36:20.070 --> 00:36:21.323

Let me move.

867

00:36:22.490 --> 00:36:23.770

Okay, slides moving?

868

00:36:23.770 --> 00:36:25.120

Yes, okay.

869

00:36:25.120 --> 00:36:29.790

So I am from a small town in East St. Louis, Illinois.

870

00:36:29.790 --> 00:36:31.160

It's across from Illinois.

871

00:36:31.160 --> 00:36:34.540

So sorry, across from Missouri, St. Louis.

872

00:36:34.540 --> 00:36:35.373

And it's

873

00:36:43.630 --> 00:36:44.893

just got a warning,

874

00:36:49.220 --> 00:36:51.673

had arise,

875

00:36:53.424 --> 00:36:54.757

I'm getting RMS.

876

00:36:58.420 --> 00:37:00.570

<v ->It sounds like your audio is cutting out.</v>

877

00:37:04.660 --> 00:37:05.493

<v ->Interesting.</v>

878

00:37:13.970 --> 00:37:15.213

<v Susan>Your Internet, girl.</v>

879

00:37:20.780 --> 00:37:22.380

Hello.

880

00:37:22.380 --> 00:37:23.220

<v ->We can still hear you,</v>

881

00:37:23.220 --> 00:37:25.847

it sounds like audio is cutting out little bit.

882

00:37:25.847 --> 00:37:27.851

<v ->Yeah, I lost Internet connection, I will do it.</v>

883

00:37:27.851 --> 00:37:28.690

<v Susan>Okay.</v>

884

00:37:28.690 --> 00:37:30.670

<v ->Let me try again.</v>

885

00:37:30.670 --> 00:37:31.503

<v Susan>Yeah.</v>

886

00:37:31.503 --> 00:37:33.303

<v ->And hopefully, this time it will work.</v>

887

00:37:36.090 --> 00:37:37.890

So I'm from a small town

888

00:37:37.890 --> 00:37:41.520

and in a town that was having really high poverty

889

00:37:41.520 --> 00:37:43.440

and really high crime.

890

00:37:43.440 --> 00:37:47.600

And when I started off in it,

891

00:37:47.600 --> 00:37:50.640

I left to go to undergrad and my undergrad degree was

892

00:37:50.640 --> 00:37:54.293

in biology and I planned on being a doctor.

893

00:37:56.310 --> 00:37:58.140

I kind of didn't really have

894

00:37:58.140 --> 00:38:00.930

or know of many like career paths.

895

00:38:00.930 --> 00:38:04.110

I knew doctor, lawyer, accountant, things like that.

896

00:38:04.110 --> 00:38:06.560

And so I went to the University of Illinois

897

00:38:06.560 --> 00:38:10.623

in Urbana-Champaign and wanted to go pre-med.

898

00:38:11.620 --> 00:38:14.690

I had always my whole life been interested in science,

899

00:38:14.690 --> 00:38:16.660

I can not remember a name to save my life

900

00:38:16.660 --> 00:38:19.400

and I cannot remember a date to save my life.

901

00:38:19.400 --> 00:38:23.210

But math was this universal language to me,

902

00:38:23.210 --> 00:38:27.220

and my mom, she had three children,

903

00:38:27.220 --> 00:38:29.280

she had her first child

904

00:38:29.280 --> 00:38:31.120

while she was in high school

905

00:38:31.120 --> 00:38:33.730

and had to drop out and then had to go back

906

00:38:33.730 --> 00:38:34.913

to get to finish.

907

00:38:37.108 --> 00:38:41.700

I'm the youngest and she was trying to go back to school

908

00:38:41.700 --> 00:38:44.150

so that she could get a promotion at work.

909

00:38:44.150 --> 00:38:48.700

And so had three children, managing a full-time job,

910

00:38:48.700 --> 00:38:52.670

a house and everything, she would leave her homework out

911

00:38:52.670 --> 00:38:55.810

and as to be a great daughter, be supportive,

912

00:38:55.810 --> 00:38:58.590

I would sit there and sometimes I would just do

913

00:38:58.590 --> 00:38:59.860

her homework for her.

914

00:38:59.860 --> 00:39:00.747

And she'd come and be like,

915

00:39:00.747 --> 00:39:02.040

"You're not supposed to do my homework,"

916

00:39:02.040 --> 00:39:04.070

but then of course she'd be extremely happy,

917

00:39:04.070 --> 00:39:06.770

but so I would always do her homework for her.

918

00:39:06.770 --> 00:39:08.880

And I just kind of had a knack for math,

919

00:39:08.880 --> 00:39:10.760

and I used to do like math competitions

920

00:39:10.760 --> 00:39:12.270

and different things.

921

00:39:12.270 --> 00:39:14.660

So when I went away for undergrad

922

00:39:14.660 --> 00:39:17.680

and I mean, I grew up in poverty and I grew up poor,

923

00:39:17.680 --> 00:39:19.443

but a lot of people do.

924

00:39:21.000 --> 00:39:24.070

When I came back, I saw a community

925

00:39:24.070 --> 00:39:26.320

that was being devastated.

926
00:39:26.320 --> 00:39:28.640
We've always had poverty that was there,

927
00:39:28.640 --> 00:39:32.250
but crack cocaine, heroin took over

928
00:39:32.250 --> 00:39:34.580
and we had always had alcohol problems.

929
00:39:34.580 --> 00:39:38.470
It really took over and the drugs fit into the crime.

930
00:39:38.470 --> 00:39:42.910
And I initially, when I came out, I was doing work

931
00:39:42.910 --> 00:39:45.340
at Washington University in St. Louis

932
00:39:45.340 --> 00:39:47.430
in one other transplant units.

933
00:39:47.430 --> 00:39:52.220
And I would hear from family and friends

934
00:39:52.220 --> 00:39:57.050
about a gang shootout or some other problems.

935
00:39:57.050 --> 00:39:58.193
And then I would be called

936
00:39:58.193 --> 00:40:01.630
into the hospital to receive organs.

937
00:40:01.630 --> 00:40:04.890
And it kind of was a connection

938
00:40:04.890 --> 00:40:07.560

that I was having a hard time making.

939

00:40:07.560 --> 00:40:12.560

And I decided to leave doing research at Wash U

940

00:40:13.640 --> 00:40:18.640

and to work at a drug treatment center doing HIV counseling

941

00:40:19.090 --> 00:40:20.763

as well as drug treatment work.

942

00:40:21.880 --> 00:40:24.180

I wanted to give back to my community.

943

00:40:24.180 --> 00:40:27.250

And so my history kinda always starts from that,

944

00:40:27.250 --> 00:40:31.670

figuring out a way that I could give back to my community.

945

00:40:31.670 --> 00:40:35.170

One of the biggest problems that I saw is

946

00:40:35.170 --> 00:40:37.750

that no one was coming to help us

947

00:40:37.750 --> 00:40:40.970

like East St. Louis still today as you saw

948

00:40:40.970 --> 00:40:43.750

in the other slide, we have the highest crime rate

949

00:40:43.750 --> 00:40:46.840

and the community is just devastated

950

00:40:46.840 --> 00:40:48.900

and there wasn't any one coming to solve those,

951

00:40:48.900 --> 00:40:51.970

we weren't getting Gates money or some other foundation.

952

00:40:51.970 --> 00:40:55.250

And definitely the state of Illinois was not coming.

953

00:40:55.250 --> 00:40:56.990

And there's lots of communities

954

00:40:56.990 --> 00:40:58.600

like East St. Louis across the US.

955

00:40:58.600 --> 00:41:01.160

And so I want it to figure out ways

956

00:41:01.160 --> 00:41:03.250

to get people the information

957

00:41:03.250 --> 00:41:05.740

and specifically around prevention and treatment.

958

00:41:05.740 --> 00:41:09.160

And I felt like and I still do today feel like the Internet

959

00:41:09.160 --> 00:41:13.623

and smartphones are ways to track, deliver,

960

00:41:15.013 --> 00:41:17.650

and do interventions as well as prevention.

961

00:41:17.650 --> 00:41:19.670

And one of the good things about smart phones is

962

00:41:19.670 --> 00:41:22.330

that everyone kind of has them.

963

00:41:22.330 --> 00:41:25.380

And so when you look at national numbers,

964

00:41:25.380 --> 00:41:27.562
you see that there's like 96%

965

00:41:27.562 --> 00:41:29.890
of the us population has a cell phone,

966

00:41:29.890 --> 00:41:32.103
about 81% have a smartphone.

967

00:41:33.010 --> 00:41:35.580
And we started looking across men and women

968

00:41:35.580 --> 00:41:38.160
that those numbers are similar.

969

00:41:38.160 --> 00:41:40.820
Yes, more younger people have a smartphone,

970

00:41:40.820 --> 00:41:43.040
but even as you get older

971

00:41:43.040 --> 00:41:44.903
we have greater populations,

972

00:41:44.903 --> 00:41:47.360
a good percentage have a smartphone.

973

00:41:47.360 --> 00:41:49.620
The age that I'm typically working in

974

00:41:49.620 --> 00:41:50.770
'cause I'm working in the clinic,

975

00:41:50.770 --> 00:41:53.810
it's kind of that 30 to 49 age range.

976

00:41:53.810 --> 00:41:55.780

And a sizable percentage

977

00:41:55.780 --> 00:41:58.143

of them around 92% have a smartphone.

978

00:41:59.240 --> 00:42:01.137

And when we look at, this is the Pew data,

979

00:42:01.137 --> 00:42:03.520

and these are the three ethnicities

980

00:42:03.520 --> 00:42:05.930

that they have represented.

981

00:42:05.930 --> 00:42:08.460

We get pretty good coverage across

982

00:42:10.080 --> 00:42:11.170

having a smartphone.

983

00:42:11.170 --> 00:42:15.810

Where we see lots of disparities is when we start dealing

984

00:42:15.810 --> 00:42:17.210

with the education level.

985

00:42:17.210 --> 00:42:19.790

And that's kind of where we start falling off

986

00:42:19.790 --> 00:42:22.263

and seeing less smartphone ownership.

987

00:42:23.834 --> 00:42:28.270

The smartphones today are super computers of yesterday.

988

00:42:28.270 --> 00:42:29.763

And we can do a lot on there.

989

00:42:29.763 --> 00:42:31.650

We can not only track,

990

00:42:31.650 --> 00:42:34.940

but we can deliver and we can monitor.

991

00:42:34.940 --> 00:42:39.080

And so that was kind of how I started was like,

992

00:42:39.080 --> 00:42:41.240

okay I want to get...

993

00:42:41.240 --> 00:42:42.470

No one's coming to help me,

994

00:42:42.470 --> 00:42:44.480

I kind of felt like I'll help my communities.

995

00:42:44.480 --> 00:42:48.387

And so how can I be a person be it to get treatment

996

00:42:48.387 --> 00:42:50.533

and information out there.

997

00:42:51.800 --> 00:42:55.240

And so one of the questions that I started

998

00:42:55.240 --> 00:42:57.393

with was kind of online surveillance.

999

00:42:58.340 --> 00:43:00.700

I initially started with looking at how drugs emerge,

1000

00:43:00.700 --> 00:43:02.440

like synthetic marijuana

1001
00:43:02.440 --> 00:43:04.720
that was taking a hold in communities.

1002
00:43:04.720 --> 00:43:08.160
And then I moved to like the online surveillance of alcohol.

1003
00:43:08.160 --> 00:43:10.910
And alcohol, it was when I got my first R01 grant

1004
00:43:10.910 --> 00:43:13.270
from NIDA, I am a NIDA baby starting

1005
00:43:13.270 --> 00:43:16.970
with a diversity supplement that I met Albert at that time.

1006
00:43:16.970 --> 00:43:19.357
But when I got my first R01,

1007
00:43:21.154 --> 00:43:23.943
the question was looking at social media language.

1008
00:43:25.300 --> 00:43:27.250
I was doing a clinic study,

1009
00:43:27.250 --> 00:43:30.170
but we wanted to test some of these questions out.

1010
00:43:30.170 --> 00:43:34.580
And so we decided to look at doing a sample online.

1011
00:43:34.580 --> 00:43:39.000
Twitter language is free, it's public.

1012
00:43:39.000 --> 00:43:40.823
And we just had a simple question,

1013
00:43:43.133 --> 00:43:47.682

can Twitter reuse to predict excessive alcohol consumption

1014

00:43:47.682 --> 00:43:49.410

at a county level?

1015

00:43:49.410 --> 00:43:52.670

So instead of doing an individual, we are looking

1016

00:43:52.670 --> 00:43:54.760

at the county level.

1017

00:43:54.760 --> 00:43:58.143

And we did a simple study.

1018

00:43:59.934 --> 00:44:01.060

We did the random sample,

1019

00:44:01.060 --> 00:44:03.690

we collected on 1% of Twitter posts

1020

00:44:03.690 --> 00:44:06.233

and then we collected national survey data.

1021

00:44:08.460 --> 00:44:11.120

One of the reasons I'm asking, can we use Twitter is

1022

00:44:11.120 --> 00:44:12.780

because it's cheaper.

1023

00:44:12.780 --> 00:44:15.550

National surveys are expensive.

1024

00:44:15.550 --> 00:44:17.490

And so one of the things we did was

1025

00:44:17.490 --> 00:44:19.820

we took some national survey data.

1026

00:44:19.820 --> 00:44:23.350

We used the behavioral risk factor surveillance system

1027

00:44:23.350 --> 00:44:26.740

across the US including,

1028

00:44:26.740 --> 00:44:31.040

Puerto Rico, DC, and the Virgin, Islands and womp.

1029

00:44:31.040 --> 00:44:33.570

We also collected demographic data

1030

00:44:33.570 --> 00:44:35.760

from the US Census Bureau,

1031

00:44:35.760 --> 00:44:39.583

and we use data from the American Community Survey.

1032

00:44:40.490 --> 00:44:42.240

So one of the questions we wanted to know

1033

00:44:42.240 --> 00:44:44.840

at the basic level is can we out predict

1034

00:44:46.550 --> 00:44:50.000

kind of the the survey data that the information

1035

00:44:50.000 --> 00:44:55.000

that we have already about who is reporting substance use?

1036

00:44:56.040 --> 00:44:57.620

And so we wanted to know,

1037

00:44:57.620 --> 00:45:01.240

could we use social media language to do that?

1038

00:45:01.240 --> 00:45:05.190

And the answer was, yes, we could out predict

1039

00:45:06.320 --> 00:45:08.920

just kind of the social demographic variables

1040

00:45:09.780 --> 00:45:11.520

that the survey data.

1041

00:45:11.520 --> 00:45:14.480

So we did some fancy things where you have all the data,

1042

00:45:14.480 --> 00:45:17.317

you train the model, and then you test it, you hold out

1043

00:45:17.317 --> 00:45:20.173

and you test it on another set of the model.

1044

00:45:21.810 --> 00:45:24.540

I mean, I'm not so interested in, can we,

1045

00:45:24.540 --> 00:45:27.020

that's the first question, always, can you?

1046

00:45:27.020 --> 00:45:30.230

But I'm always more interested in what do I gain about it?

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00:45:30.230 --> 00:45:34.140

Because my background, my undergrad was in biology.

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00:45:34.140 --> 00:45:37.350

My master's is in public health

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00:45:37.350 --> 00:45:40.150

and my PhD is in health communication.

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00:45:40.150 --> 00:45:43.050

So I'm kind of always looking from this lens

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00:45:43.050 --> 00:45:47.340
of from a public health and a health communication person,

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00:45:47.340 --> 00:45:50.180
how they can use this information

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00:45:50.180 --> 00:45:55.180
to inform treatment, prevention, and message delivery,

1054
00:45:55.780 --> 00:45:58.193
and message persuasiveness to individuals,

1055
00:45:59.330 --> 00:46:02.810
looking to how to change behavior that way.

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00:46:02.810 --> 00:46:05.250
So one of the questions is like counties

1057
00:46:05.250 --> 00:46:07.380
that had higher levels

1058
00:46:07.380 --> 00:46:10.903
of excessive drinking compared to counties with lower.

1059
00:46:12.200 --> 00:46:14.040
And one of the things that we see is

1060
00:46:14.040 --> 00:46:16.740
that when we look at the language that's used,

1061
00:46:16.740 --> 00:46:19.760
in counties that have higher levels of excessive drinking,

1062
00:46:19.760 --> 00:46:23.370
they're talking about sports, things like hockey,

1063
00:46:23.370 --> 00:46:25.400

they're talking about going to festivals

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00:46:25.400 --> 00:46:28.260
and art films, and folk films.

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00:46:28.260 --> 00:46:29.623
Of course, they talk about drinking,

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00:46:29.623 --> 00:46:33.640
they're talking about higher education type things,

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00:46:33.640 --> 00:46:36.470
research papers, projects, assignments.

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00:46:36.470 --> 00:46:39.960
And of course, they're talking about going out on weekends.

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00:46:39.960 --> 00:46:43.110
Now, if anyone's wondering, a lot of times in society,

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00:46:43.110 --> 00:46:46.480
we have this view of people who,

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00:46:46.480 --> 00:46:49.030
certain populations who have a higher levels

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00:46:49.030 --> 00:46:49.960
of excessive drinking.

1073

00:46:49.960 --> 00:46:51.410
And hopefully, I think you're saying

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00:46:51.410 --> 00:46:54.360
that if you had that view that they were black

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00:46:54.360 --> 00:46:57.010
and brown populations, the answer's, no.

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00:46:57.010 --> 00:46:59.220

We have higher rates of drinking are

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00:46:59.220 --> 00:47:03.330

in higher income, higher educated communities.

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00:47:03.330 --> 00:47:04.737

And so that's always interesting

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00:47:04.737 --> 00:47:07.170

'cause people sometimes have preconceived notion

1080

00:47:08.474 --> 00:47:12.320

about what we see because of stigma and stereotypes.

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00:47:12.320 --> 00:47:15.930

But in here, the language kind of speaks for itself.

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00:47:15.930 --> 00:47:19.320

In areas where we have less excessive drinking,

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00:47:19.320 --> 00:47:23.360

people are talking about religion, prayer, God,

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00:47:23.360 --> 00:47:25.470

doing things with family.

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00:47:25.470 --> 00:47:29.310

So one person could say, well, okay, so how is this useful?

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00:47:29.310 --> 00:47:32.660

Well, if you're in a community, you could be looking

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00:47:32.660 --> 00:47:36.150

at Twitter or you could know what's going on with festivals,

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00:47:36.150 --> 00:47:38.540

what sports activities, and you could send out

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00:47:38.540 --> 00:47:41.670

as a public health specialist or communicator,

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00:47:41.670 --> 00:47:45.770

you could send out harm reduction information on Twitter.

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00:47:45.770 --> 00:47:48.763

So that's one type of thing that you can do.

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00:47:50.390 --> 00:47:52.490

There's over 3000 counties in the US,

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00:47:52.490 --> 00:47:55.220

and we didn't think that counties was necessarily

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00:47:55.220 --> 00:47:58.800

the best way to think of this.

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00:47:58.800 --> 00:48:01.680

And so in the lab told us

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00:48:01.680 --> 00:48:04.820

about something called the American Community Project,

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00:48:04.820 --> 00:48:06.320

which was developed by a group

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00:48:06.320 --> 00:48:09.170

of researchers at George Washington University.

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00:48:09.170 --> 00:48:13.000

And what it is is it groups US counties

1100

00:48:13.000 --> 00:48:15.340

into 15 community types.

1101

00:48:15.340 --> 00:48:19.270

And these community types are based upon 36 demographic,

1102

00:48:19.270 --> 00:48:21.870

social economic, and cultural indicators

1103

00:48:21.870 --> 00:48:24.170

including population density, income, race,

1104

00:48:24.170 --> 00:48:25.510

things like that.

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00:48:25.510 --> 00:48:29.910

And so, for example, like you would see Philadelphia, LA,

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00:48:29.910 --> 00:48:31.930

New York group together in big city.

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00:48:31.930 --> 00:48:35.070

So it doesn't depend upon the actual space

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00:48:35.070 --> 00:48:37.690

that some were our proximity, that's not where it is,

1109

00:48:37.690 --> 00:48:39.460

is looking at the features.

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00:48:39.460 --> 00:48:42.540

And so we argued that clustering this way would give

1111

00:48:42.540 --> 00:48:47.320

us more information culturally, and kind of to be able

1112

00:48:47.320 --> 00:48:51.160

to start looking at excessive alcohol consumption

1113

00:48:51.160 --> 00:48:53.513

and more specifically, how to target that.

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00:48:54.480 --> 00:48:58.000

And here's just a sample of some of the results.

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00:48:58.000 --> 00:49:00.880

Using the American Communities Project,

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00:49:00.880 --> 00:49:04.610

we can say how these four communities talk

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00:49:04.610 --> 00:49:08.243

about alcohol use and drinking differently.

1118

00:49:09.360 --> 00:49:11.850

African-American communities are discussing drinking

1119

00:49:11.850 --> 00:49:16.043

in the context of night clubs and dance places.

1120

00:49:17.090 --> 00:49:19.230

College towns, not too surprising,

1121

00:49:19.230 --> 00:49:21.170

they're talking about drinking on weekends,

1122

00:49:21.170 --> 00:49:22.470

they're doing drinking humor,

1123

00:49:22.470 --> 00:49:24.540

and they do a lot of sex related topics

1124

00:49:24.540 --> 00:49:26.073

in relationship to drinking.

1125

00:49:27.310 --> 00:49:30.690

Hispanics centers were talking more about drinking

1126

00:49:30.690 --> 00:49:33.880

with family and family was related to drinking.

1127

00:49:33.880 --> 00:49:36.470

Whereas evangelical hubs were talking

1128

00:49:36.470 --> 00:49:39.430

more about their referencing sobriety

1129

00:49:39.430 --> 00:49:44.430

and different types dealing with family and responsibility,

1130

00:49:44.630 --> 00:49:46.250

things like that.

1131

00:49:46.250 --> 00:49:49.440

So it's just interesting because we can use this type

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00:49:49.440 --> 00:49:53.600

of information, not only to predict higher and lower levels

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00:49:53.600 --> 00:49:55.477

of excessive drinking in the county level,

1134

00:49:55.477 --> 00:49:58.070

but more importantly, it gives us the insight

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00:49:58.070 --> 00:50:01.340

into how we make able to develop messages

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00:50:01.340 --> 00:50:05.060

and develop interventions and treatment information.

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00:50:05.060 --> 00:50:08.150

So that was like the first level of that grant.

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00:50:08.150 --> 00:50:11.370

The second and the reason why I think NIDA gave

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00:50:11.370 --> 00:50:14.160

me the grant, the more important reason are not important,

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00:50:14.160 --> 00:50:16.930

but the reason that they were kind of concerned

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00:50:16.930 --> 00:50:21.380

with was we wanted to know from a clinical level

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00:50:21.380 --> 00:50:24.320

how we could use big data and machine learning

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00:50:24.320 --> 00:50:29.320

in order to figure out ways of predicting treatment outcomes

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00:50:29.560 --> 00:50:32.310

and excessive drinking, not excessive drinking,

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00:50:32.310 --> 00:50:35.760

but substance use and substance use outcomes.

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00:50:35.760 --> 00:50:39.230

So in here, we had a simple question,

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00:50:39.230 --> 00:50:43.290

can we use social media language,

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00:50:43.290 --> 00:50:45.350

we collect social media language literally

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00:50:45.350 --> 00:50:46.890

by a click of a button.

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00:50:46.890 --> 00:50:49.680

We wanted to know if we can predict treatment outcomes

1151
00:50:49.680 --> 00:50:51.200
of people who were attending

1152
00:50:51.200 --> 00:50:53.910
intensive outpatient drug treatment.

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00:50:53.910 --> 00:50:55.650
So these are people who are going

1154
00:50:55.650 --> 00:51:00.170
to a clinic based treatment center four times a week

1155
00:51:00.170 --> 00:51:04.960
at least, we collected this in Philadelphia,

1156
00:51:04.960 --> 00:51:08.270
the area where we're at where kind of near a lot

1157
00:51:08.270 --> 00:51:11.830
of the open area drug markets

1158
00:51:11.830 --> 00:51:14.700
that I believe have been close.

1159
00:51:14.700 --> 00:51:17.120
But so we were kind of in this community

1160
00:51:17.120 --> 00:51:19.803
where we had high poverty,

1161
00:51:21.160 --> 00:51:24.930
but yet we were able to kind of,

1162
00:51:24.930 --> 00:51:28.720
there was a lot of community groups

1163
00:51:28.720 --> 00:51:31.740

that were providing services to participants.

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00:51:31.740 --> 00:51:34.440

And so we were able to kind of partner with a lot

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00:51:34.440 --> 00:51:38.500

of them in order to come in and do this research study.

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00:51:38.500 --> 00:51:41.300

So in this study we had the question about,

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00:51:41.300 --> 00:51:43.550

can we use social media language

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00:51:43.550 --> 00:51:45.240

to predict treatment outcomes?

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00:51:45.240 --> 00:51:46.700

And so at the top, you'll see

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00:51:46.700 --> 00:51:49.070

this is kind of how the study went,

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00:51:49.070 --> 00:51:53.140

a person at baseline, which was at treatment intake.

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00:51:53.140 --> 00:51:55.350

So a treatment intake, a person they come

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00:51:55.350 --> 00:51:59.160

to the treatment center and they say, "I need treatment."

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00:51:59.160 --> 00:52:03.200

And while they're doing the intake and they're waiting,

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00:52:03.200 --> 00:52:07.403

we would come up to them and say, "Do you have social media?"

1176
00:52:08.260 --> 00:52:09.307
do you have a social media account?"

1177
00:52:09.307 --> 00:52:11.250
Any social media in which we interested

1178
00:52:11.250 --> 00:52:13.070
in our research study.

1179
00:52:13.070 --> 00:52:14.940
So we were at the treatment sites

1180
00:52:14.940 --> 00:52:18.910
and I was doing while as at the University of Pennsylvania.

1181
00:52:18.910 --> 00:52:22.450
And so we would get their social media language.

1182
00:52:22.450 --> 00:52:25.990
We also had a trained staff who would do

1183
00:52:25.990 --> 00:52:29.900
something called the ASI, the Addiction Severity Index.

1184
00:52:29.900 --> 00:52:32.279
And this right here is like a gold standard

1185
00:52:32.279 --> 00:52:35.190
of assessing addiction severity.

1186
00:52:35.190 --> 00:52:37.270
So we would do like an hour long interview

1187
00:52:37.270 --> 00:52:39.220
for the addiction severity index

1188
00:52:39.220 --> 00:52:41.250

and kind of like that click of a button

1189

00:52:41.250 --> 00:52:44.050

for your social media link.

1190

00:52:44.050 --> 00:52:45.960

We collected the social media language

1191

00:52:45.960 --> 00:52:48.320

for two years before treatment.

1192

00:52:48.320 --> 00:52:50.260

And we're using the social media language

1193

00:52:50.260 --> 00:52:54.650

and the Addiction Severity Index to predict out 90 days,

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00:52:54.650 --> 00:52:57.100

we did some fancy stuff of initially predicting

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00:52:57.100 --> 00:52:57.960

kind of three outcomes,

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00:52:57.960 --> 00:53:00.070

but we settled with one, two outcomes,

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00:53:00.070 --> 00:53:02.910

either dropping out of treatment or staying in treatment.

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00:53:02.910 --> 00:53:05.720

And the reason why we kind of threw out relapse is

1199

00:53:05.720 --> 00:53:09.470

because if you can relapse, but stay in treatment,

1200

00:53:09.470 --> 00:53:10.880

when you are out of treatment,

1201
00:53:10.880 --> 00:53:12.980
you're gone and we can't deliver interventions

1202
00:53:12.980 --> 00:53:14.710
to you anymore or treatment to you.

1203
00:53:14.710 --> 00:53:16.370
So the key is to keep people in treatment

1204
00:53:16.370 --> 00:53:18.820
and we felt that that was the best indicator.

1205
00:53:18.820 --> 00:53:20.880
So what we wanna do is we wanna predict that.

1206
00:53:20.880 --> 00:53:24.650
and our AUCs, which are area under the curve

1207
00:53:24.650 --> 00:53:27.860
And to give you an idea, 50% would indicate chance

1208
00:53:27.860 --> 00:53:30.140
and 100% indicated perfection.

1209
00:53:30.140 --> 00:53:33.960
We were able to predict 90-day treatment outcomes

1210
00:53:33.960 --> 00:53:38.580
at about 79%, which is huge, very, very good.

1211
00:53:38.580 --> 00:53:40.610
Our social behavior work is

1212
00:53:40.610 --> 00:53:44.250
typically in the 68, 70ish, early '70s.

1213
00:53:44.250 --> 00:53:49.140

So getting to a 78.9 or 79% was huge for us,

1214

00:53:49.140 --> 00:53:50.843

and so this was a win.

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00:53:52.150 --> 00:53:56.080

So we can predict treatment outcomes, that's great, right?

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00:53:56.080 --> 00:54:00.560

But I need to figure out how can we apply this?

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00:54:00.560 --> 00:54:02.840

And so that was kind of one of our big questions is

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00:54:02.840 --> 00:54:04.150

how to apply this.

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00:54:04.150 --> 00:54:08.763

And what we decided was to do was to create risk scores.

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00:54:10.120 --> 00:54:13.770

So we did kind of, we simulated a clinical application.

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00:54:13.770 --> 00:54:17.460

So what we did was we took only social media language

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00:54:17.460 --> 00:54:19.460

we have at baseline.

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00:54:19.460 --> 00:54:23.090

And we took the ASI because that's traditionally done

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00:54:23.090 --> 00:54:25.980

at baseline, treatment centers are not gonna get rid

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00:54:25.980 --> 00:54:28.630

of the ASI, so we said, let's simulate this.

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00:54:28.630 --> 00:54:30.410

So we have demographic information

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00:54:30.410 --> 00:54:32.420

which you get from the intake form,

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00:54:32.420 --> 00:54:35.930

we have social media language at baseline,

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00:54:35.930 --> 00:54:37.600

and we have the ASI.

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00:54:37.600 --> 00:54:39.540

And what we did was we use that language

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00:54:39.540 --> 00:54:42.530

to put people into four risk categories,

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00:54:42.530 --> 00:54:46.230

and then predict out where we think they should be

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00:54:46.230 --> 00:54:48.880

at 90 days or they are at 90 days.

1234

00:54:48.880 --> 00:54:51.130

So we're able to label each participant

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00:54:51.130 --> 00:54:55.250

with a risk value of how from highest risk, lowest risk.

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00:54:55.250 --> 00:54:59.530

And as you can see, those first 30 days are crucial.

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00:54:59.530 --> 00:55:02.600

And while you still get dropout from 30 to 90,

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00:55:02.600 --> 00:55:04.630

the first 30 days are crucial.

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00:55:04.630 --> 00:55:08.110

And so we can use this risk score at baseline,

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00:55:08.110 --> 00:55:10.670

the day a person walk into the treatment center

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00:55:10.670 --> 00:55:14.160

to be like, "Hey, we really need to."

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00:55:14.160 --> 00:55:17.330

hopefully we're investing time and energy into everyone,

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00:55:17.330 --> 00:55:20.310

but this group that's in the high risk category

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00:55:20.310 --> 00:55:21.910

or high risk categories,

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00:55:21.910 --> 00:55:25.610

we should start doing some more intensive treatment,

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00:55:25.610 --> 00:55:29.870

some wraparound services, they're gonna need more support.

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00:55:29.870 --> 00:55:31.890

So in a situation, unfortunately,

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00:55:31.890 --> 00:55:34.270

where we have limited resources,

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00:55:34.270 --> 00:55:38.290

this allows us to triage those resources

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00:55:38.290 --> 00:55:42.240

to people at most in need and early on because remember,

1251
00:55:42.240 --> 00:55:44.850
when we lose them they are gone,

1252
00:55:44.850 --> 00:55:48.883
so if we can keep them in treatment longer, we can succeed.

1253
00:55:50.750 --> 00:55:54.540
So that was using social media language.

1254
00:55:54.540 --> 00:55:57.040
And we we've been analyzing the data

1255
00:55:57.040 --> 00:55:59.460
and working with it and figuring out ways

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00:55:59.460 --> 00:56:02.850
of we're moving to using, we also collected language

1257
00:56:02.850 --> 00:56:05.020
while in treatment, we're looking at that.

1258
00:56:05.020 --> 00:56:06.900
And we also are looking at what is

1259
00:56:06.900 --> 00:56:09.163
it specifically about the treatment.

1260
00:56:10.870 --> 00:56:13.300
Spoiler alert, it is looking very similar

1261
00:56:13.300 --> 00:56:15.990
to the county level excessive drinking

1262
00:56:15.990 --> 00:56:17.870
where we're finding language

1263
00:56:17.870 --> 00:56:21.720

about very religious language is showing,

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00:56:21.720 --> 00:56:24.282

indicating better outcomes and treatment

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00:56:24.282 --> 00:56:25.600

or at least in this treatment.

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00:56:25.600 --> 00:56:26.640

But I would like to point out

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00:56:26.640 --> 00:56:28.730

that those were community-based treatment centers

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00:56:28.730 --> 00:56:31.780

and a lot of them have kind of a 12-step orientation.

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00:56:31.780 --> 00:56:33.890

So maybe it's just treatment matching

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00:56:33.890 --> 00:56:35.620

because that's where people kind of,

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00:56:35.620 --> 00:56:37.170

those treatment centers talk a lot

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00:56:37.170 --> 00:56:40.030

about high risk, higher power.

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00:56:40.030 --> 00:56:44.020

So COVID hit, and we have a situation,

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00:56:44.020 --> 00:56:48.480

we're at NIDA clinic and Baltimore shut down,

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00:56:48.480 --> 00:56:52.130

and we decided my team and some collaborators

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00:56:52.130 --> 00:56:55.490
to make lemonade out of lemons and felt

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00:56:55.490 --> 00:57:00.450
that who else is best suited to do a national online study

1278
00:57:00.450 --> 00:57:02.270
than someone who uses social media

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00:57:02.270 --> 00:57:04.130
and all of these digital tools.

1280
00:57:04.130 --> 00:57:06.950
So we got into the lab, not physically,

1281
00:57:06.950 --> 00:57:10.023
but virtually and designed the study.

1282
00:57:11.592 --> 00:57:14.093
We're doing a national study at NIDA,

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00:57:14.093 --> 00:57:17.470
2,500 people with and without substance use disorders,

1284
00:57:17.470 --> 00:57:19.570
including alcohol use disorder

1285
00:57:19.570 --> 00:57:22.640
in their natural environments throughout the US.

1286
00:57:22.640 --> 00:57:24.460
We have already collected data,

1287
00:57:24.460 --> 00:57:27.140
we have already met our recruitment goals.

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00:57:27.140 --> 00:57:31.860

For baseline, we have minimum of 2,500 people in the study.

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00:57:31.860 --> 00:57:34.190

We come and collect their social media language.

1290

00:57:34.190 --> 00:57:39.190

We do tons of surveys, not tons, it's minimum.

1291

00:57:39.570 --> 00:57:44.570

We collect depression, anxiety, food insecurity,

1292

00:57:44.610 --> 00:57:49.330

economic indicator, loss of jobs, treatment indicators,

1293

00:57:49.330 --> 00:57:52.910

drug use market, COVID risk type scores.

1294

00:57:52.910 --> 00:57:56.370

We did kind of pack that 'cause we weren't sure

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00:57:56.370 --> 00:57:58.840

at the beginning of COVID what was important

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00:57:58.840 --> 00:58:00.040

and we had some hypothesis

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00:58:00.040 --> 00:58:02.460

about social isolation and loneliness

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00:58:02.460 --> 00:58:04.693

and access to harm reduction and treatment.

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00:58:05.940 --> 00:58:08.020

So we have that at baseline,

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00:58:08.020 --> 00:58:12.270

we do a short 30-day initial study

1301

00:58:12.270 --> 00:58:15.830

where we use an EMA, Ecological Momentary Assessment,

1302

00:58:15.830 --> 00:58:19.320

sending out a smartphone message, a little survey.

1303

00:58:19.320 --> 00:58:21.120

We do that for 30 days.

1304

00:58:21.120 --> 00:58:23.630

And then we follow them for like the next six months

1305

00:58:23.630 --> 00:58:26.800

where we do more surveys

1306

00:58:26.800 --> 00:58:28.420

and those are a little bit tailored,

1307

00:58:28.420 --> 00:58:32.600

kind of we're ending with vaccine hesitancy

1308

00:58:32.600 --> 00:58:34.510

and things like that.

1309

00:58:34.510 --> 00:58:37.573

So as COVID changed, we were able to change our studies.

1310

00:58:39.042 --> 00:58:39.875

But the key part,

1311

00:58:39.875 --> 00:58:41.590

one of the key things we're gonna show you is

1312

00:58:41.590 --> 00:58:45.710

that of 300 of those, we're doing a smartphone sensor study

1313

00:58:45.710 --> 00:58:48.530

where they put a sensor on their smartphone

1314

00:58:48.530 --> 00:58:49.890

and we're capturing data

1315

00:58:49.890 --> 00:58:54.560

from over 15 sensors on the smartphone,

1316

00:58:54.560 --> 00:58:57.500

including social media language, as well as keystrokes.

1317

00:58:57.500 --> 00:59:00.790

So things that we can gather from that smartphone,

1318

00:59:00.790 --> 00:59:03.900

the sensor data, when you're online

1319

00:59:03.900 --> 00:59:06.200

and you have your smartphone that you carry everywhere,

1320

00:59:06.200 --> 00:59:09.350

that you text and type and do almost everything with,

1321

00:59:09.350 --> 00:59:11.490

it captures a lot of information about you.

1322

00:59:11.490 --> 00:59:14.363

It probably know better than most people that you knows.

1323

00:59:15.230 --> 00:59:17.510

It knows, we get your accelerometer data,

1324

00:59:17.510 --> 00:59:19.820

we can do things like get your physical movement

1325

00:59:19.820 --> 00:59:21.620

and daily activities.

1326
00:59:21.620 --> 00:59:24.180
Bluetooth are kind of telling us about social interactions,

1327
00:59:24.180 --> 00:59:25.490
how close you are to people,

1328
00:59:25.490 --> 00:59:29.000
and how many people you come in contact with.

1329
00:59:29.000 --> 00:59:30.610
GPS, we know where you went,

1330
00:59:30.610 --> 00:59:32.560
we know how long you were there.

1331
00:59:32.560 --> 00:59:37.200
We know if you rode, drove, walked, or ran.

1332
00:59:37.200 --> 00:59:40.030
We have the light sensors, we have WiFi scans.

1333
00:59:40.030 --> 00:59:44.220
We have your calls and texts, smartphone text logs.

1334
00:59:44.220 --> 00:59:46.820
So we can look at things like how,

1335
00:59:46.820 --> 00:59:49.650
are we getting changes in how many people you talk to

1336
00:59:49.650 --> 00:59:50.483
or you texts?

1337
00:59:50.483 --> 00:59:53.620
And we can look at things like what were your mood

1338
00:59:53.620 --> 00:59:56.880

and behaviors before then, after, during,

1339

00:59:56.880 --> 00:59:58.030

are we getting changes?

1340

00:59:58.030 --> 01:00:00.060

We also know what apps you're using

1341

01:00:00.060 --> 01:00:02.600

and we know your smartphone, your language

1342

01:00:02.600 --> 01:00:07.210

and we can predict social interactions and daily activities.

1343

01:00:07.210 --> 01:00:09.970

So to give you an idea of something that we're doing

1344

01:00:09.970 --> 01:00:14.630

with the keystroke data, we can look at, for example,

1345

01:00:14.630 --> 01:00:16.740

we can do like a just account

1346

01:00:16.740 --> 01:00:20.280

of having a list of lexicon list of words.

1347

01:00:20.280 --> 01:00:22.610

We can look at how many times you're talking

1348

01:00:22.610 --> 01:00:25.140

about vaccines or COVID.

1349

01:00:25.140 --> 01:00:28.970

We can also look at people talking about mood and stress,

1350

01:00:28.970 --> 01:00:31.980

and you know how they're doing, how they're coping.

1351

01:00:31.980 --> 01:00:33.800

Are they socially isolated?

1352

01:00:33.800 --> 01:00:35.740

Are they getting treatment and things?

1353

01:00:35.740 --> 01:00:38.620

So that's the study that we have in the field right now

1354

01:00:38.620 --> 01:00:42.280

that I'm really proud of my research lab.

1355

01:00:42.280 --> 01:00:44.520

I wouldn't be able to do this without my team.

1356

01:00:44.520 --> 01:00:46.550

We have a great group of people who are

1357

01:00:46.550 --> 01:00:49.613

from computer science, psychology, public health,

1358

01:00:50.660 --> 01:00:54.790

and we all work together to answer some of these questions.

1359

01:00:54.790 --> 01:00:57.200

But one of the key things about my lab is

1360

01:01:00.658 --> 01:01:03.260

I do a lot of ethics work.

1361

01:01:03.260 --> 01:01:04.860

I'm trying to turn off my phone,

1362

01:01:06.200 --> 01:01:09.120

sorry, yes, I have a strange ring through.

1363

01:01:09.120 --> 01:01:10.060

We have a lot of people

1364

01:01:10.060 --> 01:01:13.470

from different diverse backgrounds that we ask questions

1365

01:01:13.470 --> 01:01:15.080

all the time, especially when we're talking

1366

01:01:15.080 --> 01:01:16.790

about like releasing of data.

1367

01:01:16.790 --> 01:01:18.140

What should be released?

1368

01:01:18.140 --> 01:01:19.630

How has she been released?

1369

01:01:19.630 --> 01:01:22.750

and what are our responsibilities?

1370

01:01:22.750 --> 01:01:24.440

I like to thank my collaborators,

1371

01:01:24.440 --> 01:01:28.810

have lots of collaborators here at NIH and outside.

1372

01:01:28.810 --> 01:01:30.870

And of course, I like to think Nora

1373

01:01:30.870 --> 01:01:35.840

and Amy who have helped me from when I was at Penn

1374

01:01:35.840 --> 01:01:39.193

to being here at NIDA IRP, so thank you.

1375

01:01:43.580 --> 01:01:45.160

<v ->Thank you, Brenda, that was great.</v>

1376

01:01:45.160 --> 01:01:47.260

It was wonderful to hear about what you're working on

1377

01:01:47.260 --> 01:01:48.093

and about your career

1378

01:01:48.093 --> 01:01:50.880

as well what motivated you in to this area.

1379

01:01:50.880 --> 01:01:54.800

So thanks again to both Brenda and Kristian.

1380

01:01:54.800 --> 01:01:57.790

Next, we're gonna have questions from the audience,

1381

01:01:57.790 --> 01:01:59.430

and I'm gonna turn it over to Dr. Lindsey

1382

01:01:59.430 --> 01:02:01.170

who's gonna moderate that session.

1383

01:02:01.170 --> 01:02:04.023

So please put your questions into the chat box.

1384

01:02:06.380 --> 01:02:07.213

<v ->Awesome, thanks.</v>

1385

01:02:07.213 --> 01:02:08.700

Yeah, there's a little question and answer box

1386

01:02:08.700 --> 01:02:12.340

at the bottom and list the name of the person if you want

1387

01:02:12.340 --> 01:02:14.360

one person specifically.

1388

01:02:14.360 --> 01:02:16.660

So first, one of the purposes

1389

01:02:16.660 --> 01:02:18.400

of the seminar series is to reach out

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01:02:18.400 --> 01:02:21.390

to younger who are thinking of a career in data science.

1391

01:02:21.390 --> 01:02:24.380

So I'll ask both of our speakers today,

1392

01:02:24.380 --> 01:02:27.030

what advice you have for younger scientists

1393

01:02:27.030 --> 01:02:30.010

that are thinking of starting a career in data science

1394

01:02:34.290 --> 01:02:35.510

Either of you can jump.

1395

01:02:35.510 --> 01:02:38.150

<v ->okay, well, for me,</v>

1396

01:02:38.150 --> 01:02:40.350

I kind of came in through the back door.

1397

01:02:40.350 --> 01:02:43.030

I'm not a trained data scientist,

1398

01:02:43.030 --> 01:02:45.123

but I think just like public health,

1399

01:02:46.500 --> 01:02:48.780

I felt like public health is a very diverse group

1400

01:02:48.780 --> 01:02:50.560

of kind of who gets there.

1401

01:02:50.560 --> 01:02:53.273

And I work with data scientists to help me do,

1402

01:02:54.320 --> 01:02:56.120

using their tools and methodologies

1403

01:02:56.120 --> 01:02:58.783

in order to apply it in this domain.

1404

01:03:00.640 --> 01:03:03.723

I always spoke the language of math and science,

1405

01:03:06.989 --> 01:03:09.780

the community I came with that wasn't discouraged.

1406

01:03:09.780 --> 01:03:13.760

So I had lots of great teachers and support systems

1407

01:03:13.760 --> 01:03:15.900

and so they allowed me to be me.

1408

01:03:15.900 --> 01:03:18.670

And honestly, I felt like it was a language

1409

01:03:18.670 --> 01:03:20.943

that I could reference and understand.

1410

01:03:20.943 --> 01:03:24.093

And so I just kind of leaned into that.

1411

01:03:25.280 --> 01:03:27.580

And so that's what I would say is like don't put limits

1412

01:03:27.580 --> 01:03:31.133

on people and to just go forward in your interest.

1413

01:03:32.920 --> 01:03:35.760

<v ->Yeah, I think my advice would be</v>

1414

01:03:35.760 --> 01:03:38.450

to get some subject area expertise

1415

01:03:38.450 --> 01:03:40.550

in the areas that you want to work in as well.

1416

01:03:40.550 --> 01:03:45.550

I think the tools of data science are really powerful

1417

01:03:46.040 --> 01:03:50.070

and can be applied sort of blindly.

1418

01:03:50.070 --> 01:03:52.330

If you want to, you can just throw whatever model

1419

01:03:52.330 --> 01:03:56.810

you want at whatever data and get an answer.

1420

01:03:56.810 --> 01:03:59.700

But the answer is kind of meaningless if you don't

1421

01:03:59.700 --> 01:04:02.270

really have a good, deep understanding

1422

01:04:02.270 --> 01:04:04.930

of the subject area, the data comes from,

1423

01:04:04.930 --> 01:04:08.800

like why you would want to be applying that method or model,

1424

01:04:08.800 --> 01:04:12.250

like what are relevant questions in that area, et cetera.

1425

01:04:12.250 --> 01:04:14.820

And so I think even if you are like a very mathy person

1426
01:04:14.820 --> 01:04:17.670
like I am, and it seems like Dr. Curtis as well,

1427
01:04:17.670 --> 01:04:22.500
I think , it's always worth the time to also dive deep

1428
01:04:22.500 --> 01:04:26.053
on the non-mathematical aspects of the field.

1429
01:04:33.240 --> 01:04:34.690
<v ->Lindsey, we can't hear you.</v>

1430
01:04:35.749 --> 01:04:39.850
<v ->Sorry, so I'm gonna read one question from the attendees,</v>

1431
01:04:39.850 --> 01:04:41.377
thank you for your answers.

1432
01:04:41.377 --> 01:04:44.430
"Dr. Curtis, very interesting application of data science

1433
01:04:44.430 --> 01:04:46.750
to drug use in the phone center project.

1434
01:04:46.750 --> 01:04:47.963
How might predictive algorithms

1435
01:04:47.963 --> 01:04:51.530
on the phone sensors be used in a real world setting?

1436
01:04:51.530 --> 01:04:54.470
And how do you see this being the ultimate end user?

1437
01:04:54.470 --> 01:04:57.620
For example, clinicians, hospitals, healthcare,

1438
01:04:57.620 --> 01:04:59.270

healthcare providers, et cetera."

1439

01:05:00.580 --> 01:05:02.640

<v ->My approach because probably I started</v>

1440

01:05:02.640 --> 01:05:06.560

in treatment is how we can help treatment

1441

01:05:06.560 --> 01:05:10.330

and also strongly support peer support and family support.

1442

01:05:10.330 --> 01:05:13.780

And so I'm looking at coming from the lens

1443

01:05:13.780 --> 01:05:15.900

of not necessarily providing intervention

1444

01:05:15.900 --> 01:05:18.440

to the person per se the smartphone,

1445

01:05:18.440 --> 01:05:19.850

but how we can use that information

1446

01:05:19.850 --> 01:05:22.090

that can inform clinical treatment.

1447

01:05:22.090 --> 01:05:25.310

And so, for example, things I've heard from families is

1448

01:05:25.310 --> 01:05:27.170

like, they just want to have an indicator

1449

01:05:27.170 --> 01:05:29.993

that their family member is doing okay.

1450

01:05:30.830 --> 01:05:32.620

Are they similar to they were yesterday

1451

01:05:32.620 --> 01:05:35.130

or two the last week, has there been any changes?

1452

01:05:35.130 --> 01:05:37.560

So I can see an application there that can kind

1453

01:05:37.560 --> 01:05:40.400

of be this early detection, early warning tool,

1454

01:05:40.400 --> 01:05:44.920

which can also be applied to treatment centers.

1455

01:05:44.920 --> 01:05:48.320

Case managers and counselors have a huge number

1456

01:05:48.320 --> 01:05:51.960

of participants, patients per clinician.

1457

01:05:51.960 --> 01:05:54.750

And so if we can provide them with just like a

1458

01:05:54.750 --> 01:05:56.810

daily dashboard or an idea

1459

01:05:56.810 --> 01:05:58.930

of how their patients are doing

1460

01:05:58.930 --> 01:06:02.450

and who they may wanna check in with earlier,

1461

01:06:02.450 --> 01:06:04.190

I think that that would be great.

1462

01:06:04.190 --> 01:06:07.090

So I'm using these tools for,

1463

01:06:07.090 --> 01:06:09.060

we don't know if there's a certain level

1464

01:06:09.060 --> 01:06:10.880

that indicate risk, right?

1465

01:06:10.880 --> 01:06:12.630

Or is it a change of behavior?

1466

01:06:12.630 --> 01:06:14.977

Is it that a person comes from treatment,

1467

01:06:14.977 --> 01:06:17.260

you know from their treatment records

1468

01:06:17.260 --> 01:06:18.410

and you've been talking to them.

1469

01:06:18.410 --> 01:06:20.910

If these three weeks they've been doing great

1470

01:06:20.910 --> 01:06:22.300

and now. we see a change,

1471

01:06:22.300 --> 01:06:25.130

we don't know if that changes a good change or a bad change,

1472

01:06:25.130 --> 01:06:26.850

but that would allow the counselor to then

1473

01:06:26.850 --> 01:06:29.370

or the treatment provider or peer support to check

1474

01:06:29.370 --> 01:06:31.960

in with that person to see what's going on

1475

01:06:31.960 --> 01:06:35.480

and then to start initiating if higher level

1476
01:06:35.480 --> 01:06:36.950
of intervention is needed.

1477
01:06:36.950 --> 01:06:39.510
We can also use this with medication adherence

1478
01:06:39.510 --> 01:06:41.830
and monitoring to see how people are doing

1479
01:06:41.830 --> 01:06:43.703
on medication assisted treatments.

1480
01:06:45.410 --> 01:06:47.860
<v ->Awesome, thank you for that answer, very great.</v>

1481
01:06:49.230 --> 01:06:51.373
Another question that we had was regarding COVID,

1482
01:06:51.373 --> 01:06:52.970
I mean, both of you have mentioned

1483
01:06:52.970 --> 01:06:54.650
how your projects have changed

1484
01:06:54.650 --> 01:06:57.050
because of the pandemic, but what kind

1485
01:06:57.050 --> 01:06:58.530
of advice or outreach would you have

1486
01:06:58.530 --> 01:07:02.730
for younger scientists and this kind of pandemic field

1487
01:07:03.580 --> 01:07:04.530
that they find themselves

1488
01:07:04.530 --> 01:07:06.683

kind of growing up in education wise?

1489

01:07:09.610 --> 01:07:13.330

<v ->I'll go first, I have a lot of like small things,</v>

1490

01:07:13.330 --> 01:07:16.220

I think one of them, and I have sort of fallen victim

1491

01:07:16.220 --> 01:07:18.010

to not using this advice is

1492

01:07:18.010 --> 01:07:20.480

that you don't have to chase every,

1493

01:07:20.480 --> 01:07:22.120

I don't wanna call it COVID like a news cycle,

1494

01:07:22.120 --> 01:07:24.490

but just because something's the hot area,

1495

01:07:24.490 --> 01:07:25.400

you can still keep working

1496

01:07:25.400 --> 01:07:26.510

on the thing that you're working on.

1497

01:07:26.510 --> 01:07:30.390

You don't have to necessarily pivot to COVID.

1498

01:07:30.390 --> 01:07:32.350

Yes, it's an important topic, and yes,

1499

01:07:32.350 --> 01:07:34.120

we obviously need a lot of research on it,

1500

01:07:34.120 --> 01:07:35.750

but also we still need to keep doing research

1501

01:07:35.750 --> 01:07:36.730
on other things as well.

1502

01:07:36.730 --> 01:07:39.330
And I think for a lot of people,

1503

01:07:39.330 --> 01:07:42.367
myself obviously included, it's tempting to be like,

1504

01:07:42.367 --> 01:07:43.670
"Okay, this is the thing right now."

1505

01:07:43.670 --> 01:07:46.993
So like I should pivot all my resources towards that,

1506

01:07:48.280 --> 01:07:52.000
but other research is valuable as well.

1507

01:07:52.000 --> 01:07:54.360
I also think one of the really great things

1508

01:07:54.360 --> 01:07:59.360
about the sort of online or like virtual type

1509

01:07:59.780 --> 01:08:02.240
of research that's been happening since COVID,

1510

01:08:02.240 --> 01:08:03.540
I don't want to say great thing about COVID

1511

01:08:03.540 --> 01:08:05.770
'cause I think there's nothing great about that,

1512

01:08:05.770 --> 01:08:08.650
but this sort of virtual world is that I can be here

1513

01:08:08.650 --> 01:08:12.700

with you all, people can tune in from all over the place.

1514

01:08:12.700 --> 01:08:15.260

Recently, I was involved in organizing a conference

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01:08:15.260 --> 01:08:16.520

that was fully remote.

1516

01:08:16.520 --> 01:08:18.300

It was much cheaper than normal,

1517

01:08:18.300 --> 01:08:20.630

people from all over the world could attend.

1518

01:08:20.630 --> 01:08:23.900

We actually do try to make it accessible even in-person,

1519

01:08:23.900 --> 01:08:25.670

but it was much more accessible this year.

1520

01:08:25.670 --> 01:08:27.690

And so I think sort of taking advantage

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01:08:27.690 --> 01:08:29.850

of this situation where everything is online,

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01:08:29.850 --> 01:08:31.710

there's so many resources,

1523

01:08:31.710 --> 01:08:34.970

you have access to people who maybe normally,

1524

01:08:34.970 --> 01:08:37.453

it would be hard to talk to.

1525

01:08:38.480 --> 01:08:40.390

There's a lot less overhead and a lot less cost,

1526

01:08:40.390 --> 01:08:43.080

I think too for people to just do a quick Zoom

1527

01:08:43.080 --> 01:08:45.010

or tune into something like this.

1528

01:08:45.010 --> 01:08:47.580

And so that's been one of the positives

1529

01:08:47.580 --> 01:08:51.060

of this increasingly virtual research world

1530

01:08:51.060 --> 01:08:52.110

that we're living in.

1531

01:08:53.030 --> 01:08:56.220

<v ->For us, it was a kind of an interesting point</v>

1532

01:08:56.220 --> 01:08:59.240

because we were like, as our center was closing down,

1533

01:08:59.240 --> 01:09:02.050

we realized that a lot of treatment centers

1534

01:09:02.050 --> 01:09:04.620

across the US where people remember they go four

1535

01:09:04.620 --> 01:09:06.630

or five days a week was closing down.

1536

01:09:06.630 --> 01:09:09.540

And so we were really afraid of that,

1537

01:09:09.540 --> 01:09:13.330

we also know that during the the opiate epidemic

1538

01:09:13.330 --> 01:09:16.050

that our fear was that places

1539

01:09:16.050 --> 01:09:18.974

that were providing harm reduction like Naloxone

1540

01:09:18.974 --> 01:09:22.270

and other things were going to disappear.

1541

01:09:22.270 --> 01:09:25.820

And so we were going to have higher levels of overdoses.

1542

01:09:25.820 --> 01:09:29.550

So we were afraid of that because harm reduction,

1543

01:09:29.550 --> 01:09:32.330

treatment sites closing their doors

1544

01:09:32.330 --> 01:09:34.040

and having social distance,

1545

01:09:34.040 --> 01:09:37.260

we were also afraid because substance use is

1546

01:09:37.260 --> 01:09:41.890

already a social isolating type of situation,

1547

01:09:41.890 --> 01:09:44.760

and people who are in treatment are highly stigmatized

1548

01:09:44.760 --> 01:09:46.390

and already kind of like an outcast

1549

01:09:46.390 --> 01:09:48.030

unfortunately many times.

1550

01:09:48.030 --> 01:09:49.010

And so we were afraid

1551

01:09:49.010 --> 01:09:51.110

that that isolation was gonna continue.

1552

01:09:51.110 --> 01:09:53.710

And even when they come into treatment centers,

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01:09:53.710 --> 01:09:55.860

if they have COVID that they were going

1554

01:09:55.860 --> 01:09:57.420

to get discriminated against.

1555

01:09:57.420 --> 01:09:59.840

And so we had these kinds of questions that were happening

1556

01:09:59.840 --> 01:10:04.050

to this population and that we were concerned about.

1557

01:10:04.050 --> 01:10:07.990

And so I was like on the phone, and texting, and typing

1558

01:10:07.990 --> 01:10:12.047

with different people, my collaborator and saying like,

1559

01:10:12.047 --> 01:10:13.847

"This is a problem, we should do it,

1560

01:10:14.740 --> 01:10:16.430

this could be horrible."

1561

01:10:16.430 --> 01:10:18.730

And more importantly, we were really afraid

1562

01:10:18.730 --> 01:10:21.000

that this population wasn't gonna get heard,

1563

01:10:21.000 --> 01:10:22.340

that this was gonna happen

1564

01:10:22.340 --> 01:10:23.900

and we weren't gonna know about it,

1565

01:10:23.900 --> 01:10:26.340

the impact of COVID to this population.

1566

01:10:26.340 --> 01:10:29.740

So that was that, and then at the same time,

1567

01:10:29.740 --> 01:10:34.170

I'm doing this, Nora puts out this statement article

1568

01:10:34.170 --> 01:10:35.417

about the harms and I was like,

1569

01:10:35.417 --> 01:10:36.640

"Oh, my God," now I was teasing,

1570

01:10:36.640 --> 01:10:38.530

I was like, "Does she have access to my share drive?"

1571

01:10:38.530 --> 01:10:39.730

No, I know she doesn't,

1572

01:10:39.730 --> 01:10:42.020

but it was like, this timing was perfect

1573

01:10:42.020 --> 01:10:44.280

and it gave me kind of this,

1574

01:10:44.280 --> 01:10:48.170

I felt like, okay, I'm one track and I can push through.

1575

01:10:48.170 --> 01:10:50.280

Getting something like this type of study done

1576

01:10:50.280 --> 01:10:53.190

at NIDA, inside the government was kind of hard.

1577

01:10:53.190 --> 01:10:55.520

I'm talking about a lot of stuff

1578

01:10:55.520 --> 01:10:58.340

that normally governments don't collect.

1579

01:10:58.340 --> 01:11:00.500

So we had to partner with Penn

1580

01:11:00.500 --> 01:11:02.000

and we're able to kind of get this done.

1581

01:11:02.000 --> 01:11:05.530

Everyone was supportive, Nora Collins, everyone.

1582

01:11:05.530 --> 01:11:08.613

And so we got it and Amy and we got it done.

1583

01:11:09.850 --> 01:11:12.380

So COVID in that sense really helped me push

1584

01:11:12.380 --> 01:11:15.130

my research more national.

1585

01:11:15.130 --> 01:11:18.460

And also, I think hopefully, give voice to a population

1586

01:11:18.460 --> 01:11:22.260

that were not going to get heard,

1587

01:11:22.260 --> 01:11:24.378

and I was really, we were really kind of scared

1588

01:11:24.378 --> 01:11:26.003

of what was gonna be going on.

1589

01:11:27.560 --> 01:11:29.000

<v ->Awesome, thanks, it's good that you're able</v>

1590

01:11:29.000 --> 01:11:31.853

to find some positives at a time of so many negatives.

1591

01:11:32.910 --> 01:11:35.370

I know Wilson has a question, so I'll turn it over to Wilson

1592

01:11:35.370 --> 01:11:37.653

to ask his question to the panelists.

1593

01:11:38.960 --> 01:11:42.680

<v ->Thanks very much, this has been a wonderful session today,</v>

1594

01:11:42.680 --> 01:11:46.760

and I really appreciate both the career path

1595

01:11:46.760 --> 01:11:49.440

that Dr. Lum you described and Dr. Curtis,

1596

01:11:49.440 --> 01:11:53.030

your wonderful research, as well as some new information

1597

01:11:53.030 --> 01:11:54.620

for me about your career path,

1598

01:11:54.620 --> 01:11:56.190

how you started and how you got launched,

1599

01:11:56.190 --> 01:11:58.870

it was really important.

1600

01:11:58.870 --> 01:12:00.950

I have a couple of questions,

1601

01:12:00.950 --> 01:12:04.160

First, I think applies mostly to you, Dr. Lum,

1602

01:12:04.160 --> 01:12:08.000

look at a question about you've moved

1603

01:12:08.000 --> 01:12:12.220

between academic and NGO and start up

1604

01:12:12.220 --> 01:12:17.210

and shifting back to academic is not

1605

01:12:17.210 --> 01:12:19.540

always so easy after you been out

1606

01:12:19.540 --> 01:12:20.840

of an academic environment,

1607

01:12:20.840 --> 01:12:22.450

what has it been like coming back

1608

01:12:22.450 --> 01:12:26.300

and any advice about how to make those shifts?

1609

01:12:26.300 --> 01:12:29.520

<v ->Yeah, I mean, I think what you said is totally true.</v>

1610

01:12:29.520 --> 01:12:32.050

It's not always, I guess I'd say

1611

01:12:32.050 --> 01:12:33.730

something sort of that it's not ever easy

1612

01:12:33.730 --> 01:12:38.463

to completely shift sectors like that.

1613

01:12:39.350 --> 01:12:42.080

How it's been for me, it's been okay.

1614

01:12:42.080 --> 01:12:46.840

I think for the most part continued doing the same type

1615

01:12:46.840 --> 01:12:48.630

of research that I was doing before.

1616

01:12:48.630 --> 01:12:51.240

I feel like some of the differences are

1617

01:12:51.240 --> 01:12:53.530

when I was at the NGO, we were doing research.

1618

01:12:53.530 --> 01:12:56.700

And so I think working in a career

1619

01:12:56.700 --> 01:12:58.710

where you're still doing research

1620

01:12:58.710 --> 01:13:00.760

even if your affiliation is an academic,

1621

01:13:00.760 --> 01:13:04.020

that does leave doors open to go back to academia

1622

01:13:04.020 --> 01:13:05.490

'cause you're still publishing,

1623

01:13:05.490 --> 01:13:07.360

you're sort of still sort of like keeping a presence

1624

01:13:07.360 --> 01:13:08.370

in the research world.

1625

01:13:08.370 --> 01:13:12.023

So I think if somebody wanted to keep those doors open,

1626
01:13:13.970 --> 01:13:17.380
picking a career path where there are still opportunities

1627
01:13:17.380 --> 01:13:19.960
to continue publishing would would sort of help keep

1628
01:13:19.960 --> 01:13:20.793
those doors open.

1629
01:13:20.793 --> 01:13:21.870
I think in terms of differences,

1630
01:13:21.870 --> 01:13:26.280
one thing I really appreciated about working in research,

1631
01:13:26.280 --> 01:13:29.980
but in an NGO was we measured our impact,

1632
01:13:29.980 --> 01:13:32.300
I think a little bit differently than people in academia do.

1633
01:13:32.300 --> 01:13:36.050
So it wasn't, I don't wanna say this is true in academia,

1634
01:13:36.050 --> 01:13:37.130
but I think a caricature is

1635
01:13:37.130 --> 01:13:40.060
that it's purely about citation counts

1636
01:13:40.060 --> 01:13:43.020
and impact factors and all of those things

1637
01:13:43.020 --> 01:13:47.340
whereas when I was at the NGO, it was, yeah,

1638
01:13:47.340 --> 01:13:49.110

we were publishing 'cause we were researchers,

1639

01:13:49.110 --> 01:13:51.993

but also part of how we measured our impact was

1640

01:13:51.993 --> 01:13:54.800

were we helping advocacy groups understand the issues,

1641

01:13:54.800 --> 01:13:56.150

were we talking to journalists

1642

01:13:56.150 --> 01:13:58.760

and helping them understand the important issues

1643

01:13:58.760 --> 01:14:03.760

to be covering, were we working with other say NGOs

1644

01:14:05.690 --> 01:14:08.450

to help them say, scope their research

1645

01:14:08.450 --> 01:14:10.120

in say the AI space, right?

1646

01:14:10.120 --> 01:14:14.870

So I think we had a sort of more flexible definition

1647

01:14:14.870 --> 01:14:18.750

of success and that's something that I really appreciated.

1648

01:14:18.750 --> 01:14:21.500

So coming back to academia has been a little bit interesting

1649

01:14:21.500 --> 01:14:25.400

in the sense that I feel like I need to shift

1650

01:14:25.400 --> 01:14:26.690

my attention more to publishing

1651
01:14:26.690 --> 01:14:28.820
rather than these other aspects

1652
01:14:28.820 --> 01:14:29.900
that I've focused on for so long.

1653
01:14:29.900 --> 01:14:30.733
And it can be difficult,

1654
01:14:30.733 --> 01:14:33.450
I think to reramp up on the like publish, publish,

1655
01:14:33.450 --> 01:14:37.530
publish side of things, but it's been enjoyable.

1656
01:14:37.530 --> 01:14:39.680
And I like my current job, that sounds negative,

1657
01:14:39.680 --> 01:14:41.320
I don't mean it to be negative about academia,

1658
01:14:41.320 --> 01:14:43.890
but I think one of the difficulties is

1659
01:14:43.890 --> 01:14:46.550
kind of reworking how you think

1660
01:14:46.550 --> 01:14:48.640
about your work and sort of revamping

1661
01:14:48.640 --> 01:14:51.730
where you put your priorities to sort of set

1662
01:14:51.730 --> 01:14:53.810
yourself up for success in areas

1663
01:14:53.810 --> 01:14:55.313

where the incentives are different.

1664

01:14:56.370 --> 01:14:58.060

<v ->Certainly strikes me that both you</v>

1665

01:14:58.060 --> 01:15:00.330

and Dr. Curtis really approach your work

1666

01:15:00.330 --> 01:15:04.963

from a passion for improving lives of groups

1667

01:15:05.830 --> 01:15:09.290

that have been traditionally ignored by so much of society.

1668

01:15:09.290 --> 01:15:12.780

And I'm curious about how you balance

1669

01:15:12.780 --> 01:15:17.500

sort of the academic nature of your work with this,

1670

01:15:17.500 --> 01:15:21.470

the real world, changing the world that we live in

1671

01:15:21.470 --> 01:15:22.980

and making lives better,

1672

01:15:22.980 --> 01:15:25.150

which clearly motivates both of you very much,

1673

01:15:25.150 --> 01:15:26.750

it's just wonderful to see.

1674

01:15:26.750 --> 01:15:28.100

But how do you balance some of those

1675

01:15:28.100 --> 01:15:30.230

'cause because there's a need for the rigorous

1676

01:15:30.230 --> 01:15:32.280

in terms of the math and the science

1677

01:15:32.280 --> 01:15:35.763

as well as a focus on the long term goals?

1678

01:15:37.530 --> 01:15:39.230

<v ->Well, I actually think one of the things</v>

1679

01:15:39.230 --> 01:15:43.400

that Kristian talked about, which is very key is

1680

01:15:43.400 --> 01:15:47.650

that while we are doing our research and we are publishing

1681

01:15:47.650 --> 01:15:51.540

and getting information out to an academic institution,

1682

01:15:51.540 --> 01:15:55.310

it is if not more important to make sure

1683

01:15:55.310 --> 01:15:58.360

that that information is getting to people on the ground,

1684

01:15:58.360 --> 01:16:00.690

that's going to be using it.

1685

01:16:00.690 --> 01:16:03.380

So when, for example, I published an article,

1686

01:16:03.380 --> 01:16:05.890

we may make an infographic or something that we can put

1687

01:16:05.890 --> 01:16:09.237

on social media that can boil down this information

1688

01:16:09.237 --> 01:16:13.133

and make it accessible to lots of different people.

1689

01:16:15.550 --> 01:16:17.040

So that's one of the things I do

1690

01:16:17.040 --> 01:16:19.430

as well as the organizations and groups

1691

01:16:19.430 --> 01:16:22.720

that I've participated in the type of community outreach

1692

01:16:22.720 --> 01:16:25.840

that I give and making myself accessible

1693

01:16:25.840 --> 01:16:30.840

to high school students and students all along the pipeline

1694

01:16:31.730 --> 01:16:35.630

and my trainees in order to kind of, to do this.

1695

01:16:35.630 --> 01:16:39.770

So I think that we have an obligation and well,

1696

01:16:39.770 --> 01:16:41.030

I can't, at this point,

1697

01:16:41.030 --> 01:16:43.900

I used to be with the Public Policy Center at Annenberg,

1698

01:16:43.900 --> 01:16:46.300

but while at this point I don't do policy work,

1699

01:16:46.300 --> 01:16:49.290

I can make sure that my publications

1700

01:16:49.290 --> 01:16:52.500

as well as information that I put out could easily be used

1701
01:16:52.500 --> 01:16:54.900
and accessed by policy makers

1702
01:16:54.900 --> 01:16:57.290
so that they can have empirical evidence

1703
01:16:57.290 --> 01:16:59.113
in order to support decisions.

1704
01:17:01.780 --> 01:17:03.580
<v ->Thank you, I think that's a wonderful idea.</v>

1705
01:17:03.580 --> 01:17:06.800
And one I would personally strive for and encourage people

1706
01:17:06.800 --> 01:17:11.000
to emulate to make sure that work has multiple levels

1707
01:17:11.000 --> 01:17:13.463
that it can be used.

1708
01:17:15.970 --> 01:17:17.720
<v ->Awesome, thank you.</v>

1709
01:17:17.720 --> 01:17:20.490
I wanted to ask about one phrase I heard both of you use

1710
01:17:20.490 --> 01:17:23.500
and that was the lens that I look at the data through.

1711
01:17:23.500 --> 01:17:25.430
You both kind of mentioned like your individual lens,

1712
01:17:25.430 --> 01:17:26.670
so I was wondering if you could like speak

1713
01:17:26.670 --> 01:17:29.370

to that a little bit more, kind of your unique perspective

1714

01:17:29.370 --> 01:17:30.940
on your tackling these problems

1715

01:17:30.940 --> 01:17:33.163
and how that maybe has evolved over time.

1716

01:17:34.640 --> 01:17:36.513
<v ->So I guess, I can start.</v>

1717

01:17:40.483 --> 01:17:42.080
I think the the way I think about data

1718

01:17:42.080 --> 01:17:44.820
or the lens through which I look at data is very much shaped

1719

01:17:44.820 --> 01:17:48.405
by my time working at the NGO at HRDAG

1720

01:17:48.405 --> 01:17:52.280
where we really were sort of trying to think

1721

01:17:52.280 --> 01:17:53.590
of data from a different perspective.

1722

01:17:53.590 --> 01:17:55.770
I think often that lens is shaped

1723

01:17:55.770 --> 01:17:58.270
by one of the other things I was talking about earlier,

1724

01:17:58.270 --> 01:18:01.180
sort of diving deep into a substantive area,

1725

01:18:01.180 --> 01:18:04.700
trying to understand what the issues are there,

1726

01:18:04.700 --> 01:18:06.930

thinking about how the data was generated

1727

01:18:07.770 --> 01:18:11.433

and taking seriously concerns from people,

1728

01:18:11.433 --> 01:18:13.470

like when we're talking about, for example,

1729

01:18:13.470 --> 01:18:15.930

data that's going to be used to make recommendations

1730

01:18:15.930 --> 01:18:19.450

about real humans, taking seriously the current concerns

1731

01:18:19.450 --> 01:18:22.240

of the people about whom those recommendations will be made.

1732

01:18:22.240 --> 01:18:25.120

Right, so looking at the data, trying at least look

1733

01:18:25.120 --> 01:18:27.160

at the data from their perspective, often I'm not the person

1734

01:18:27.160 --> 01:18:30.580

at the sharp end of these models, in fact, almost never.

1735

01:18:30.580 --> 01:18:32.590

And so I can't say that I can fully adopt

1736

01:18:32.590 --> 01:18:34.650

someone's perspective, I don't know what it's like,

1737

01:18:34.650 --> 01:18:39.650

for example, to be a person evaluated for recidivism risk,

1738

01:18:40.000 --> 01:18:41.450

that's just not my experience,

1739

01:18:41.450 --> 01:18:43.660

but listening is as well as you can

1740

01:18:43.660 --> 01:18:45.620

and taking seriously the concerns

1741

01:18:45.620 --> 01:18:46.870

and incorporating those concerns

1742

01:18:46.870 --> 01:18:49.603

or that perspective into the data analysis.

1743

01:18:50.780 --> 01:18:53.473

<v ->For me, my lens has been coming</v>

1744

01:18:57.567 --> 01:18:59.880

from two lenses, maybe three,

1745

01:18:59.880 --> 01:19:02.240

one lens is coming from the communities

1746

01:19:02.240 --> 01:19:04.530

that are highly stigmatized

1747

01:19:04.530 --> 01:19:07.790

and highly stereotyped.

1748

01:19:07.790 --> 01:19:12.680

And knowing that I hear constantly those types

1749

01:19:12.680 --> 01:19:15.800

of stereotypes being, and data coming out

1750

01:19:17.306 --> 01:19:18.500

of what I should be

1751

01:19:18.500 --> 01:19:22.003

or if we go by the averages, what I would look like.

1752

01:19:23.140 --> 01:19:27.573

Also having a lens of being educated from an Ivy league,

1753

01:19:28.477 --> 01:19:31.180

what that means as well as you're not a researcher.

1754

01:19:31.180 --> 01:19:35.270

So trying to look at my data from various perspectives

1755

01:19:35.270 --> 01:19:36.451

as well as the gender,

1756

01:19:36.451 --> 01:19:40.023

the gender is an interesting lens to use.

1757

01:19:41.644 --> 01:19:42.530

And having, making sure

1758

01:19:42.530 --> 01:19:44.750

that that's why I like to have a diverse lab rule,

1759

01:19:44.750 --> 01:19:47.280

all different areas of peoples so that we can try

1760

01:19:47.280 --> 01:19:48.670

to get as many voices.

1761

01:19:48.670 --> 01:19:53.180

But more importantly, going back to participants.

1762

01:19:53.180 --> 01:19:56.180

I do a lot of ethics works where I ask participants

1763

01:19:56.180 --> 01:19:59.430

about the data, about the information we're collecting,

1764

01:19:59.430 --> 01:20:00.780
features they would like to have

1765

01:20:00.780 --> 01:20:02.980
to ensure privacy, confidentiality,

1766

01:20:02.980 --> 01:20:05.550
who they would like to have access to the data.

1767

01:20:05.550 --> 01:20:08.130
What types of apps are features

1768

01:20:08.130 --> 01:20:09.930
they would like built in of who,

1769

01:20:09.930 --> 01:20:13.060
when to turn on access and turn off access.

1770

01:20:13.060 --> 01:20:14.800
So I think it's key

1771

01:20:14.800 --> 01:20:17.700
to ask people who you are collecting data

1772

01:20:17.700 --> 01:20:20.890
from what they feel about it and what they want used,

1773

01:20:20.890 --> 01:20:24.580
and then to incorporate that into your research.

1774

01:20:24.580 --> 01:20:29.370
But also to let IRBs and regulators know this information

1775

01:20:29.370 --> 01:20:30.540
'cause they could be making demands

1776

01:20:30.540 --> 01:20:32.140

and you could say, no, no, no, my population,

1777

01:20:32.140 --> 01:20:34.540

they literally said they don't want that.

1778

01:20:34.540 --> 01:20:36.170

And so applying it that way.

1779

01:20:36.170 --> 01:20:39.890

I think it's key to include our participants

1780

01:20:39.890 --> 01:20:44.000

and people in our community and community advisory boards

1781

01:20:44.000 --> 01:20:46.470

in our research so that they can have a voice

1782

01:20:46.470 --> 01:20:47.303

and not necessarily like,

1783

01:20:47.303 --> 01:20:48.820

"We're just gonna give you a voice,"

1784

01:20:48.820 --> 01:20:51.830

but they need to have a real legitimate voice

1785

01:20:51.830 --> 01:20:55.270

and stake into our research and into what we're developing

1786

01:20:55.270 --> 01:20:56.570

and how we disseminate it.

1787

01:20:58.380 --> 01:20:59.930

<v ->Awesome, thank you very much.</v>

1788

01:21:00.870 --> 01:21:03.057

Another question was for you, Dr. Curtis specifically,

1789

01:21:03.057 --> 01:21:04.570

"Are you active on Twitter?"

1790

01:21:04.570 --> 01:21:06.220

Does your lab have a Twitter presence?

1791

01:21:06.220 --> 01:21:10.287

How does that social media outreach come from you, guys?"

1792

01:21:11.920 --> 01:21:16.653

<v ->My lab currently here, we are not active on Twitter.</v>

1793

01:21:19.130 --> 01:21:22.067

Being part of NIH and NIDA, I just came in 2019,

1794

01:21:22.067 --> 01:21:24.620

and I'm still trying to figure out the whole communication

1795

01:21:24.620 --> 01:21:26.690

of what we can distribute versus not.

1796

01:21:26.690 --> 01:21:30.710

So what I typically do is we make things,

1797

01:21:30.710 --> 01:21:34.147

and then we will let our partners and our collaborators,

1798

01:21:34.147 --> 01:21:36.960

and a lot of times they put stuff out.

1799

01:21:36.960 --> 01:21:39.390

So we have not done it.

1800

01:21:39.390 --> 01:21:41.030

I don't even know if we're allowed to have

1801

01:21:41.030 --> 01:21:43.390
our own lab Twitter page.

1802

01:21:43.390 --> 01:21:45.068
I actually think the answer is no.

1803

01:21:45.068 --> 01:21:45.901
(chuckle loudly)

1804

01:21:45.901 --> 01:21:47.170
I'm not sure Wilson will say,

1805

01:21:47.170 --> 01:21:48.461
yeah, the answer is no.

1806

01:21:48.461 --> 01:21:50.670
I will let you know though, but yeah,

1807

01:21:50.670 --> 01:21:52.490
I haven't read their bioethics.

1808

01:21:52.490 --> 01:21:54.500
<v ->I'm not totally sure, I'll find out.</v>

1809

01:21:54.500 --> 01:21:55.452
It's a great question.

1810

01:21:55.452 --> 01:21:56.285
<v ->I think the answer is...</v>

1811

01:21:56.285 --> 01:21:58.470
<v ->Intramural is a little different than extramural,</v>

1812

01:21:58.470 --> 01:21:59.940
but we'll find out.

1813

01:21:59.940 --> 01:22:01.550

<v Brenda>Yeah.</v>

1814

01:22:01.550 --> 01:22:04.103

<v ->Great, thanks, I think Albert has a question.</v>

1815

01:22:05.977 --> 01:22:06.927

Yup, there you are.

1816

01:22:07.920 --> 01:22:12.920

<v ->Thanks to both of you guys for your talk</v>

1817

01:22:13.710 --> 01:22:15.290

in your career journey.

1818

01:22:15.290 --> 01:22:18.300

One of the things that I would like you, guys,

1819

01:22:18.300 --> 01:22:21.270

to both address so that we could share

1820

01:22:21.270 --> 01:22:25.660

this with investigators down the road is

1821

01:22:25.660 --> 01:22:28.200

you've touched upon you have to be passionate,

1822

01:22:28.200 --> 01:22:31.710

and you have to really be interested in your field,

1823

01:22:31.710 --> 01:22:33.760

but you guys came from two different places

1824

01:22:33.760 --> 01:22:34.810

and your journeys have been

1825

01:22:34.810 --> 01:22:36.773

really different and interesting.

1826
01:22:37.980 --> 01:22:40.040
I would like to hear if you guys could share

1827
01:22:40.040 --> 01:22:41.580
actually like tangible

1828
01:22:41.580 --> 01:22:44.880
things that someone right now who's an undergrad

1829
01:22:44.880 --> 01:22:47.140
who may be kind of wondering if data science

1830
01:22:47.140 --> 01:22:49.520
or computer science is something for them,

1831
01:22:49.520 --> 01:22:51.040
what are tangibles that they could do

1832
01:22:51.040 --> 01:22:53.580
besides following your passion and your gut

1833
01:22:53.580 --> 01:22:54.460
and things like that,

1834
01:22:54.460 --> 01:22:56.270
what are the things that they could take?

1835
01:22:56.270 --> 01:22:58.290
People who may not have the same sort

1836
01:22:58.290 --> 01:23:01.620
of opportunities that others may have, what could they do

1837
01:23:04.700 --> 01:23:06.380
<v ->Should I start?</v>

1838
01:23:06.380 --> 01:23:09.640

I think one thing that people could do to sort of test

1839

01:23:09.640 --> 01:23:12.918

this out and this I'm not entirely obeying the question

1840

01:23:12.918 --> 01:23:15.780

by saying, by veering away from like find

1841

01:23:15.780 --> 01:23:18.280

something you're passionate about, but learn to code.

1842

01:23:18.280 --> 01:23:20.650

Like I think that is obviously a big part

1843

01:23:20.650 --> 01:23:22.870

of this career path.

1844

01:23:22.870 --> 01:23:25.610

And I personally find it really miserable trying

1845

01:23:25.610 --> 01:23:28.120

to learn a new language when you're just like walking

1846

01:23:28.120 --> 01:23:31.230

through a book and like copying in the commands

1847

01:23:31.230 --> 01:23:32.970

and then just fiddling with them.

1848

01:23:32.970 --> 01:23:36.380

I find the best way to do it is to find a question

1849

01:23:36.380 --> 01:23:37.290

you really wanna answer

1850

01:23:37.290 --> 01:23:39.230

and some data that can help you answer it

1851
01:23:39.230 --> 01:23:41.750
or maybe something you really wanna build

1852
01:23:41.750 --> 01:23:44.880
and maybe requires data, maybe it doesn't.

1853
01:23:44.880 --> 01:23:46.000
But if it does require data,

1854
01:23:46.000 --> 01:23:47.050
the data you'd need to build it.

1855
01:23:47.050 --> 01:23:48.150
And actually just try,

1856
01:23:49.250 --> 01:23:50.750
even if you don't have a lot of resources

1857
01:23:50.750 --> 01:23:52.670
there are a lot of free resources on the Internet.

1858
01:23:52.670 --> 01:23:54.960
I'm not even gonna like, just Google it.

1859
01:23:54.960 --> 01:23:56.580
There's tons of stuff out there.

1860
01:23:56.580 --> 01:23:59.170
There are a lot of free programming languages,

1861
01:23:59.170 --> 01:24:01.910
I think in the past when people were all using SAS

1862
01:24:01.910 --> 01:24:03.650
and everything and that's expensive

1863
01:24:03.650 --> 01:24:06.670

and you needed university access if you were a student.

1864

01:24:06.670 --> 01:24:09.660

Now, you can just use R, now there's Python.

1865

01:24:09.660 --> 01:24:12.160

There's all sorts of tools out there that are free

1866

01:24:13.479 --> 01:24:15.880

and there are great resources to learn as well.

1867

01:24:15.880 --> 01:24:19.500

So I think if you really want to figure out

1868

01:24:19.500 --> 01:24:22.340

if data science is a good career path for you,

1869

01:24:22.340 --> 01:24:25.570

I would just start by like, yeah, figuring out a project

1870

01:24:25.570 --> 01:24:29.850

you wanna do and just give it a shot, just go around browse.

1871

01:24:29.850 --> 01:24:31.990

People will explain things differently depending

1872

01:24:31.990 --> 01:24:32.970

on the resource you find,

1873

01:24:32.970 --> 01:24:36.020

find somebody who's explanations work for you,

1874

01:24:36.020 --> 01:24:36.853

work through it

1875

01:24:36.853 --> 01:24:38.500

and see it'll be really frustrating at first.

1876

01:24:38.500 --> 01:24:40.790

I remember when I first came to programming

1877

01:24:40.790 --> 01:24:43.950

I was in a class and it was like, i equals i plus one

1878

01:24:43.950 --> 01:24:48.950

and I was like false, no, unless it's infinity, no,

1879

01:24:48.950 --> 01:24:51.610

that's not how math works and having to like work through,

1880

01:24:51.610 --> 01:24:54.000

sort of rewire your brain to think,

1881

01:24:54.000 --> 01:24:56.490

okay, no, I'm actually assigning it.

1882

01:24:56.490 --> 01:24:57.580

Yeah. It'll be frustrating.

1883

01:24:57.580 --> 01:24:59.780

It'll be like banging your head against a wall,

1884

01:24:59.780 --> 01:25:00.630

I think at times.

1885

01:25:00.630 --> 01:25:04.070

But if you enjoy that kind of like head banging,

1886

01:25:04.070 --> 01:25:07.940

brain twisting sort of work, this is probably a good place.

1887

01:25:07.940 --> 01:25:08.780

<v ->Thanks Dr. Lum.</v>

1888

01:25:08.780 --> 01:25:09.920

So give it a try and see

1889

01:25:09.920 --> 01:25:11.930

if that's something that you're pulled to,

1890

01:25:11.930 --> 01:25:13.670

gravitate towards, that sounds great.

1891

01:25:13.670 --> 01:25:14.970

What about you, Dr. Curtis?

1892

01:25:14.970 --> 01:25:17.350

<v ->Well, actually, one of the things I didn't include is</v>

1893

01:25:17.350 --> 01:25:20.750

when I was in high school, my dad saw

1894

01:25:20.750 --> 01:25:22.570

or he heard some people at work talking

1895

01:25:22.570 --> 01:25:24.420

about a computer programming class

1896

01:25:24.420 --> 01:25:27.100

that the college was doing for high school students.

1897

01:25:27.100 --> 01:25:30.330

So of course, that's how I spent my Saturday afternoons.

1898

01:25:30.330 --> 01:25:34.240

So I did learn initially to do some basic coding

1899

01:25:34.240 --> 01:25:37.720

'cause my dad was like computers are gonna are the future.

1900

01:25:37.720 --> 01:25:41.930

And so I did that, and then when I got my job

1901

01:25:41.930 --> 01:25:43.480
at the drug treatment center

1902

01:25:43.480 --> 01:25:45.290
where I was doing HIV counseling,

1903

01:25:45.290 --> 01:25:49.050
the first thing I developed, (chuckles) do you, guys,

1904

01:25:49.050 --> 01:25:50.460
know those like cosmic quiz

1905

01:25:50.460 --> 01:25:52.440
like what type of person are you?

1906

01:25:52.440 --> 01:25:54.763
I did one where it was like,

1907

01:25:56.780 --> 01:25:59.670
it was about your sexual activity and your risk.

1908

01:25:59.670 --> 01:26:02.500
So I was asking questions like what type of sex do like,

1909

01:26:02.500 --> 01:26:05.520
how do you it, all the different flavors.

1910

01:26:05.520 --> 01:26:07.370
And then it was like a risk score

1911

01:26:07.370 --> 01:26:09.730
and depending on their different behaviors,

1912

01:26:09.730 --> 01:26:11.080
it would assign risks.

1913

01:26:11.080 --> 01:26:13.280

And so I would do that instead of doing,

1914

01:26:13.280 --> 01:26:15.530

we had this boring regular questionnaire.

1915

01:26:15.530 --> 01:26:17.018

So I would still have them do the questionnaire

1916

01:26:17.018 --> 01:26:18.600

because I have to have it for documentation.

1917

01:26:18.600 --> 01:26:21.780

But then I put people in a computer program, they do it.

1918

01:26:21.780 --> 01:26:24.990

And then we would talk about this in harm reduction.

1919

01:26:24.990 --> 01:26:27.643

So for me, while I'm not a computer programmer

1920

01:26:27.643 --> 01:26:29.540

or a data scientist, in that sense,

1921

01:26:29.540 --> 01:26:31.280

you had to use some of those tools.

1922

01:26:31.280 --> 01:26:34.850

It was more like how to get people engaged

1923

01:26:34.850 --> 01:26:38.370

in the content to assess risk, to assess things,

1924

01:26:38.370 --> 01:26:40.580

and then to deliver something that was cool

1925

01:26:40.580 --> 01:26:42.350

and innovative that would keep them engaged

1926

01:26:42.350 --> 01:26:44.143

so that we could work on treatment.

1927

01:26:47.110 --> 01:26:50.270

<v ->So then the advice would be to those young...</v>

1928

01:26:50.270 --> 01:26:53.060

<v ->My advice would be to have fun with it.</v>

1929

01:26:53.060 --> 01:26:55.820

You're not gonna wanna do something that's boring.

1930

01:26:55.820 --> 01:26:58.270

You do need to learn the skills, which is sometime boring,

1931

01:26:58.270 --> 01:27:00.120

but then apply it to what you wanna do.

1932

01:27:00.120 --> 01:27:00.953

I mean, I don't know,

1933

01:27:00.953 --> 01:27:04.050

even you wanna play with what Pokemon character are you

1934

01:27:04.050 --> 01:27:05.210

or whatever it is,

1935

01:27:05.210 --> 01:27:07.640

apply it to something that you're interested in

1936

01:27:07.640 --> 01:27:10.280

so that you can just keep getting the skills.

1937

01:27:10.280 --> 01:27:13.970

You do need the skills, but fortunately those skills,

1938

01:27:13.970 --> 01:27:15.860

a lot of times, they're not reliant

1939

01:27:15.860 --> 01:27:17.630

of one of your social economic class.

1940

01:27:17.630 --> 01:27:18.830

As you plan out,

1941

01:27:18.830 --> 01:27:21.130

there are lots of free things that you can do.

1942

01:27:22.002 --> 01:27:24.020

Well, I'm not gonna list a couple of the programs either,

1943

01:27:24.020 --> 01:27:27.060

but there's lots of stuff that have broken

1944

01:27:27.060 --> 01:27:28.740

those barriers of income.

1945

01:27:28.740 --> 01:27:30.150

And if you have a smart phone, some of them,

1946

01:27:30.150 --> 01:27:32.360

you can even just use your smart phone and code on.

1947

01:27:32.360 --> 01:27:34.810

So go after it, have fun,

1948

01:27:34.810 --> 01:27:37.750

apply it to something you're interested in.

1949

01:27:37.750 --> 01:27:38.741

<v ->Thanks to you both,</v>

1950

01:27:38.741 --> 01:27:41.093

I think that's very, very useful info.

1951
01:27:44.410 --> 01:27:45.243
<v ->Great, thank you.</v>

1952
01:27:45.243 --> 01:27:46.740
So we had one question come in

1953
01:27:46.740 --> 01:27:50.580
and it was about with all of the different ways

1954
01:27:50.580 --> 01:27:52.463
to gather data and to survey,

1955
01:27:53.480 --> 01:27:56.320
how do you decide what is a high priority

1956
01:27:56.320 --> 01:27:58.010
for you to spend your time on?

1957
01:27:58.010 --> 01:27:59.950
How do you decide what kind

1958
01:27:59.950 --> 01:28:00.980
of questions do you wanna tackle?

1959
01:28:00.980 --> 01:28:02.423
What's a high priority?

1960
01:28:03.360 --> 01:28:04.550
Where does that drive come from

1961
01:28:04.550 --> 01:28:06.000
or how do you narrow it down?

1962
01:28:09.020 --> 01:28:10.730
<v ->Okay, I'll start.</v>

1963
01:28:10.730 --> 01:28:14.030

I go for clinical relevance.

1964

01:28:14.030 --> 01:28:18.430

What I think is going to have a high impact in the clinic

1965

01:28:18.430 --> 01:28:23.430

as well as what could have a high level

1966

01:28:24.240 --> 01:28:27.510

nationally kind of surveillance tool

1967

01:28:27.510 --> 01:28:30.820

and it's quick and easy and something that can apply to them

1968

01:28:30.820 --> 01:28:33.463

at a level from the community to the nation.

1969

01:28:34.550 --> 01:28:39.550

Also has some indicators that could not be applied well,

1970

01:28:40.000 --> 01:28:41.650

but so I kind of go from there,

1971

01:28:41.650 --> 01:28:45.173

more of an application use case scenario.

1972

01:28:47.050 --> 01:28:49.030

<v ->Yeah, I think I borrow some of my thinking</v>

1973

01:28:49.030 --> 01:28:52.930

on this from my time at HRDAG where there are a suite

1974

01:28:52.930 --> 01:28:54.860

of questions that we asked ourselves there.

1975

01:28:54.860 --> 01:28:56.290

I don't even honestly remember

1976

01:28:56.290 --> 01:28:59.540

all of them off the top of my head, but one of them is

1977

01:28:59.540 --> 01:29:02.610

about how can my skills actually contribute to this?

1978

01:29:02.610 --> 01:29:04.450

Do I have something unique to say in this area?

1979

01:29:04.450 --> 01:29:06.380

So that's one of the things I ask myself.

1980

01:29:06.380 --> 01:29:10.080

And the other one is in this case say we do a great job

1981

01:29:10.080 --> 01:29:13.440

and we get to some sort of truth, does the truth matter?

1982

01:29:13.440 --> 01:29:15.890

And in some sense, the truth always matters,

1983

01:29:15.890 --> 01:29:17.750

just sort of as a value I hold,

1984

01:29:17.750 --> 01:29:19.700

the truth is important just in general,

1985

01:29:20.750 --> 01:29:21.583

but also does it matter?

1986

01:29:21.583 --> 01:29:22.680

Is it policy relevant?

1987

01:29:22.680 --> 01:29:25.100

Is this something where people know the truth,

1988

01:29:25.100 --> 01:29:27.940

they can act on it, they can maybe just sort of speaking

1989

01:29:27.940 --> 01:29:29.010

and generally do

1990

01:29:29.010 --> 01:29:31.140

something about it to make the world better?

1991

01:29:31.140 --> 01:29:32.670

Will this be relevant to policy makers,

1992

01:29:32.670 --> 01:29:34.600

will this actually move things forward?

1993

01:29:34.600 --> 01:29:37.840

And so those are sort of the two main questions

1994

01:29:37.840 --> 01:29:39.580

that I think I've, again, borrowed from my time

1995

01:29:39.580 --> 01:29:43.830

at HRDAG that helped me sort of whittle down all the

1996

01:29:43.830 --> 01:29:45.860

sort of infinite possibilities of projects

1997

01:29:45.860 --> 01:29:47.860

that one could work on at this point to try

1998

01:29:47.860 --> 01:29:52.793

to find one that I think is a good use of time and skill.

1999

01:29:54.050 --> 01:29:54.883

<v ->Thank you very much.</v>

2000

01:29:54.883 --> 01:29:56.600

Thank you for answering all of the questions

2001
01:29:56.600 --> 01:29:59.050
and for your presentations.

2002
01:29:59.050 --> 01:29:59.980
I see that we're out of time,

2003
01:29:59.980 --> 01:30:01.963
so I will turn it back over to Susan.

2004
01:30:04.350 --> 01:30:05.183
<v ->Thank you, Lindsey.</v>

2005
01:30:05.183 --> 01:30:08.363
I just wanna thank you again, Dr. Lum and Dr. Curtis.

2006
01:30:08.363 --> 01:30:10.200
It's been great hearing about your careers

2007
01:30:10.200 --> 01:30:11.120
and what you're working on.

2008
01:30:11.120 --> 01:30:13.720
And I think you've given a lot of great advice

2009
01:30:13.720 --> 01:30:15.870
to the the next generation of data scientists.

2010
01:30:15.870 --> 01:30:17.460
It's been really enjoyable this morning hearing

2011
01:30:17.460 --> 01:30:18.547
from both of you.

2012
01:30:18.547 --> 01:30:20.650
And I also just wanna announce that next week,

2013
01:30:20.650 --> 01:30:22.880

next Monday at 9:00 a.m. is our final session

2014

01:30:22.880 --> 01:30:24.890

of this four-part series.

2015

01:30:24.890 --> 01:30:26.500

We'll be hearing from Dan Jacobson

2016

01:30:26.500 --> 01:30:28.410

from Oak Ridge National Laboratory

2017

01:30:28.410 --> 01:30:30.120

and Mike Tamir from SIG

2018

01:30:30.120 --> 01:30:32.500

and he's also a faculty at UC Berkeley.

2019

01:30:32.500 --> 01:30:33.610

So please, go ahead and register.

2020

01:30:33.610 --> 01:30:36.020

I think the registration link is in the chat

2021

01:30:36.020 --> 01:30:37.610

and hopefully, we'll see you all again next week.

2022

01:30:37.610 --> 01:30:39.320

And thanks again to our speakers

2023

01:30:39.320 --> 01:30:40.406

and thanks again to the organizers

2024

01:30:40.406 --> 01:30:41.840

and everyone that's on the team.

2025

01:30:41.840 --> 01:30:43.630

And I know it's hard in the virtual environment,

2026

01:30:43.630 --> 01:30:44.560
but some virtual applause

2027

01:30:44.560 --> 01:30:45.940
for everybody for being a part

2028

01:30:45.940 --> 01:30:47.580
of this this morning.

2029

01:30:47.580 --> 01:30:48.430
Thanks, everyone.