

#CovidComplete

An Inside Look at Covid-19 Forecasting

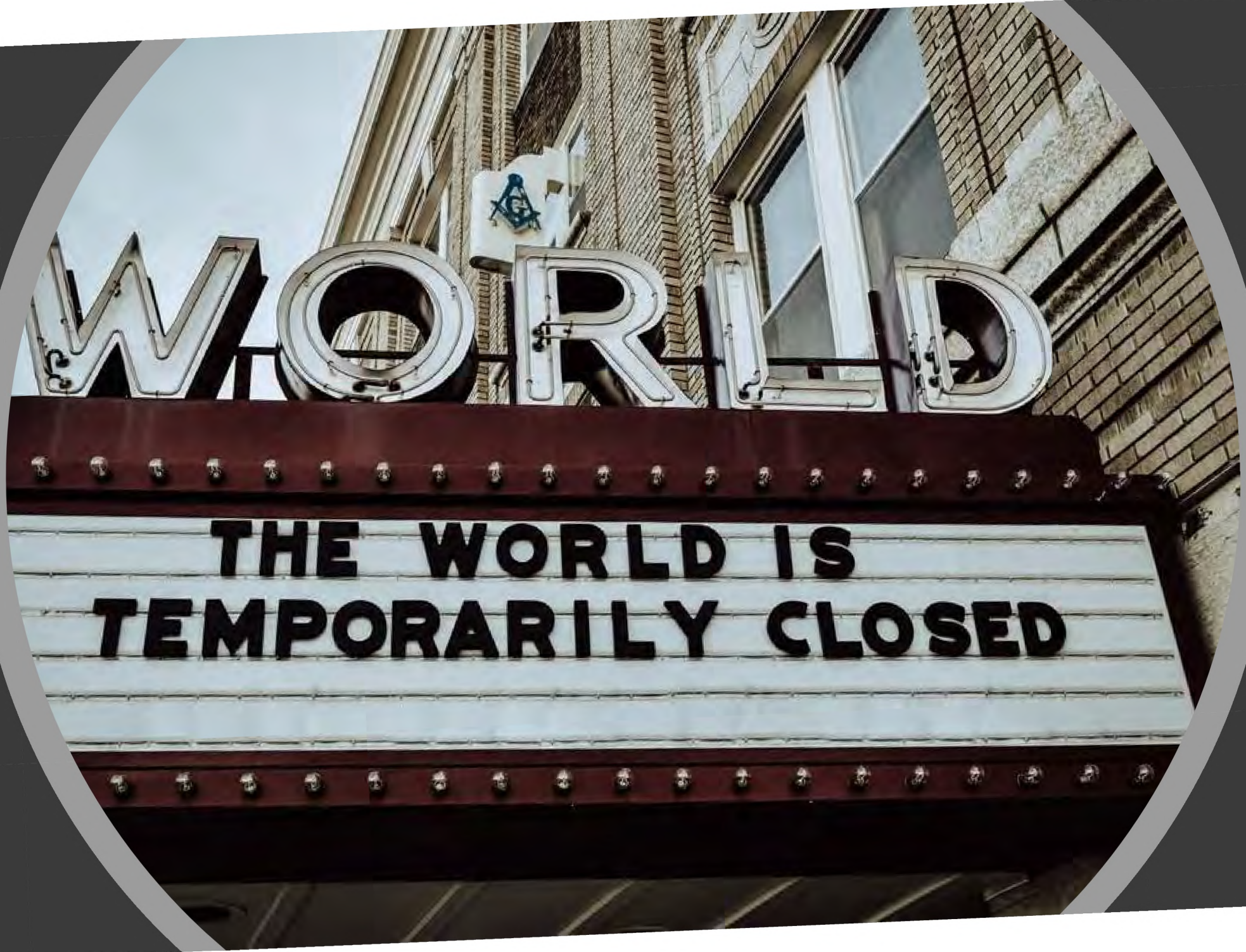
By Steve McConnell

— Copyright

© Copyright Steve McConnell.
All Rights Reserved.

These materials may be freely copied and distributed for personal, educational, and non-commercial purposes so long as the following attribution is included:

*These materials are © Copyright Steve McConnell.
Additional information is available at stevemcconnell.com/covid.*



**THE WORLD IS
TEMPORARILY CLOSED**

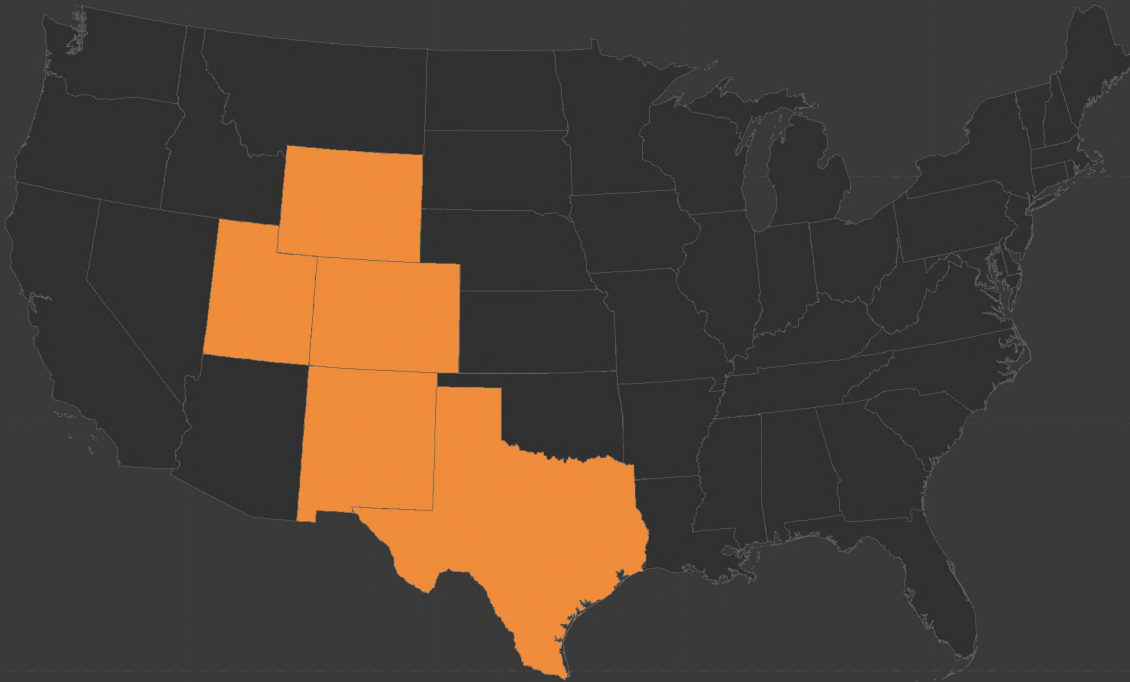
Goals for this Talk

- Diffuse emotion and politics; show why you should ignore IHME's forecasts
- Demystify forecasting (vs. speculation)
- Explore Covid-19 data issues; correct common data misreporting issues
- Provide background on the CDC's forecasting process
- Explain what's possible and what is not possible in Covid-19 forecasting
- Demonstrate state of the art of current Covid-19 forecasts

— Why Create State Forecasts?

- ▣ Guides near-term activity planning and event planning (1-4 weeks)
- ▣ Anticipates regional surges
- ▣ Identify areas that will need additional supplies, equipment, staff
- ▣ Identify areas that can spare supplies, equipment, staff
- ▣ Provides data-based feedback on policy decisions (as long as the forecast is based on data rather than the assumptions being tested)

— Why create National Forecasts?



Guides near-term activity and event planning (1-4 weeks)



Provides data-based feedback on policy decisions



Provides important context for state-level decisions (how much capacity is expected to be available nationally?)



Avoid panic (or create it)



My personal role in Covid-19 Forecasting

Microsoft

CODE COMPLETE

2
Second Edition



A practical handbook of software construction

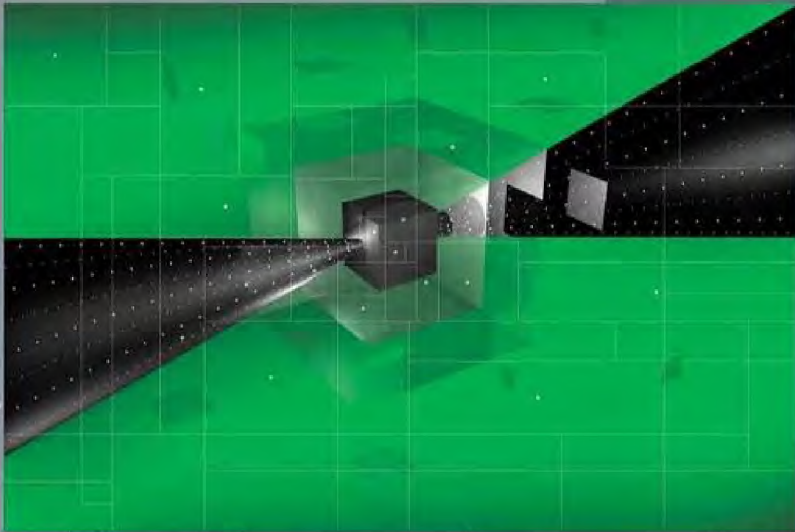
Steve McConnell

Two-time winner of the Software Development Magazine Jolt Award

“

For most of my career I have focused on understanding the data analytics of software development, including quality, productivity, and estimation. The techniques I've learned from working with noisy data, bad data, uncertainty, and forecasting all apply to COVID-19.

SOFTWARE ESTIMATION



Demystifying the Black Art

Steve McConnell

Two-time winner of *Software Development* magazine's Jolt Award



Experts tend to use simple estimation strategies, even when their level of expertise in the subject being estimated is high — *Steve McConnell, Software Estimation*



Centers for Disease Control and Prevention
CDC 24/7: Saving Lives, Protecting People™

[A-Z Index](#)

Search

COVID-19 ' Q

[Advanced Search](#) 103

Coronavirus Disease 2019 (COVID-19)



Your Health v Community, Work & School v Healthcare Workers & Labs v Health Depts v Cases & Data v More v

n Cases, Data & Surveillance

[US Cases & Deaths](#) +[Cases & Deaths by County](#)[Testing Data in the US](#) +[Hospitalizations &](#) +

CASES, DATA & SURVEILLANCE

COVID-19 Forecasts: Deaths

Updated Sept, 30, 2020

[Print](#)

0 0000

Observed and forecasted new and total reported COVID-19 deaths as of September 28, 2020.

Tutoiot—tin Af

Forecasts Af

[On This Page](#)

acdc.gov/coronavirLfs/2019-ncov/covid-dat3/forecasting-us.html

Personal Home Page O JHU Data ^ Covki-19 0 State Data 0 US Data V CDC Dropbox ^ Construe OnDemand S Cisco □ Offsite

» Other bookmarks

1 :

Special Populations Data

National Forecast

COVIDView Weekly

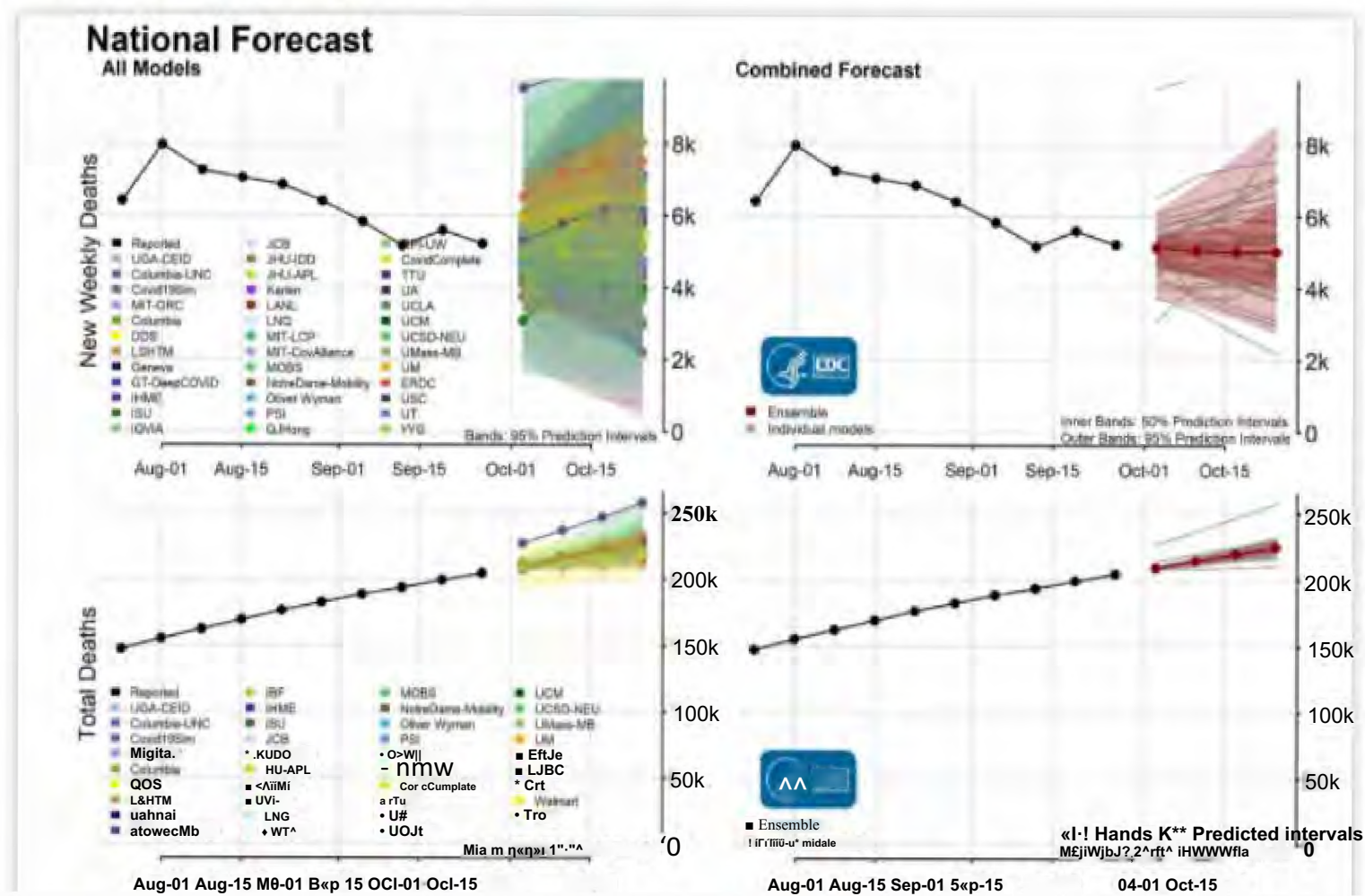
Summary

Sequencing for SARS-CoV-2 (SPHERES)

Epidemiology for COVID-19

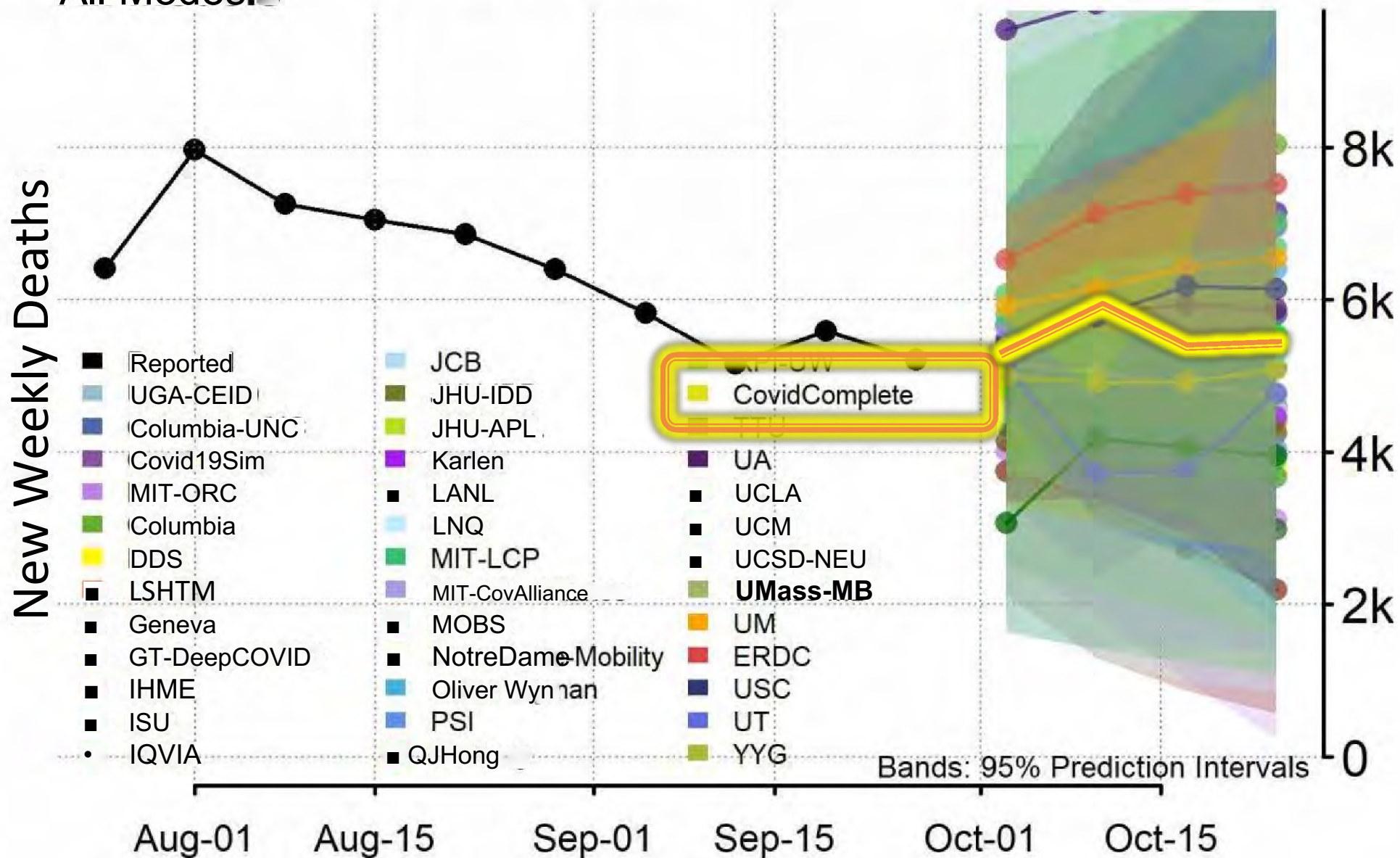
M Get Email Updates

To receive email updates about COVID-19, enter your email address:



National Forecast

All Models



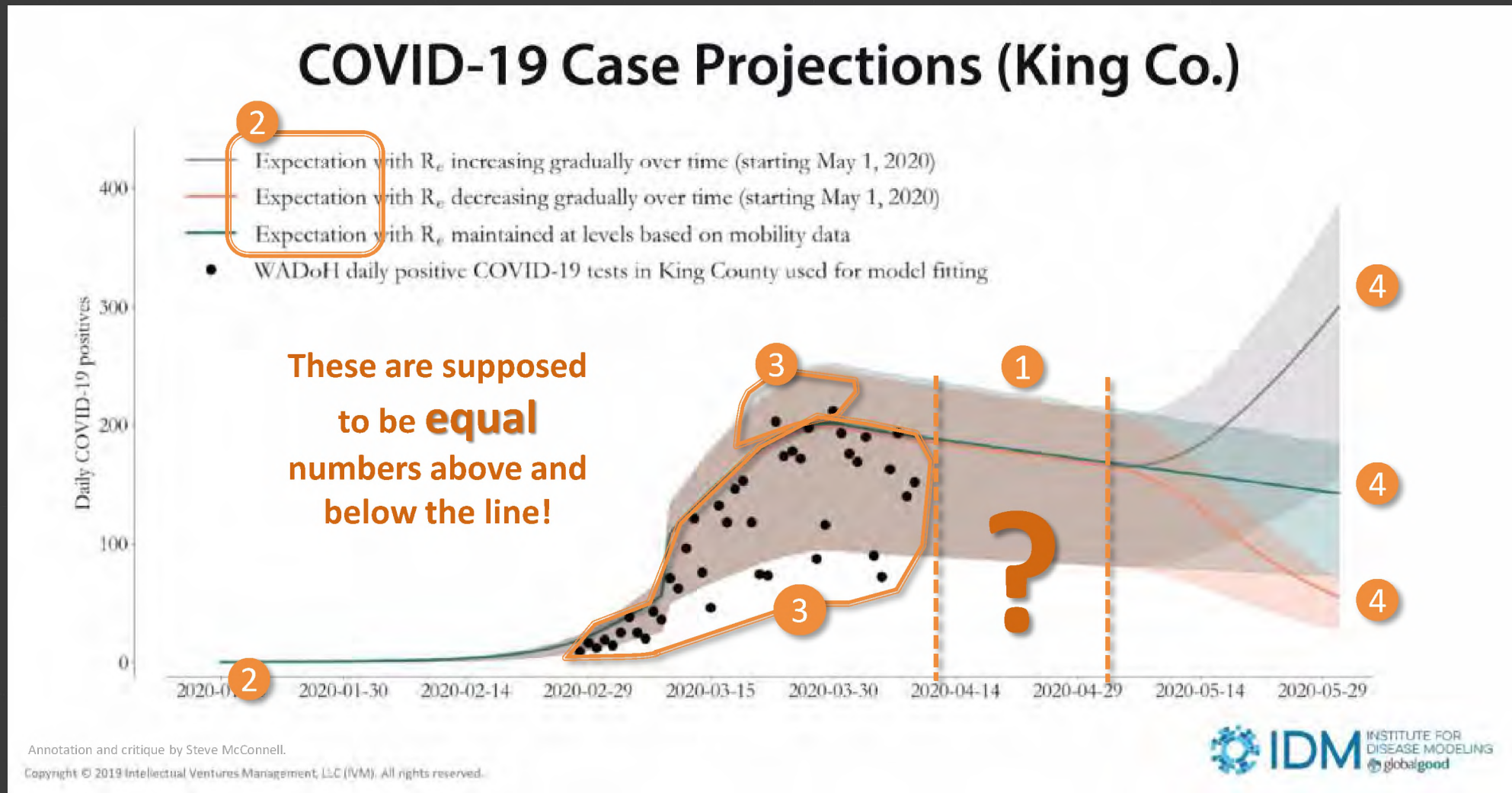


Why I got Involved

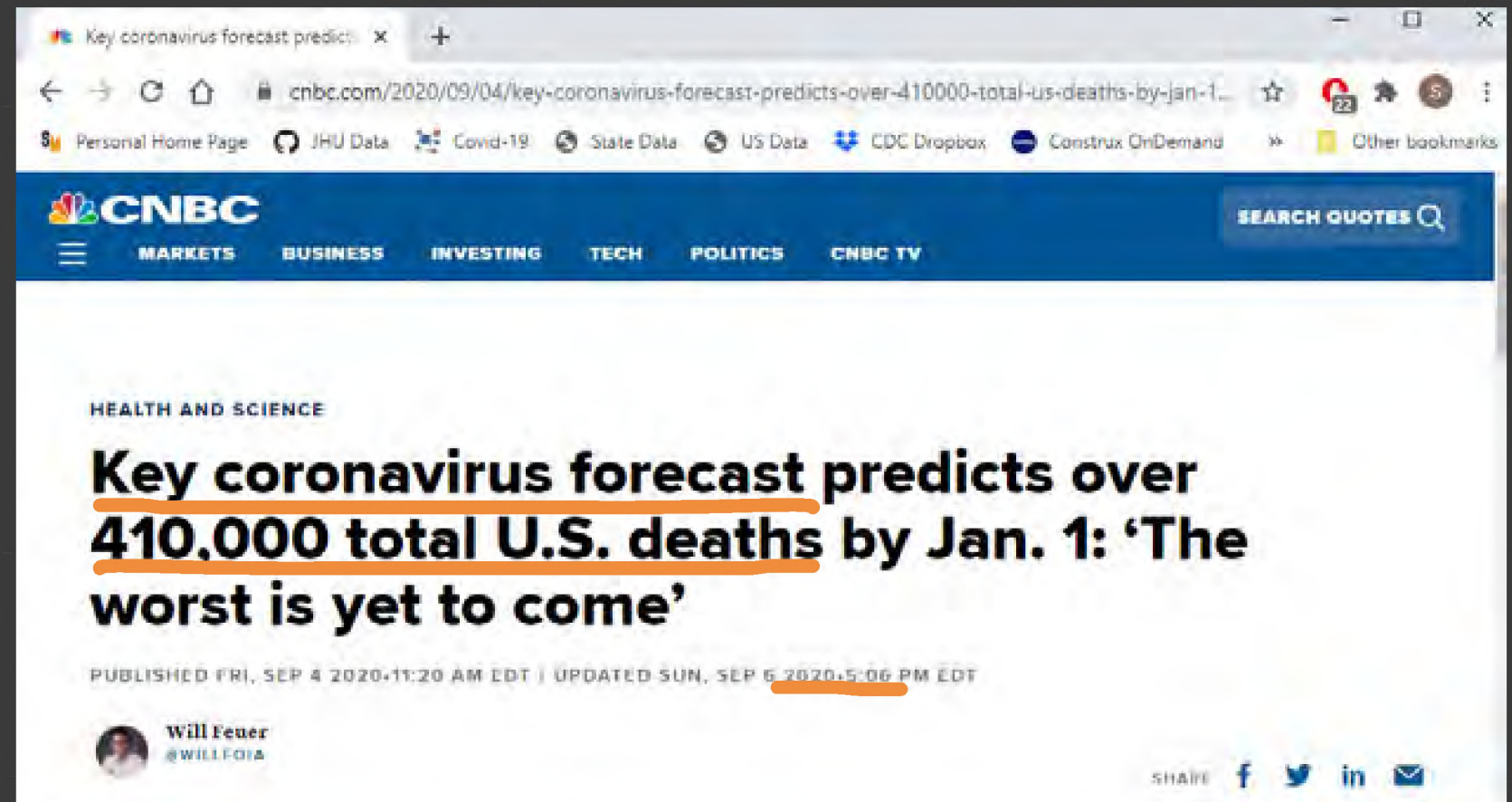


I don't like playing games with forecasts during a global pandemic

— Playing games with data – earlier

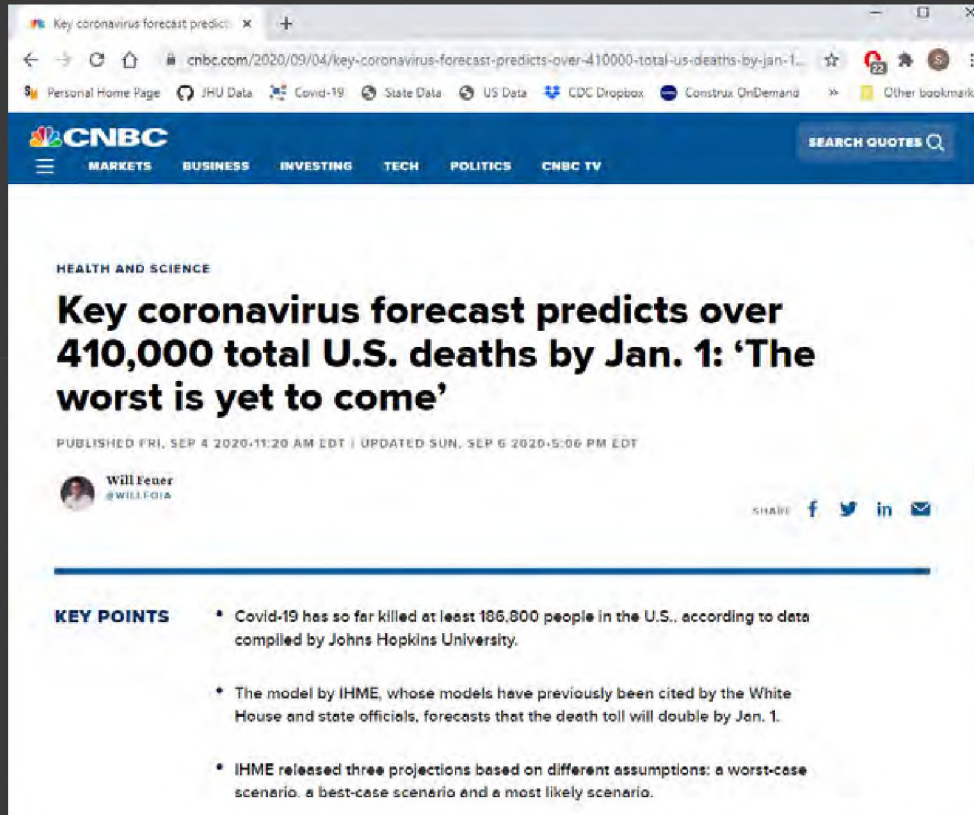


Playing games with Data, Recently



scenario, a best-case scenario and a most likely scenario. The most likely estimates that Covid-19 will kill 410,450 people in the U.S. by Jan. 1.

- The model by IHME, whose models have previously been cited by the White House and state officials, forecasts that the death toll will double by Jan. 1.
- IHME released three projections based on different assumptions: a worst-case scenario, a best-case scenario and a most likely scenario.

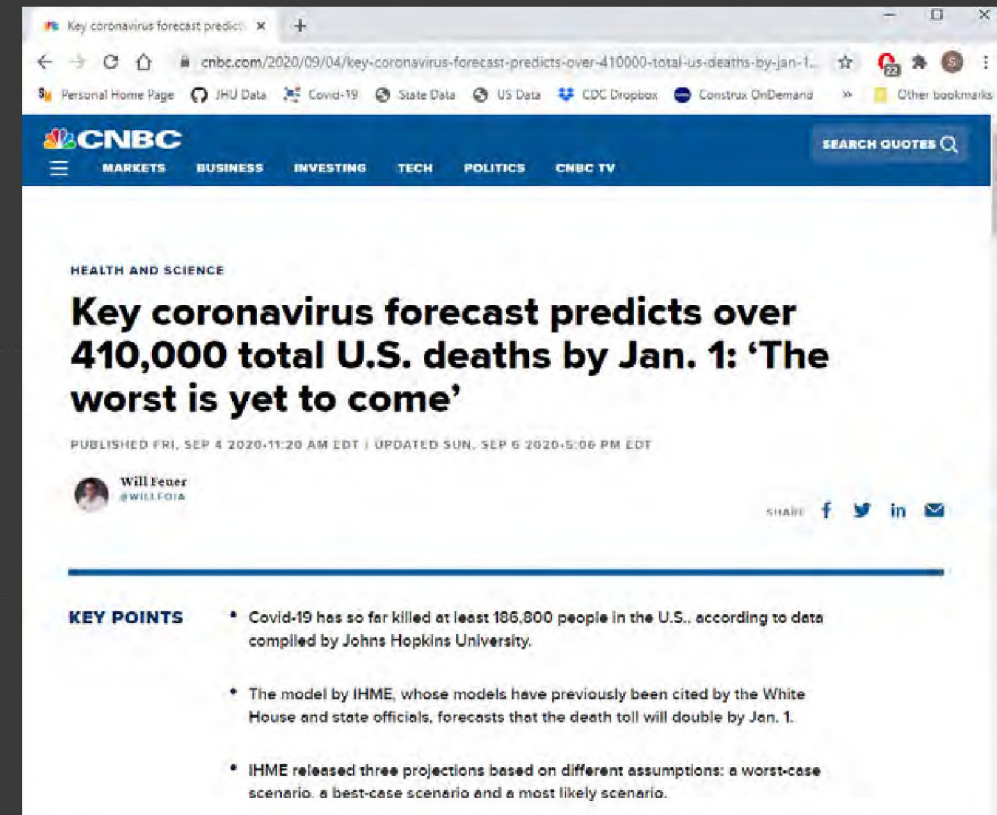


What would need to happen for this to be true?

- The death trend would need to more than double, immediately, despite trending down 10%/week for the past few weeks, along with leading indicators also trending down
- Need to incur more deaths in four months than we've had since the beginning of the pandemic

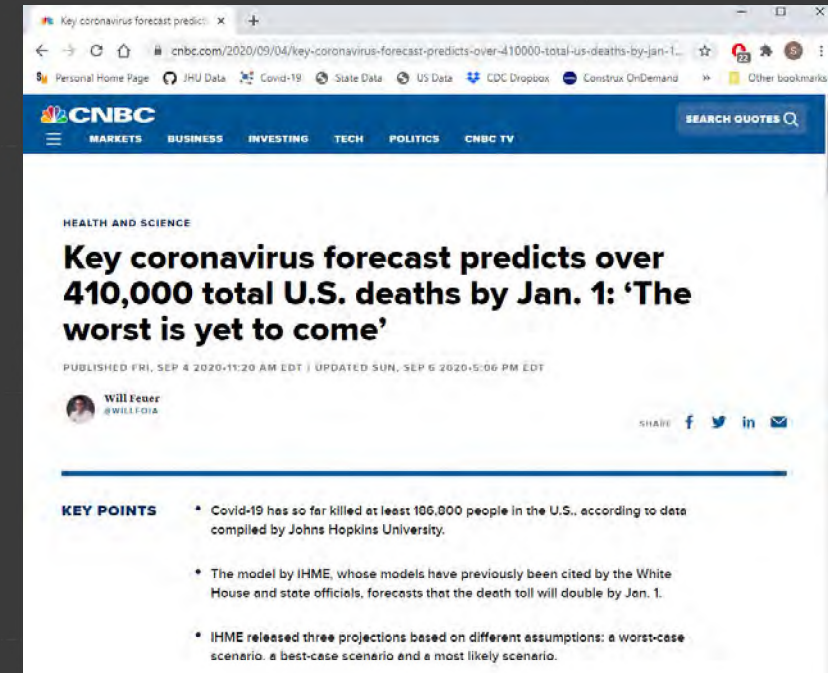
What would need to happen for this to be true?

- Need to average 1800 deaths per day for 115 days straight, which was 100 more days than we'd had that many deaths so far
- No one can take any corrective action to reduce deaths even after deaths skyrocket, and they must continue not to take any action for **4 months**

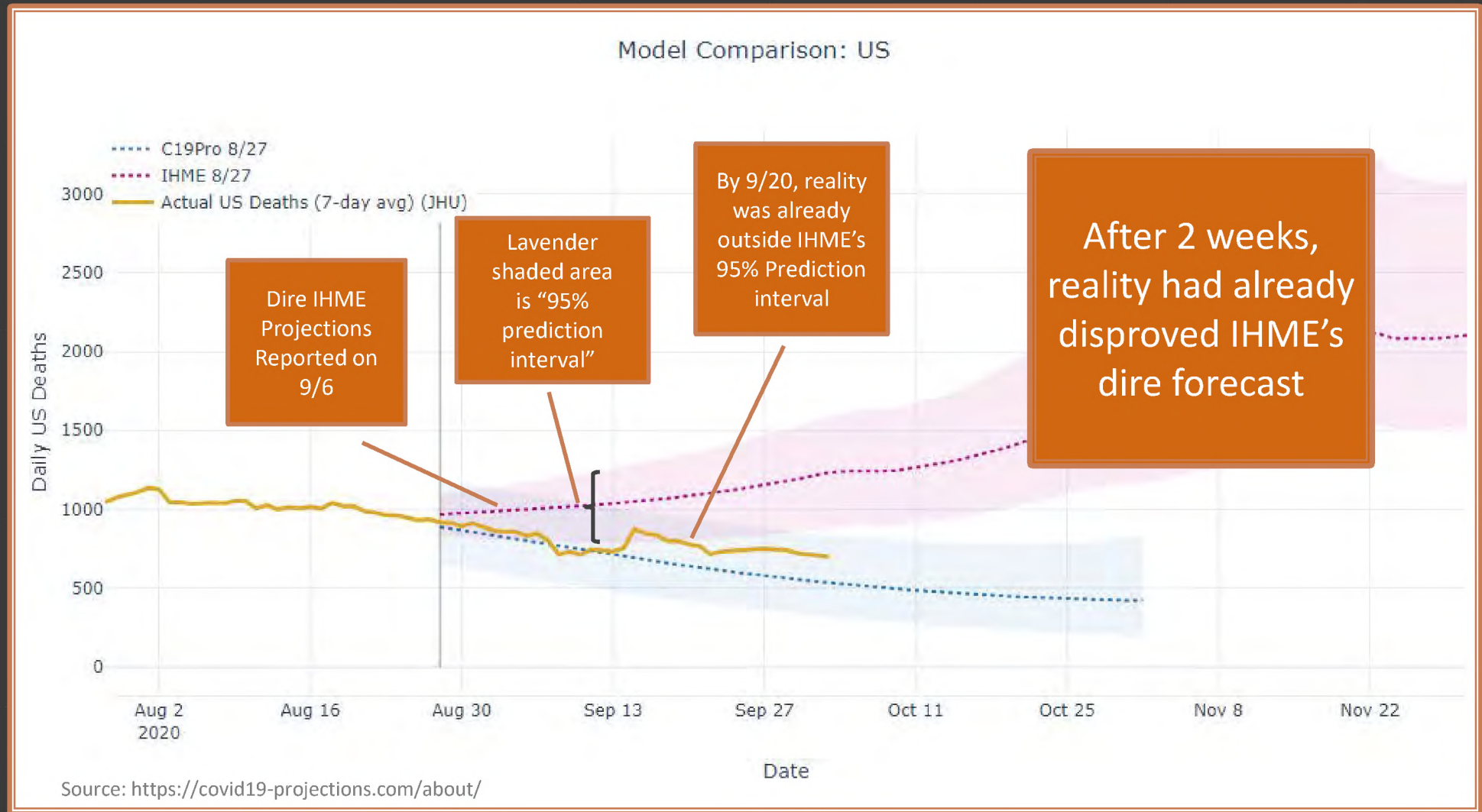


My Observations

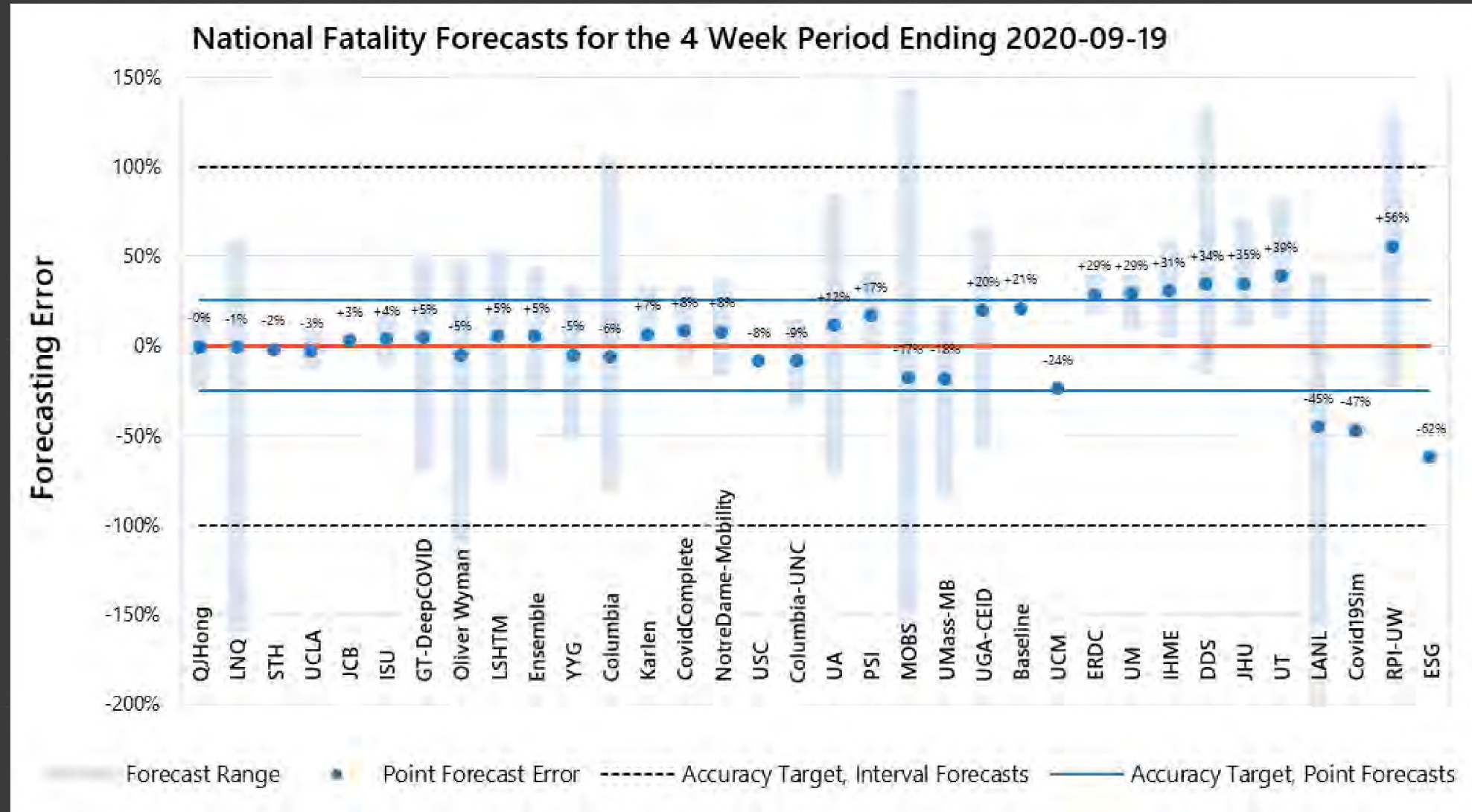
- ❑ This forecast is not going to happen
- ❑ This forecast is completely unfounded
- ❑ This forecast is in no way “most likely”



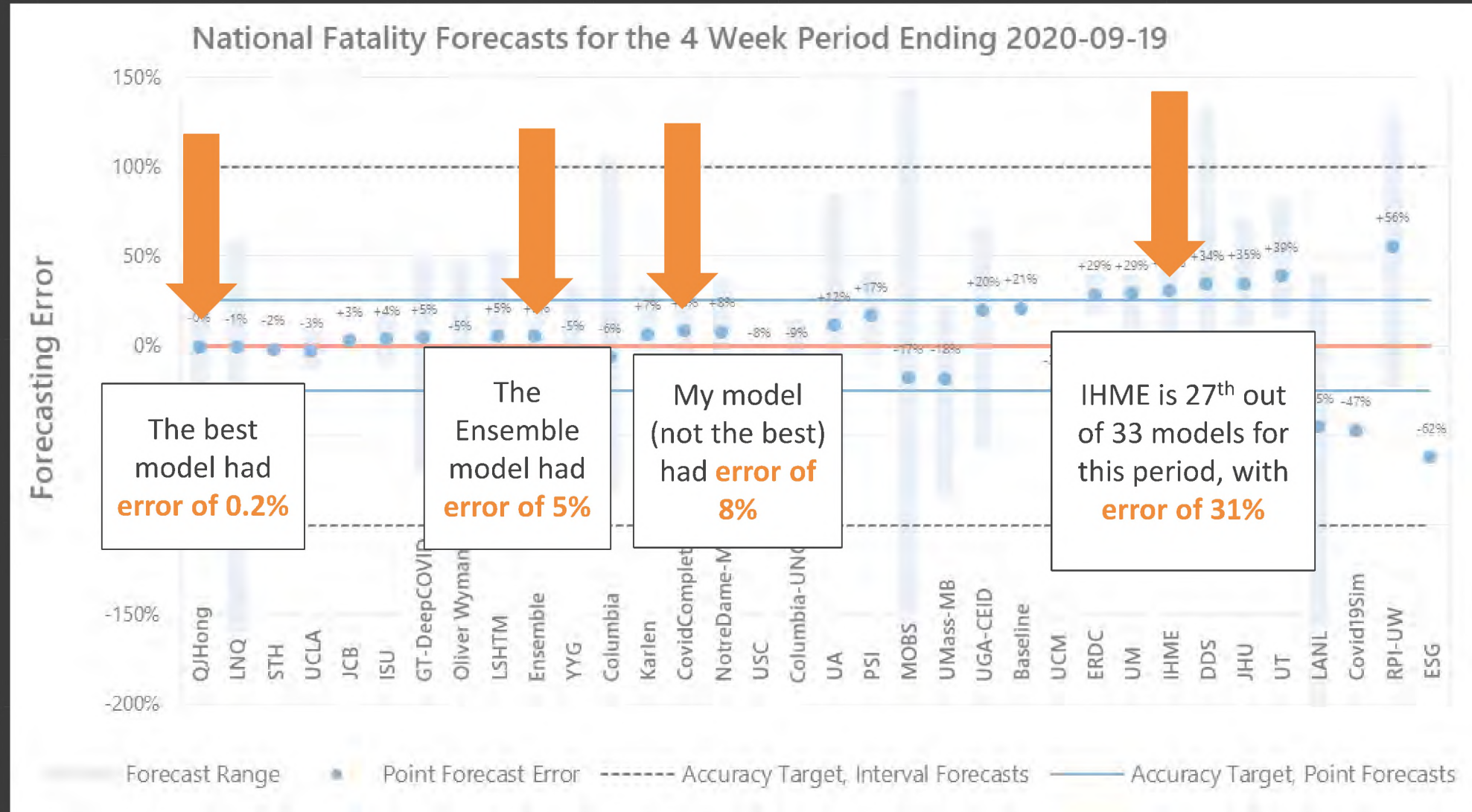
— This “Forecast” is completely unfounded



How Accurate were Other Forecast for the same period?



How Accurate were Other Forecasts for the same period?





Why I got Involved

I just wanted to know what is going on with
the pandemic, without all the spin



- The Basics: What is Forecasting?

Forecasting is
not the same as
speculation, but we are
seeing them treated
interchangeably



[PostEverything](#) • [Perspective](#)

Scientists want to predict covid-19's long-term trajectory. Here's why they can't.

Our research suggests that forecasts are unreliable further than a few weeks out



Dining outdoors in Manhattan on Monday. (Jeenah Moon/Reuters)

By **Nicholas Reich** and **Caitlin Rivers**
by
September 15, 2020 at 11:22 a.m. PDT

What is possible with forecasting

1-4 week time horizons

Accurate national forecasts

Pretty accurate state forecasts, especially for more active states

- Anything further out than that is speculation (even if it's dressed up with graphs and numbers, which most of it is)

Forecasting vs. Speculation

- **Forecasting**
 - Inputs are known
 - Outputs (forecasts) are calculated based on known relationships
- **Speculation**
 - Inputs are guesses
 - Outputs are guesses piled on top of other guesses



Knowns

Number of people who have already tested positive

Approximate relation between positive tests and cases

Approximate fatality rates by age and comorbidity status

Progression of the virus in individuals

Unknowns

Specific policies that specific states will implement (or relax) in the future

Dates those policies will be implemented or relaxed

Effectiveness of the policies, with good compliance

Effectiveness of the policies, with whatever compliance we actually get

Availability of vaccines

Effectiveness of vaccines, when and if available

The Basis of Forecasting
"What We Know"



We Know Much More About Covid-19 Now Than We Did on 1/22/20



Fatality is highly age-related (~50% of deaths are age 80+)



Fatality is highly co-morbidity related (~90% of fatalities involve at least one co-morbidity)



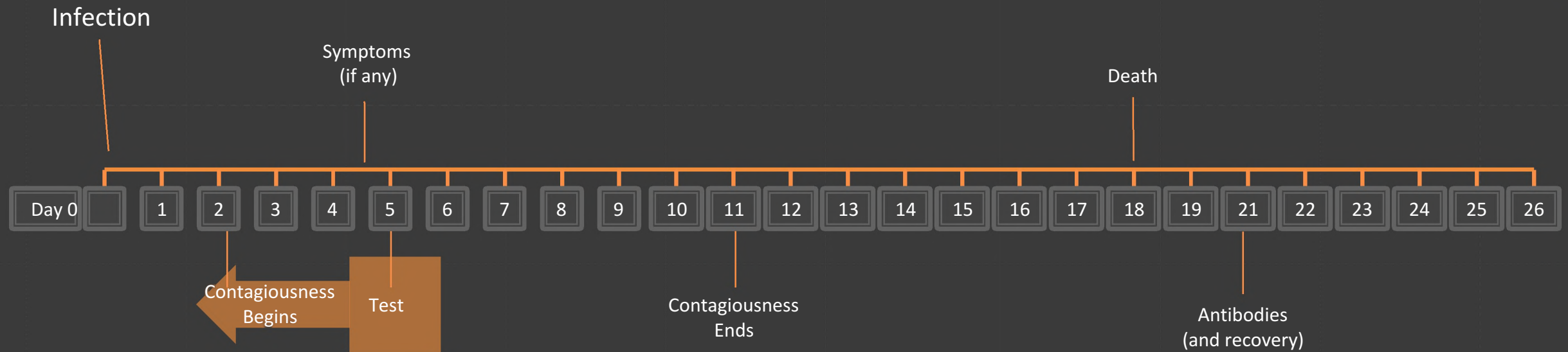
A lot more people are infected than have tested positive

- Early in the pandemic: 10-20x cases per positive test
- Now—3-5x cases per positive test

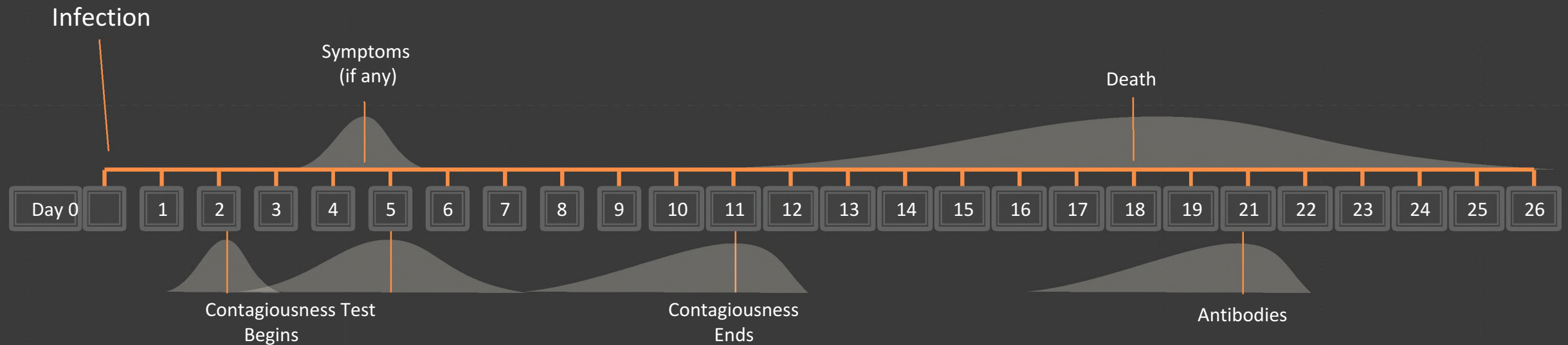


We know the timing of the course of the disease

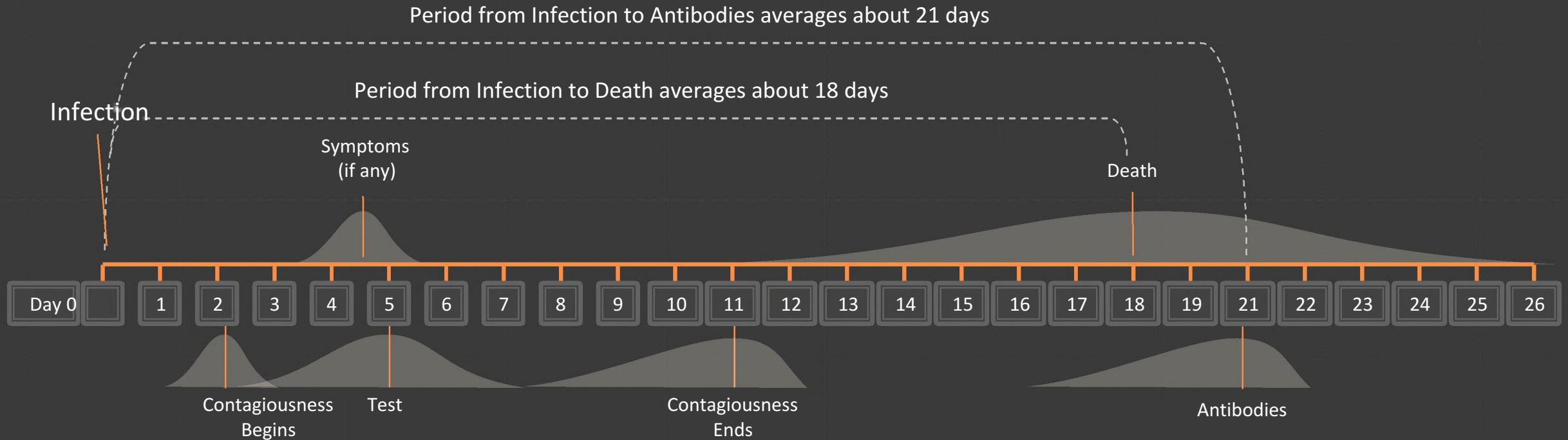
— We have a pretty good view of the timing of the disease



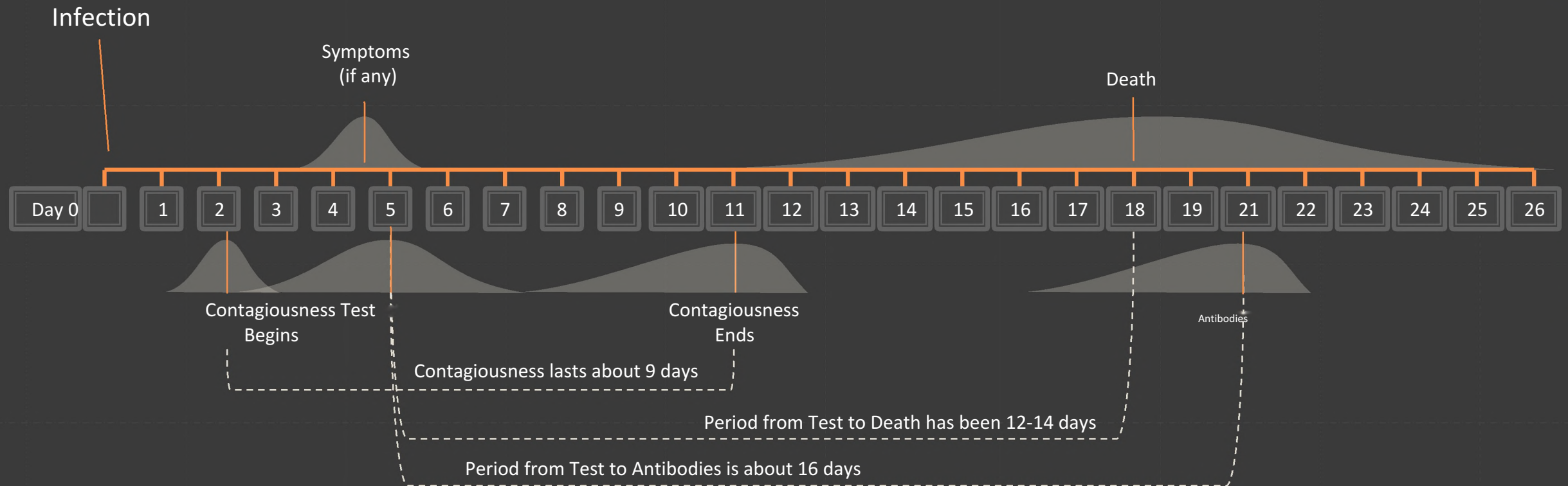
— Of course there's variability in all of this



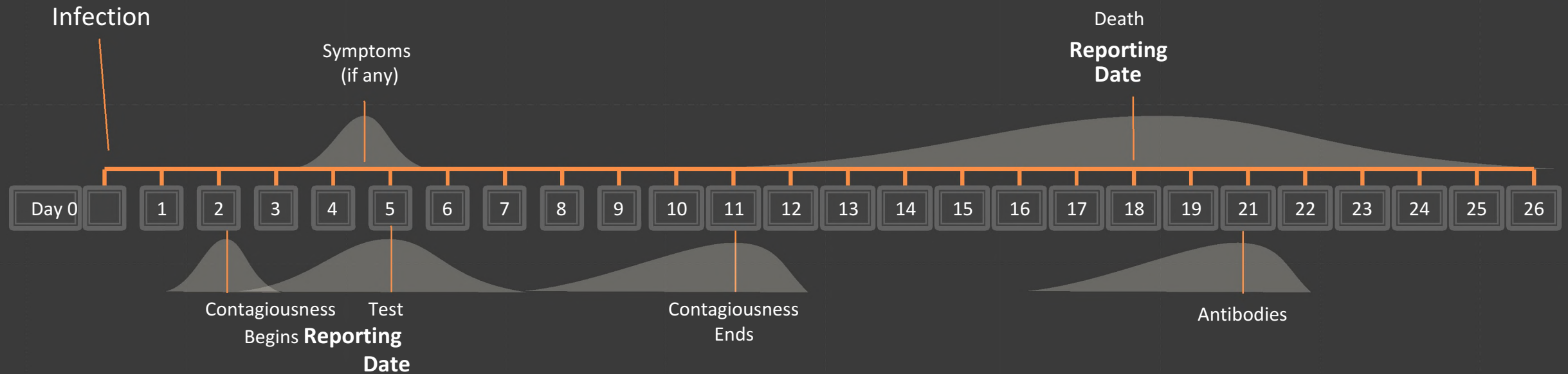
— This allows us to draw some inferences



— This allows us to draw some inferences



— This is more about reporting dates, more than actual dates



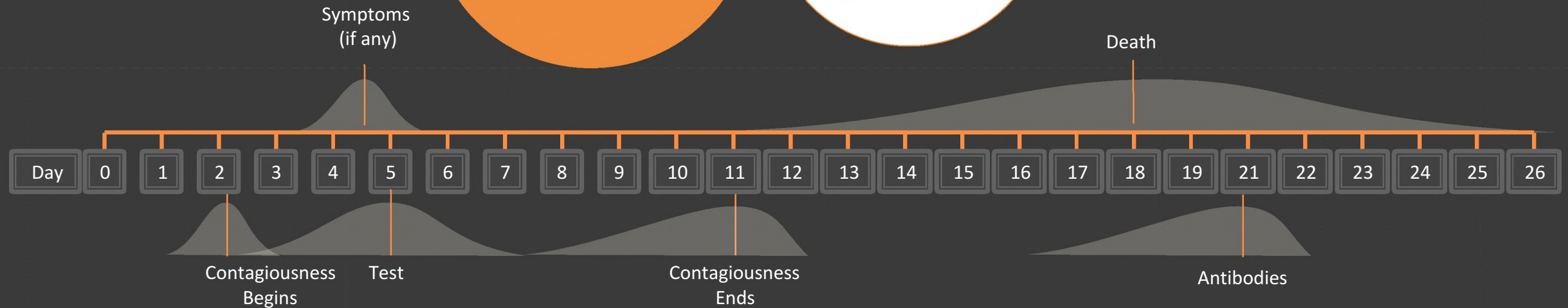


What Does This Have to do With
Forecasting?

Timing forms the basis of forecasting (for my method)

We can do this with
terrific accuracy
at the national level,
and with fair accuracy
at the state level

And because
trends tend to
continue, we can
project forward
another 14 days



We can take
this number, which
we know on date X

We can apply what we've
learned about the ratio of
positive tests to deaths

We can forecast
this number,
14 days ahead

— Forecasting Begins with Data

Current State of Data, Part 1



Current State of Data, Part 2





Problems with Cumulative Data

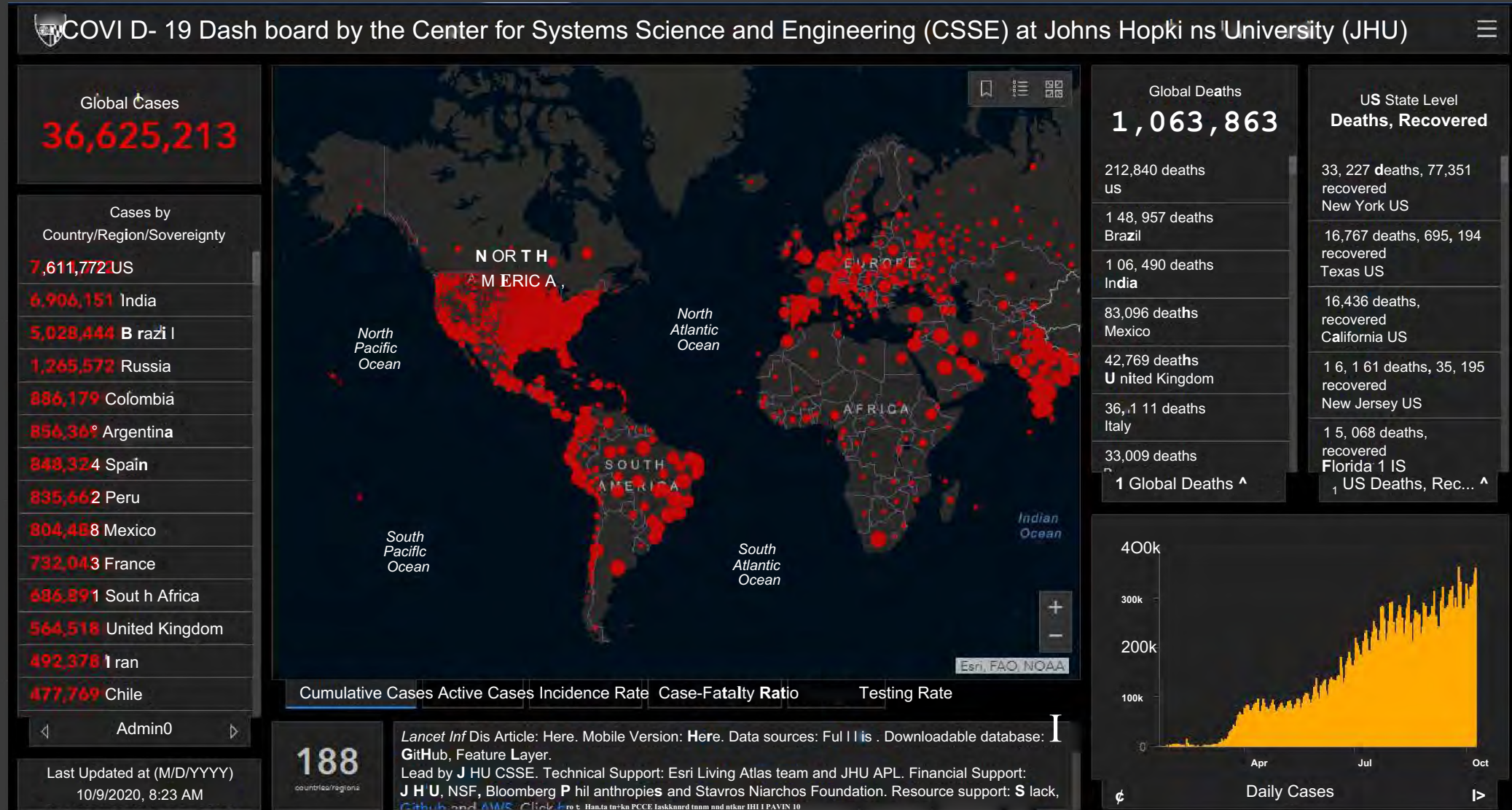
Issues with cumulative data, examples

- Bing/covid graph
- JHU graphic
- What is cumulative data good for?

— TTyypicaall CCuummuullaattivvee GGrraapph

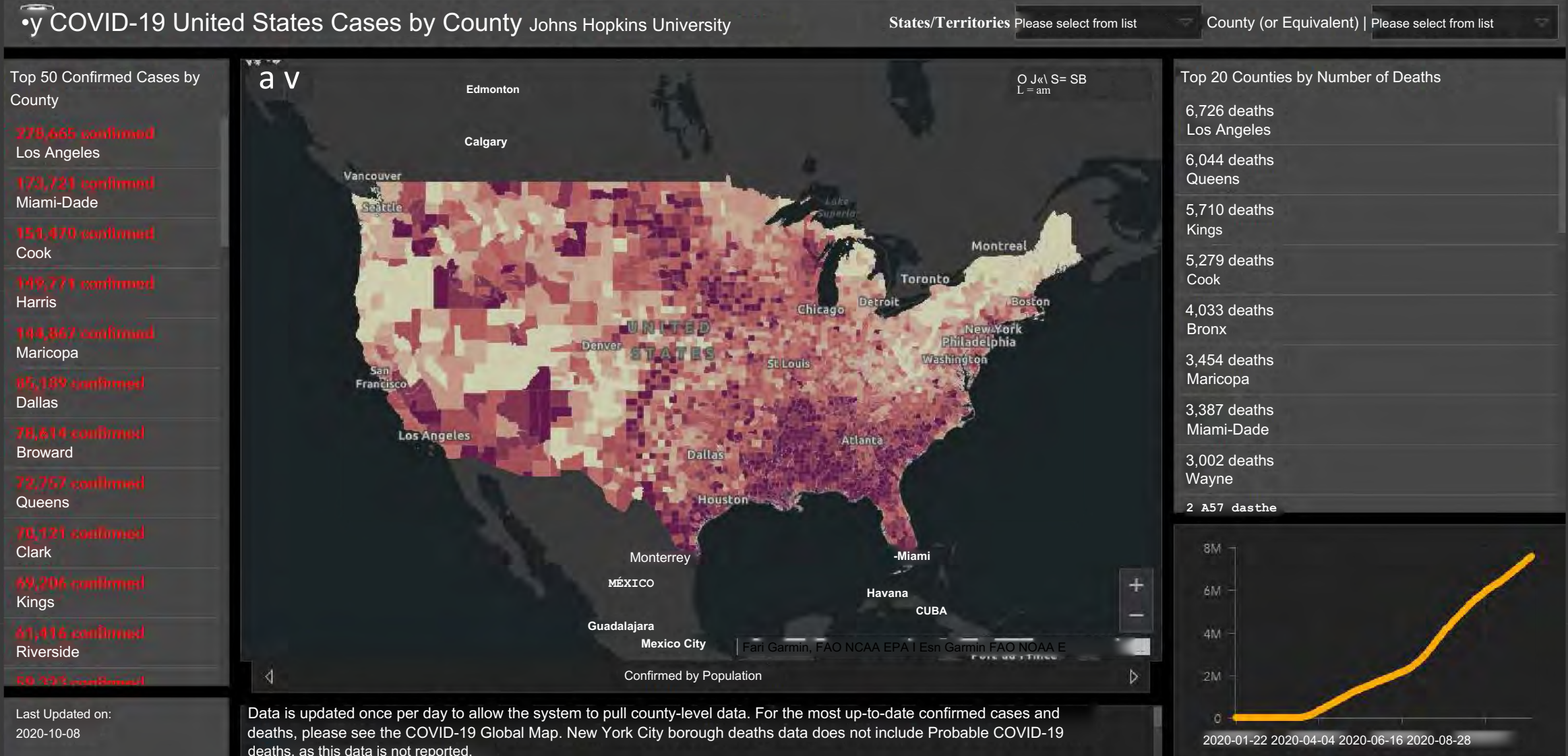


— JJoohhnns HHooppkkiinnss CCoorronnaavviirruuss DDaas



Source: <https://covid19.jhu.edu/map.html>

— Johns Hopkins Coronavirus Dashboard - Newer Graphic



Source: <https://coronavirus.jhu.edu/us-map>

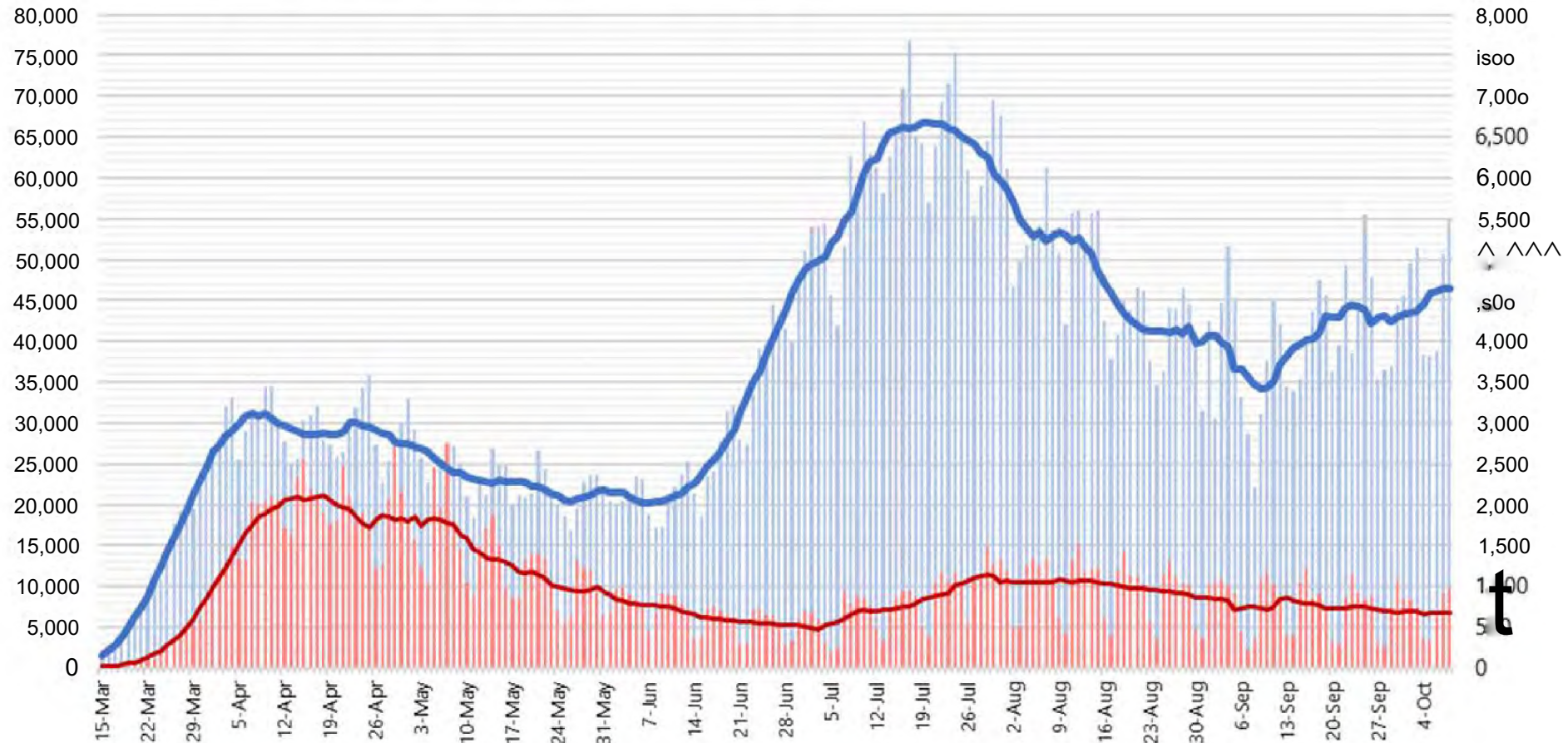


Incremental Data

- Incremental data is more meaningful than cumulative data
- But it also opens the door to a whole new set of data quality issues

— Daily Graphic

US Daily Positive Tests and Deaths as of 10/8/20

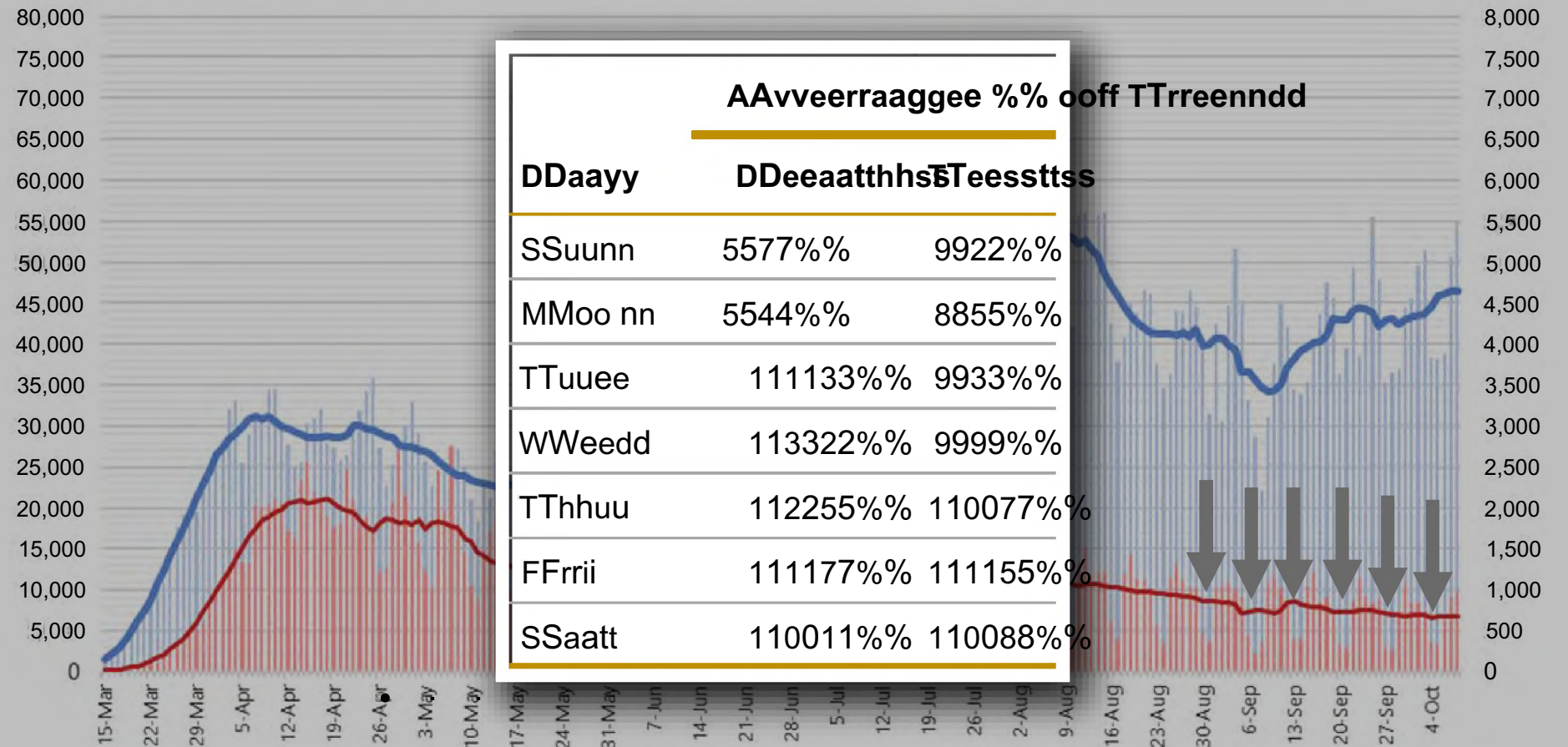


Source: stevemccconnell.com/covid

Positive Tests Deaths 7- Day Average Positive Tests 7- Day Average Deaths

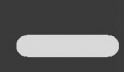
— IIssssuuee ##11 wwiitthh DDaaiilly DDaatta -- SSuunnda

US Daily Positive Tests and Deaths as of 10/8/20



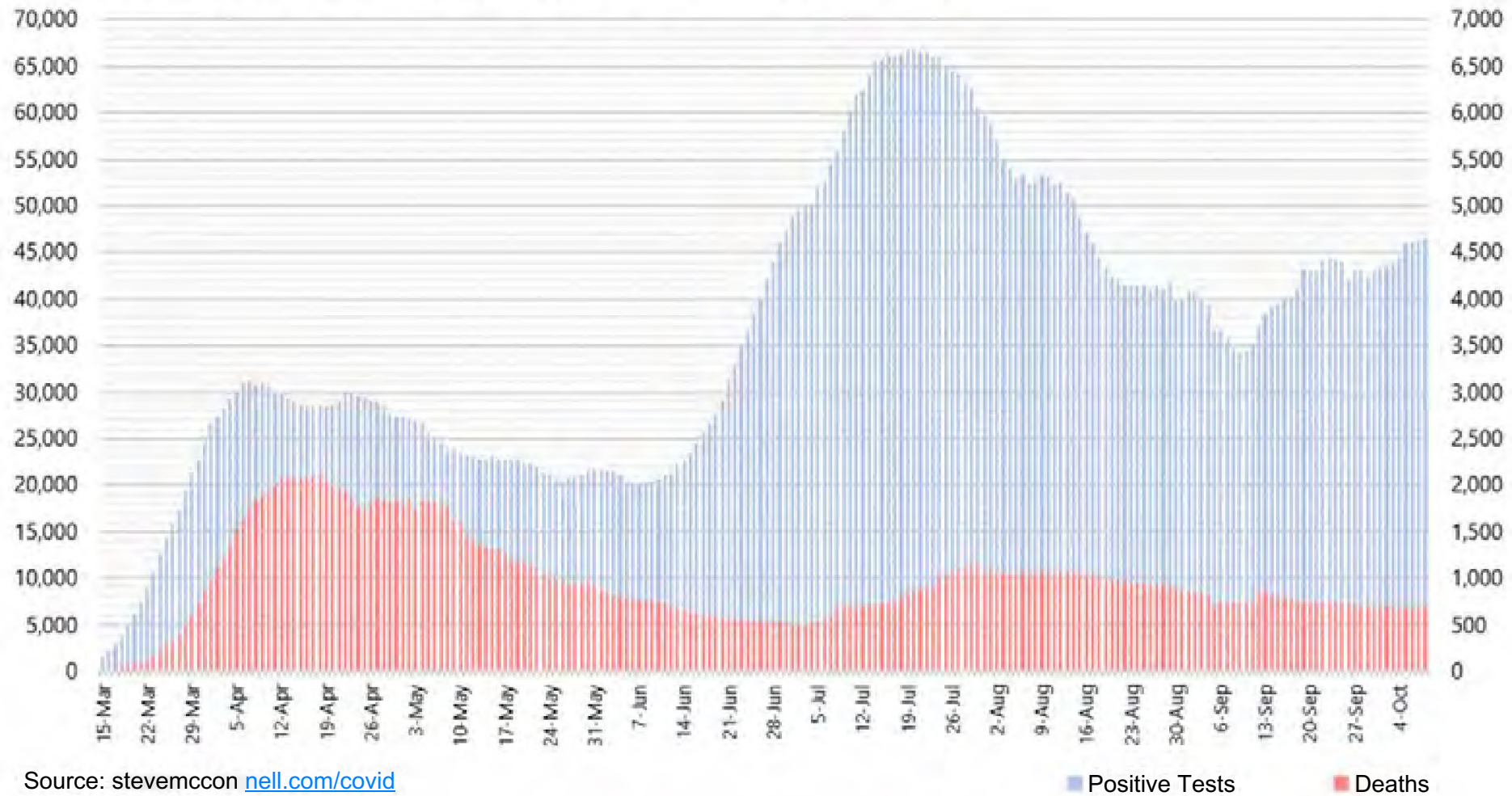
Source: stevemccconnell.com/covid

Positive Tests Deaths 7-Day Average Positive Tests 7-Day Average Deaths

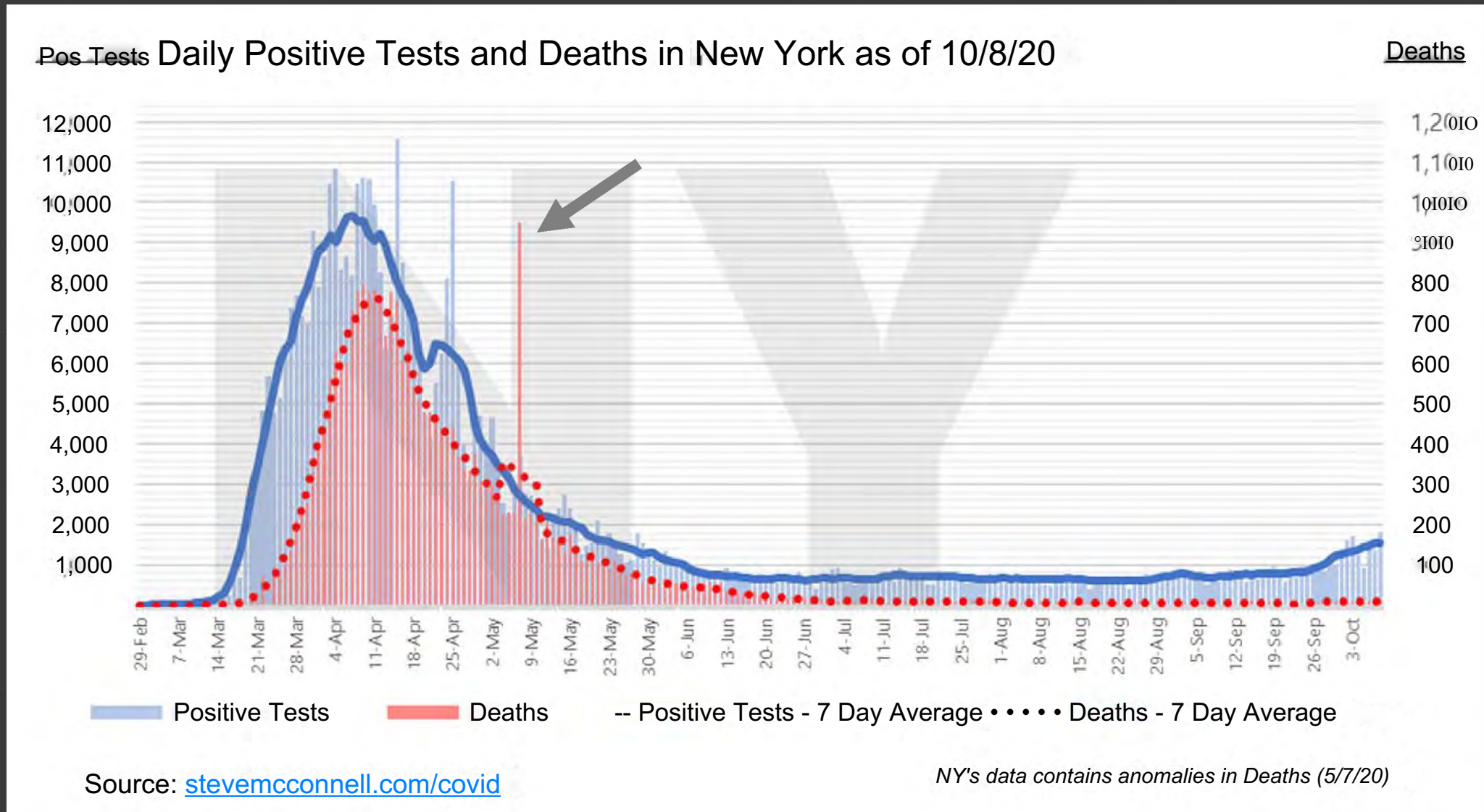


Issue #11 with Daily Data – Sunday

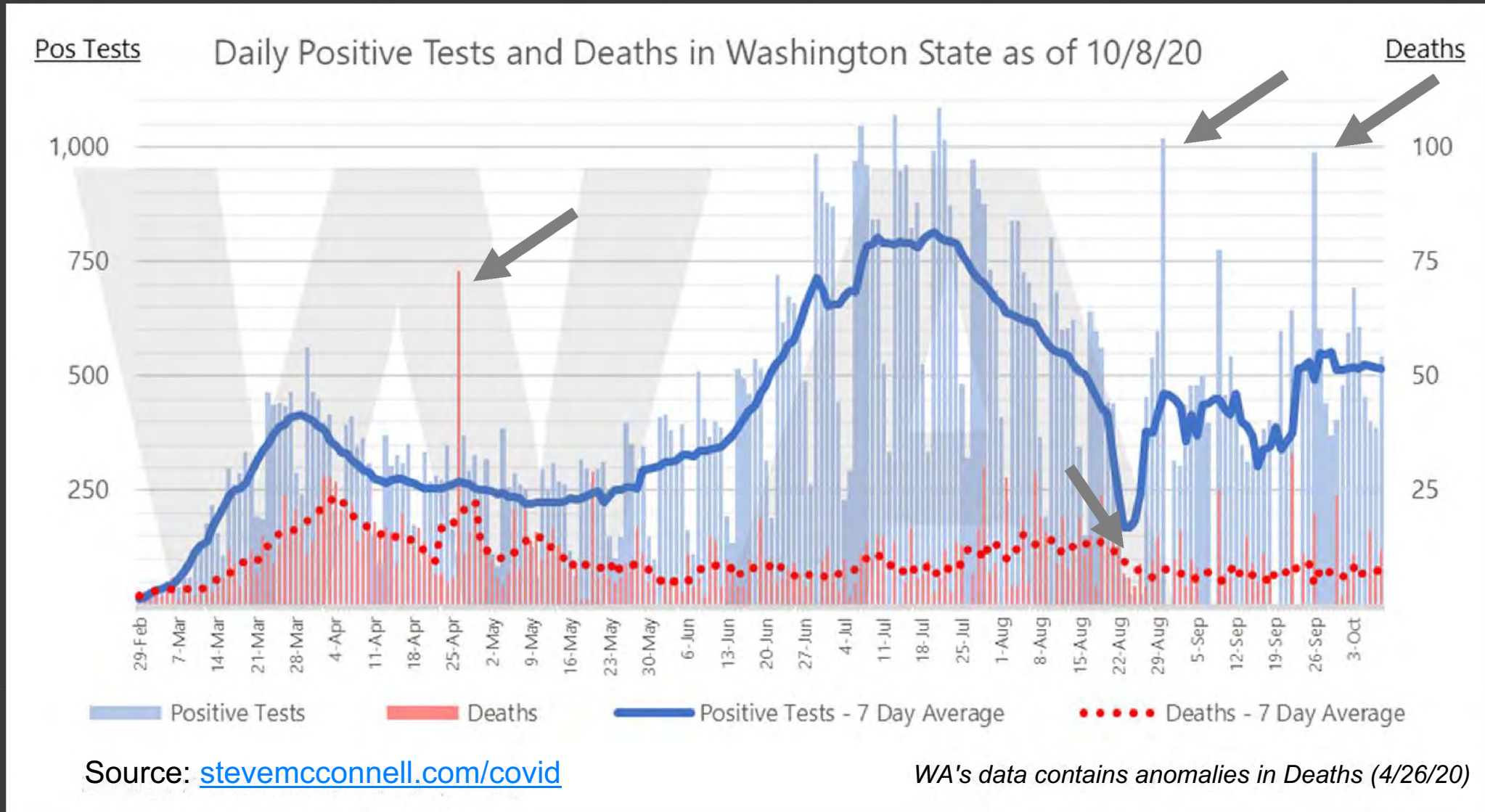
US Smoothed Daily Positive Tests and Deaths as of 10/8/20



— Isssuee ##22 wwiitthh DDaaiillyy DDaatta -- SSppiikkeess



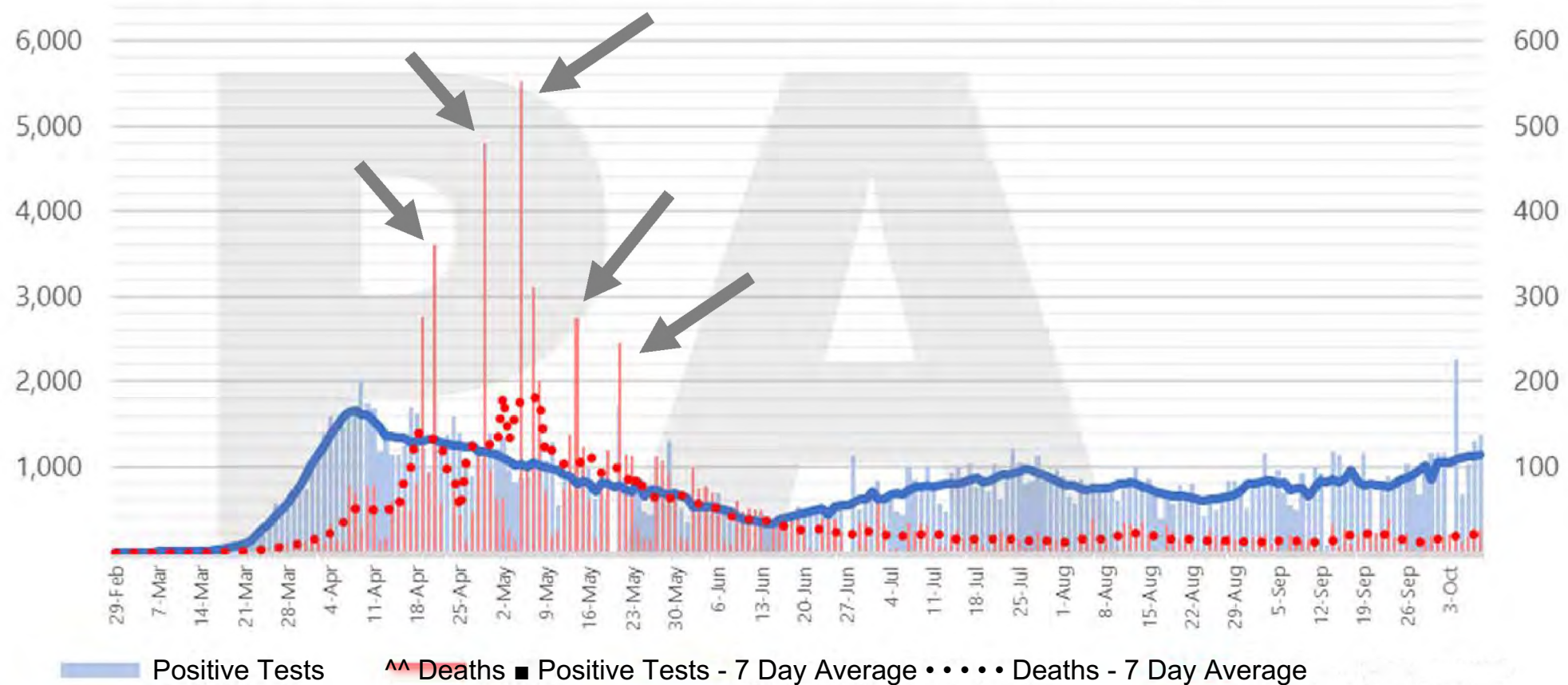
— IIssssuuee ##22 wwiitthh DDaaiillyy DDaatta -- SSttaatte L



— IIssssuuee ##22 wwiitthh DDaaiillyy DDaatta -- SSttaatte L

Pos Tests Daily Positive Tests and Deaths in Pennsylvania as of 10/8/20

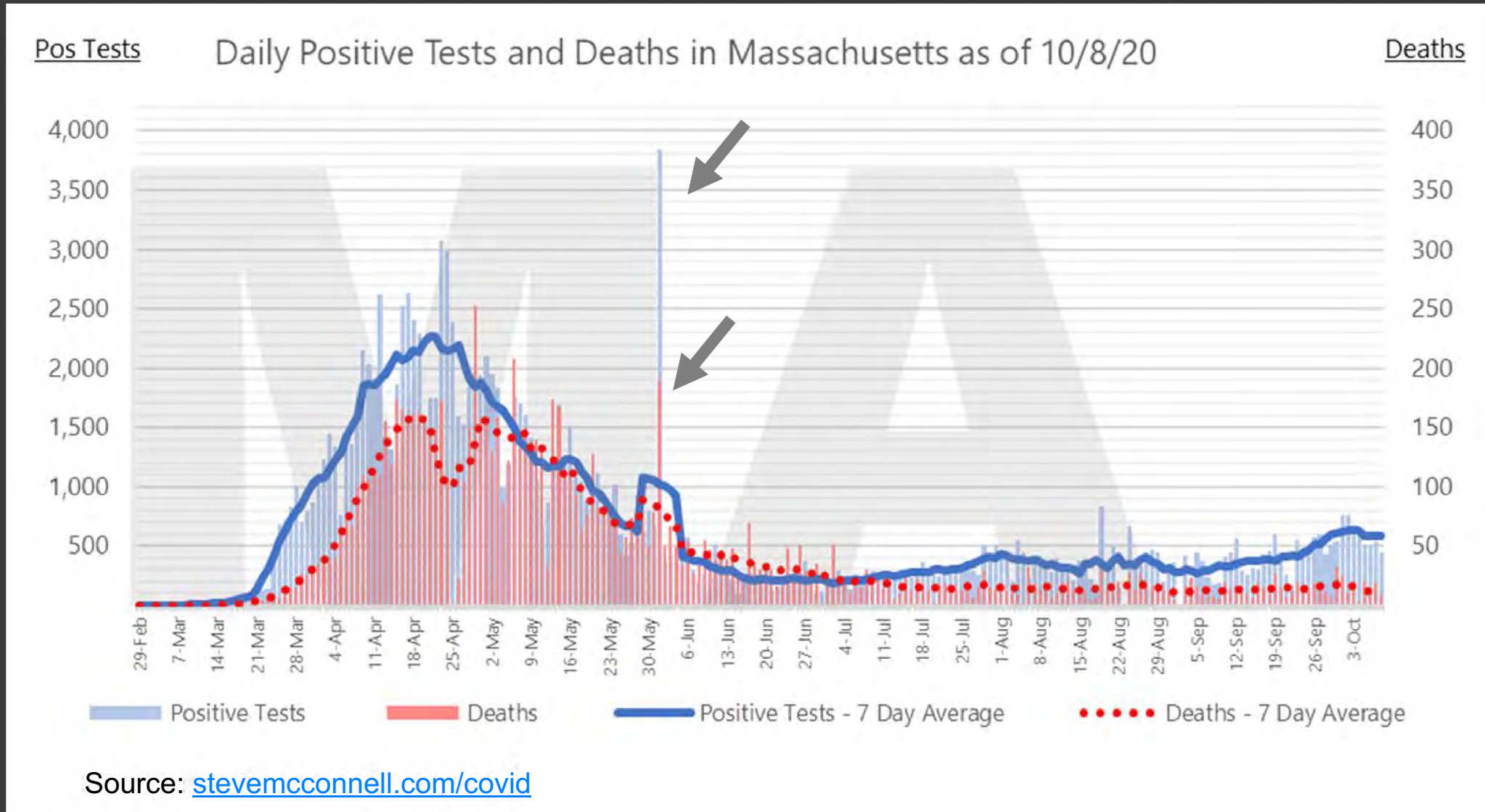
Deaths



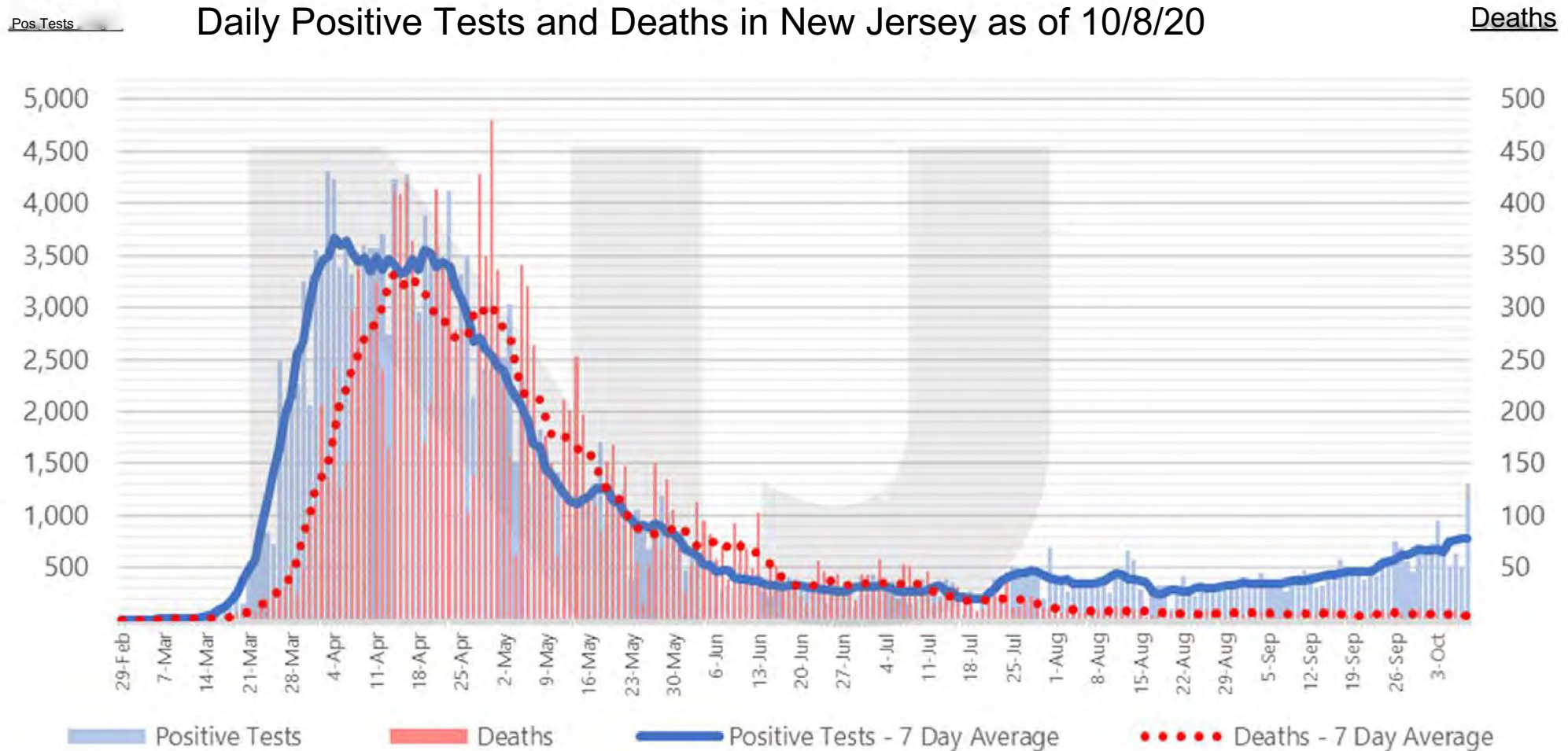
Source: stevenmcconnell.com/covid

PA's data contains anomalies in Deaths (5/5/20)

— Issue #22 with Daily Data — State L



— IIsssuuee ##22 wwiitthh DDaaiillyy DDaatta -- SSttaatte L



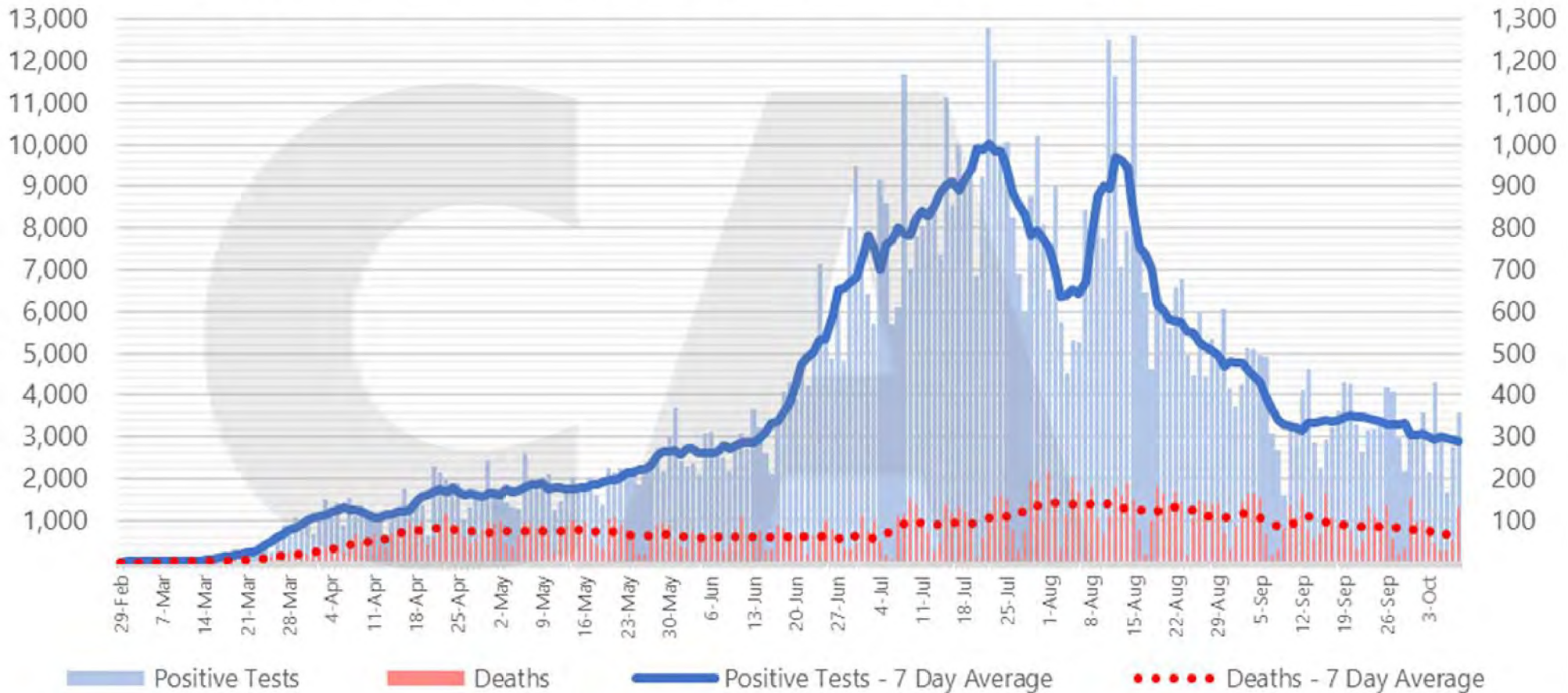
Source: stevemcconnell.com/covid

NJ's data contains anomalies in Negative Tests (8/8/20)

— IIssssuuee ##22 wwiitthh DDaaiilly DDaatta -- SSttaatte L

Pos Tests Daily Positive Tests and Deaths in California as of 10/8/20

Deaths



Source: stevemcconnell.com/covid

CA 's data contains anomalies in Deaths (8/1/20)

— State Level Data Smoothing

- Due to non-regular reporting frequencies, smoothed data is actually more accurate than non-smoothed data
- The smoothing period must be a multiple of 7 days

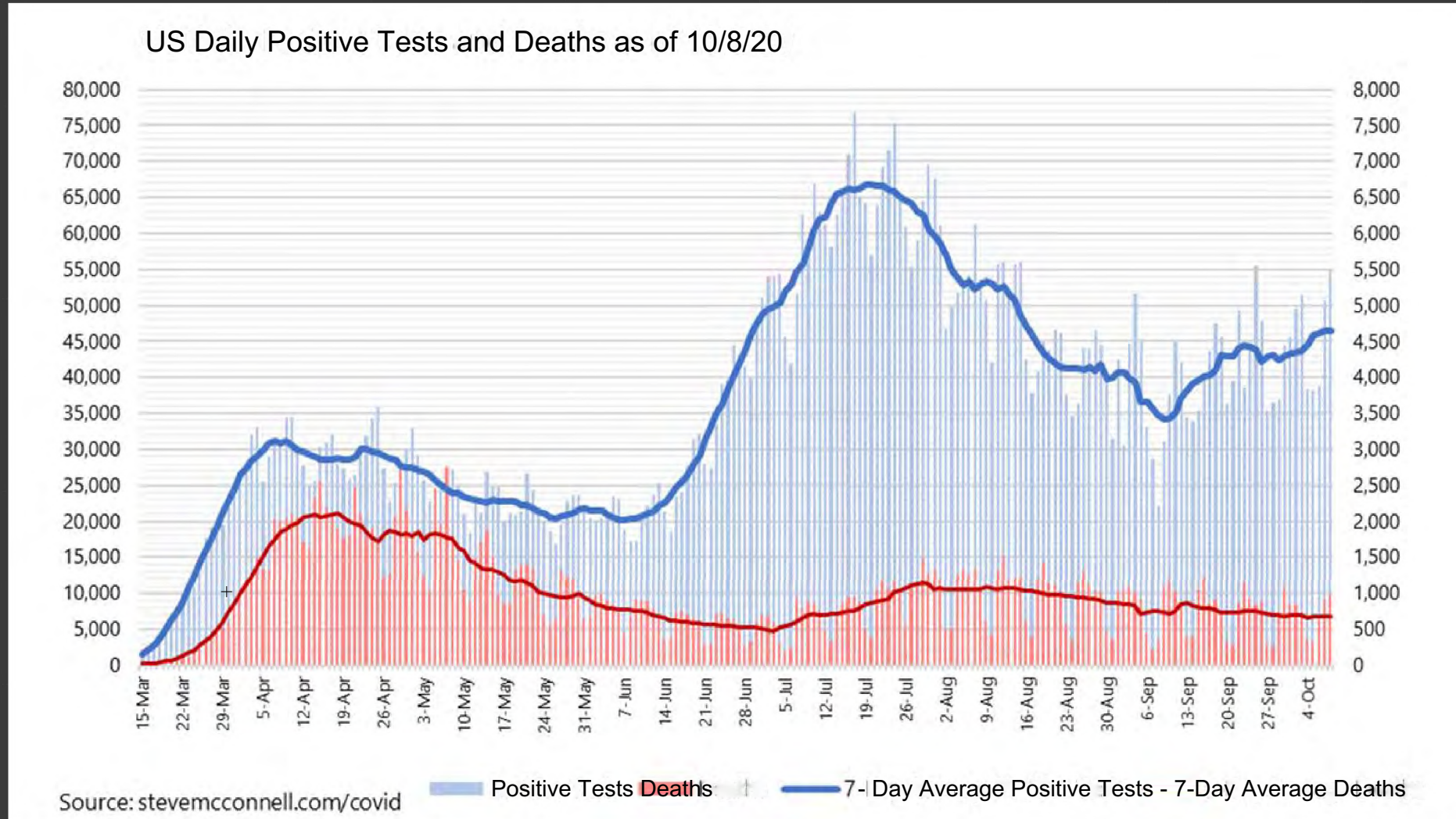
By the same reasoning, weekly data is more accurate than daily data

- Smoothing helps, but it doesn't change the fact that there are still weird spikes in the data

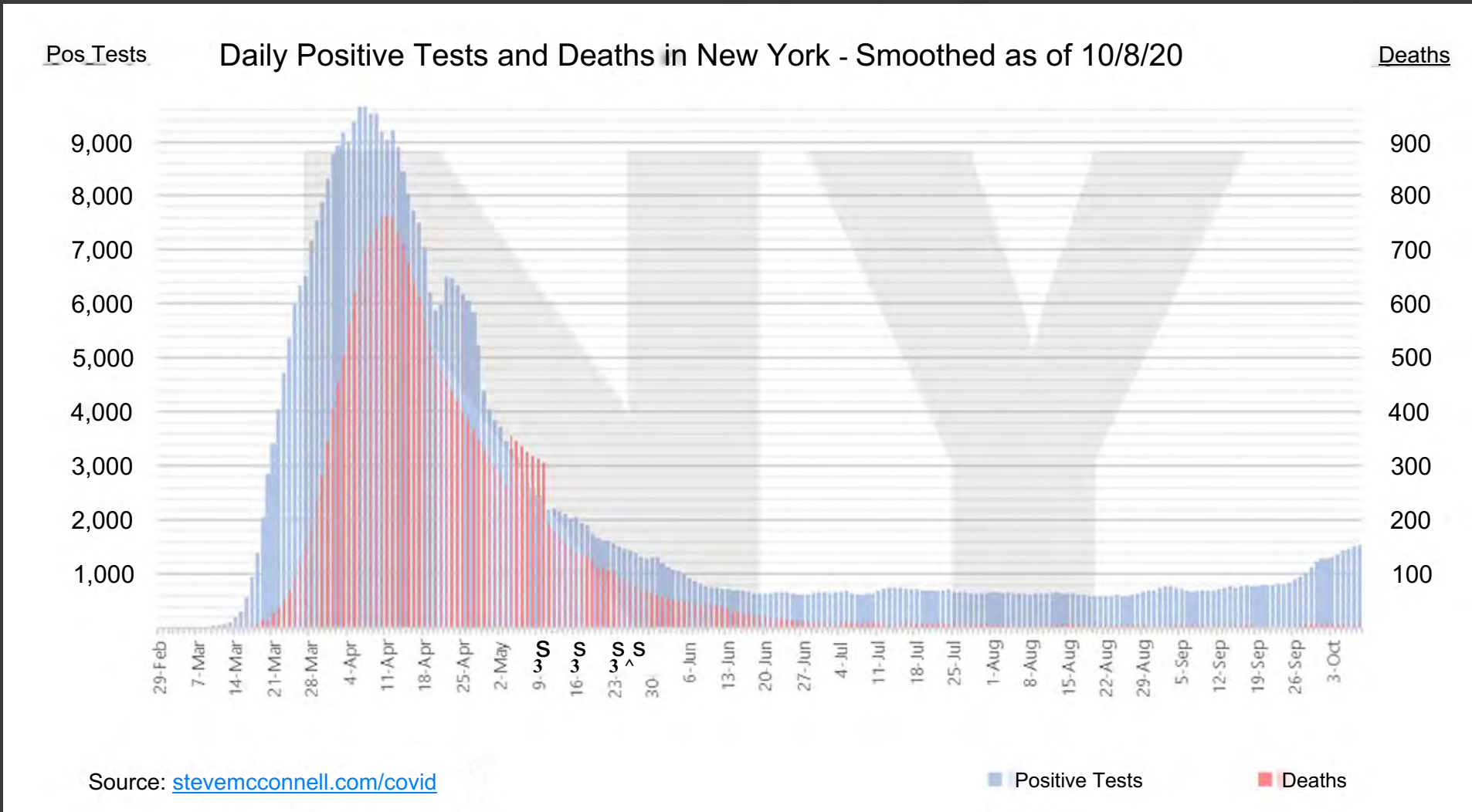
Covid-19 Data Issues

- Issues with daily data
 - Part 1: Sundays (the weekly cycle)
 - Part 2: State-level data spikes and corrections
- Alternatives
 - Running averages
 - Deltas

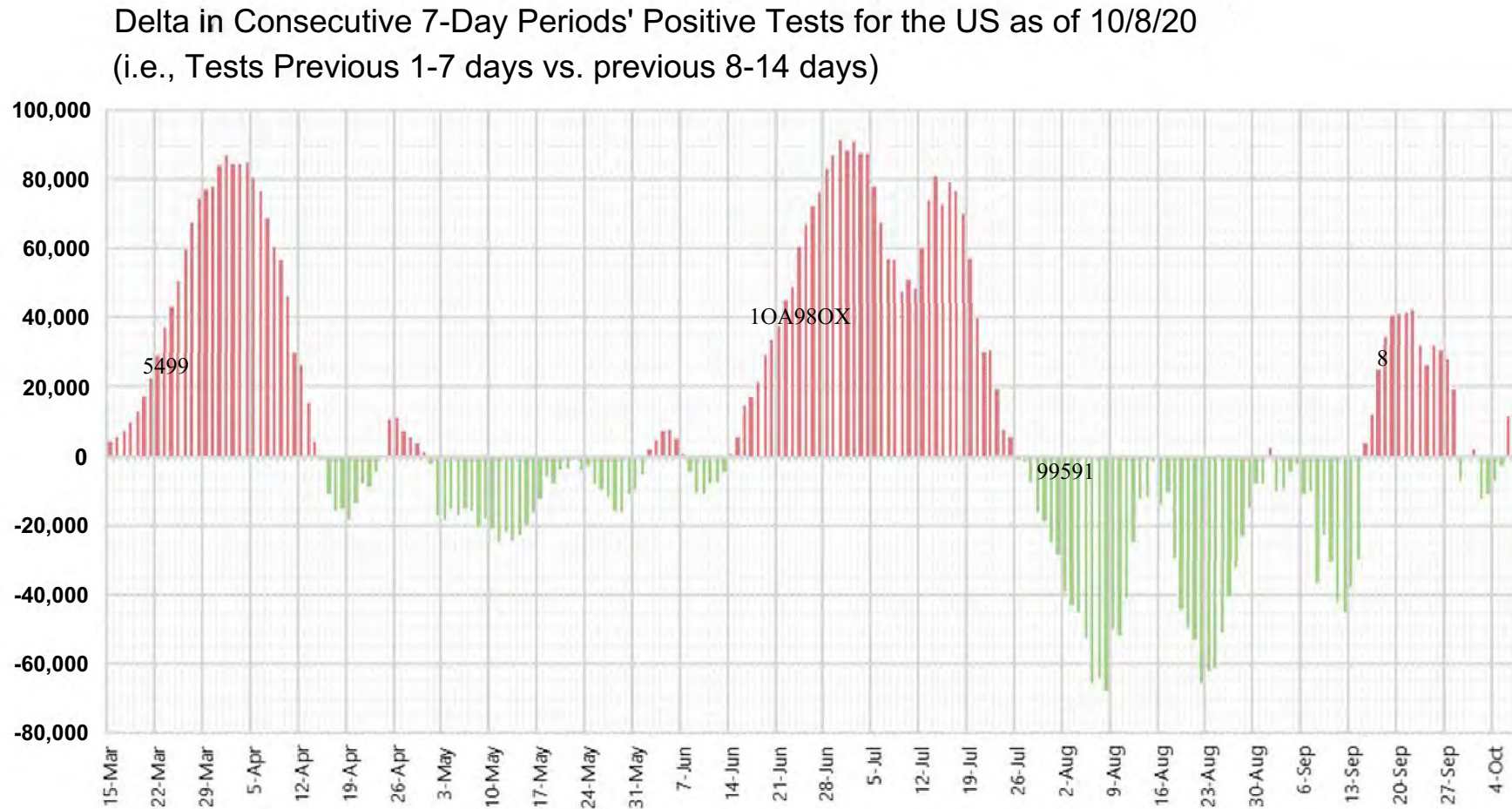
— DDaattaa SSmooootthhiinnngg — RRuunnnniinnngg AAvveer



— DDaattaa SSmooootthhiinnngg — RRuunnnniinnngg AAveer



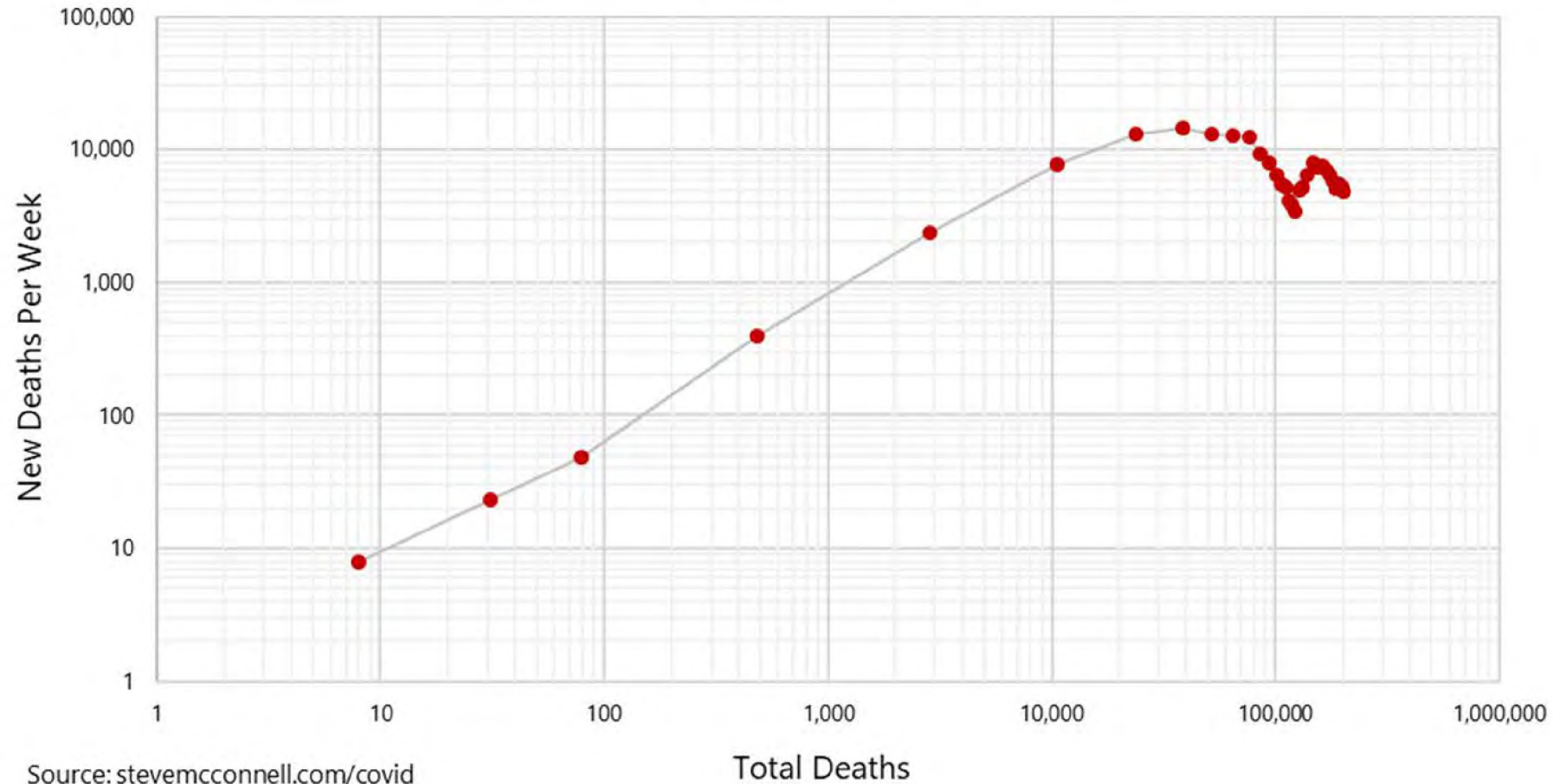
— DDaattaa SSmooootthhiinnngg — DDeelttaass ((Ilinneea



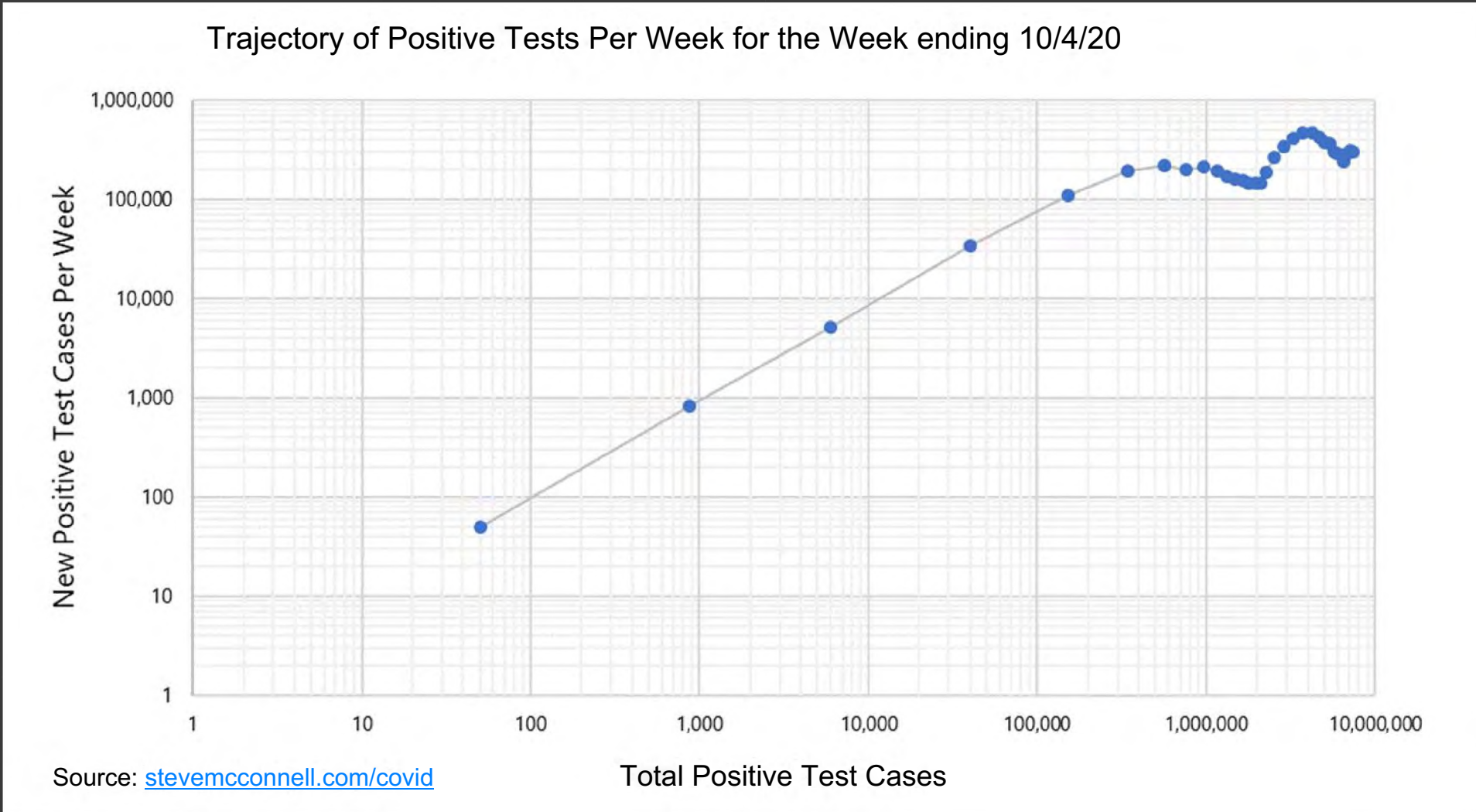
Source: stevemcconnell.com/covid

— DDaattaa SSmooootthhiinnngg -- DDeelttaass ((lloogg))

Trajectory of Deaths Per Week for the Week ending 10/4/20



— DDaattaa SSmooootthhiinnngg -- DDeelttaass ((lloogg))



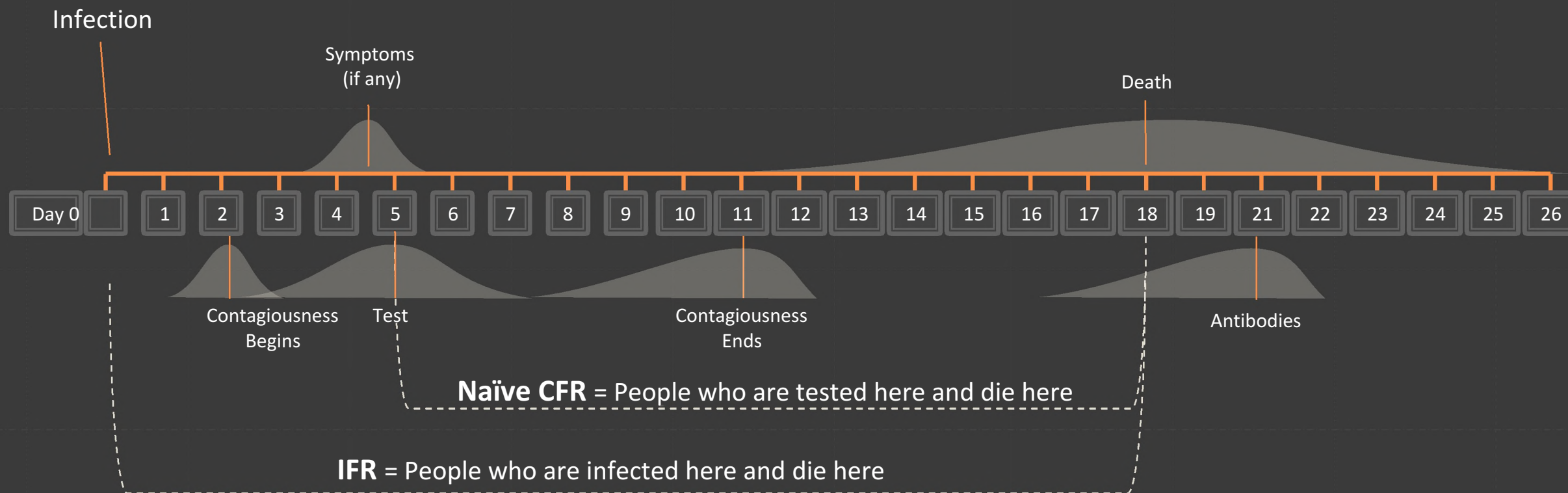
- Specific Data Relationships

Calculating CFR/IFR

CFR vs. IFR - Terminology

- “Case” – people who are diagnosed/tested as positive
- “Infected” – people who get the virus whether they’re tested or not
- “CFR” = Case Fatality Rate
 - This is number of fatalities divided by number of “cases” (positive tests)
- “Junk CFR” = Today’s fatalities divided by today’s positive tests
- “Naïve CFR” = Today’s fatalities divided by the number of positive tests at some sensible point in the past (e.g., 14 days)
- “IFR” = Infection Fatality Rate – percentage of people infected who die, whether they’ve been tested or not
- BOTH Naïve CFR and IFR are useful for certain purposes

— IFR vs. CFR



Usefulness of IFR

Useful for assessing risk

To an overall population

To specific groups within the population

Usefulness of CFR

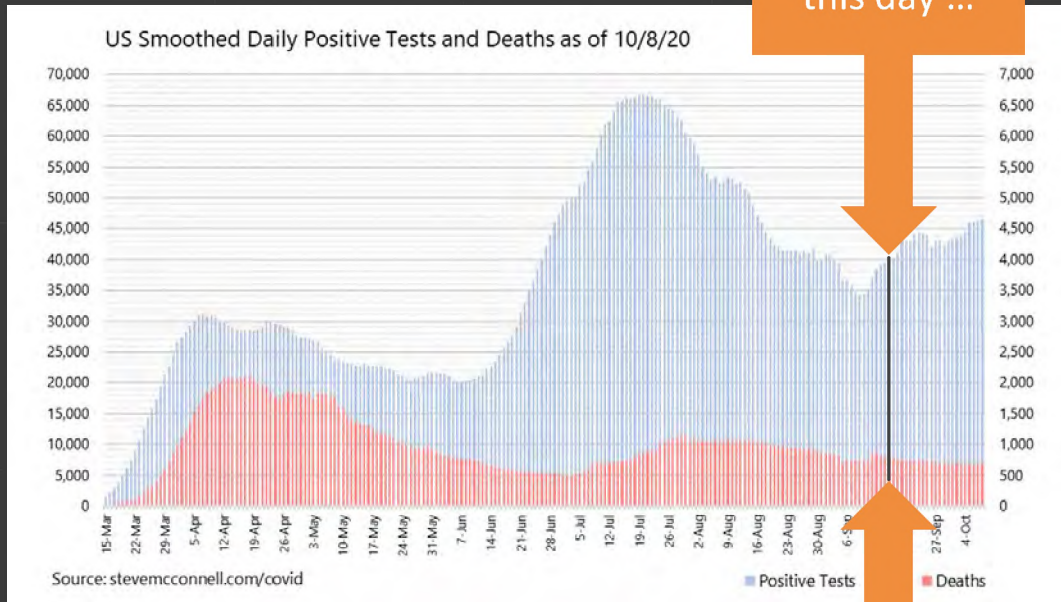
Useful for estimation purposes

- IF you get all your facts straight (data assumptions)
- Calculations are usually presented incorrectly in ways that undermine clear understanding of the data

Data
Correction #1

Lining up time periods correctly
for Naïve CFR

The people
infected on
this day ...

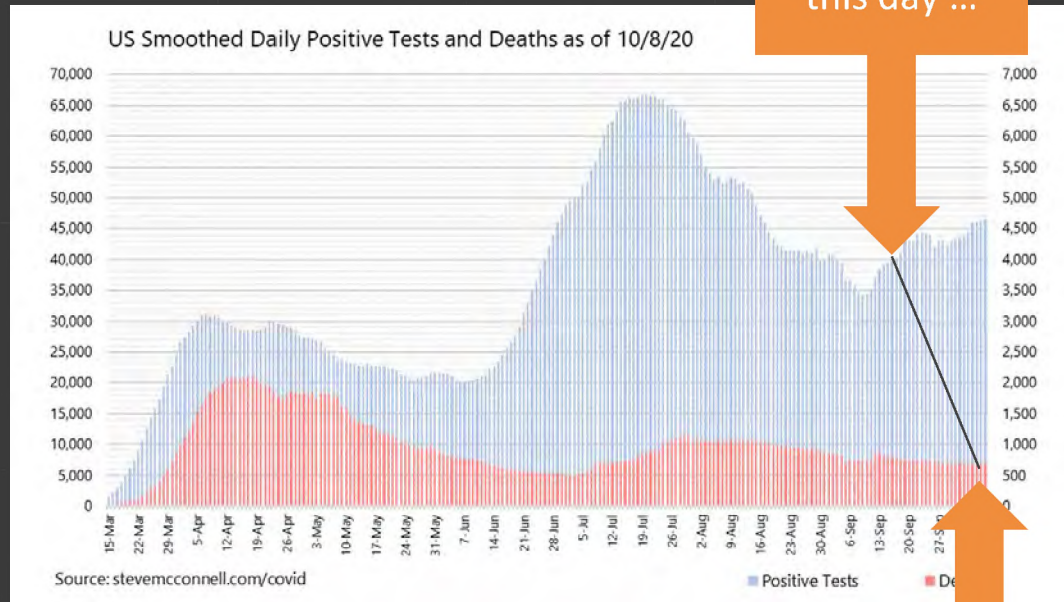


... do not die
on the same
day

Correcting “Junk CFR” (line up time periods correctly)

- People do not get infected, test positive, and die on the same day
- But commonly reported CFRs effectively assume that’s the way it happens

The people infected on this day ...



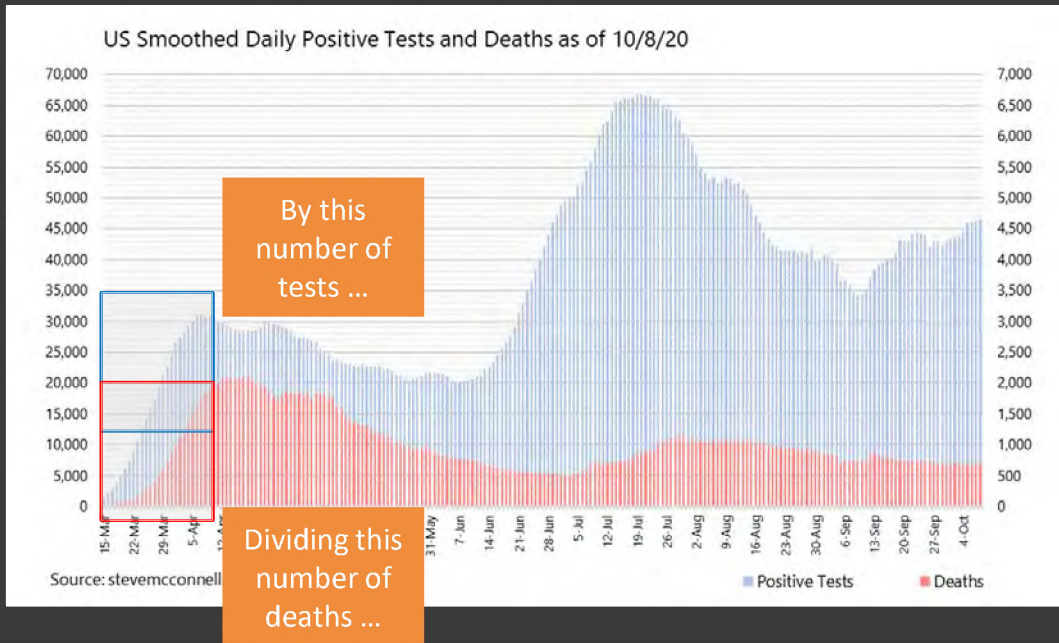
... die, on average, about 14 days later

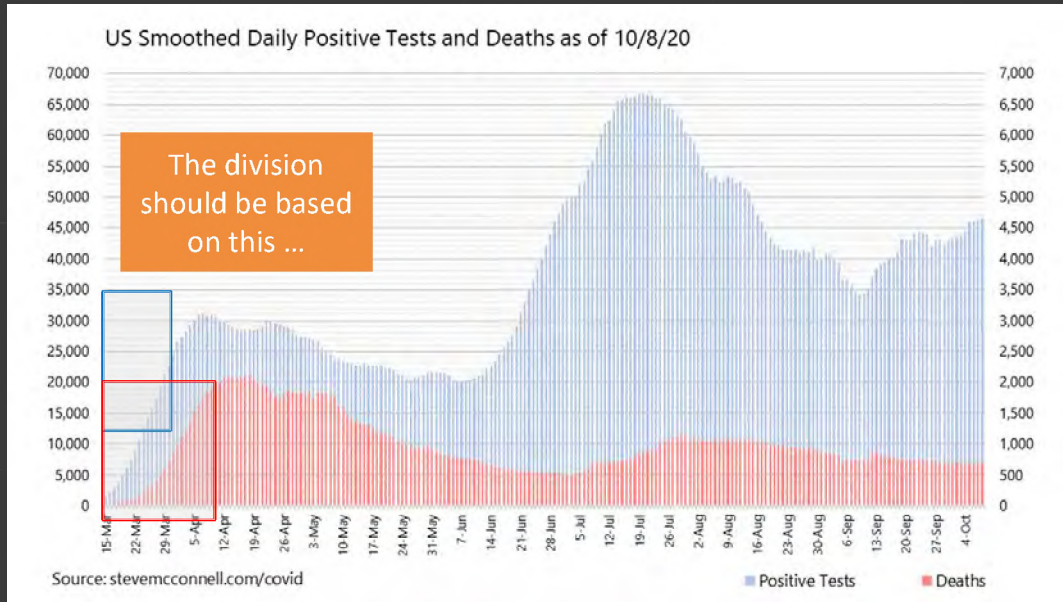
Correcting “Junk CFR” (line up time periods correctly)

- People do not get infected, test positive, and die on the same day
- But commonly reported CFRs effectively assume that’s the way it happens

Correcting “Junk CFR” (line up time periods correctly)

- This mistake gives rise to dramatically inaccurate calculations of fatality rates, especially in the early days of the pandemic





Correcting “Junk CFR” (line up time periods correctly)

- This mistake gives rise to dramatically inaccurate calculations of fatality rates, especially in the early days of the pandemic
- This correction results in a **higher** CFR than was reported by the media at the time (but that doesn't imply the IFR was higher)

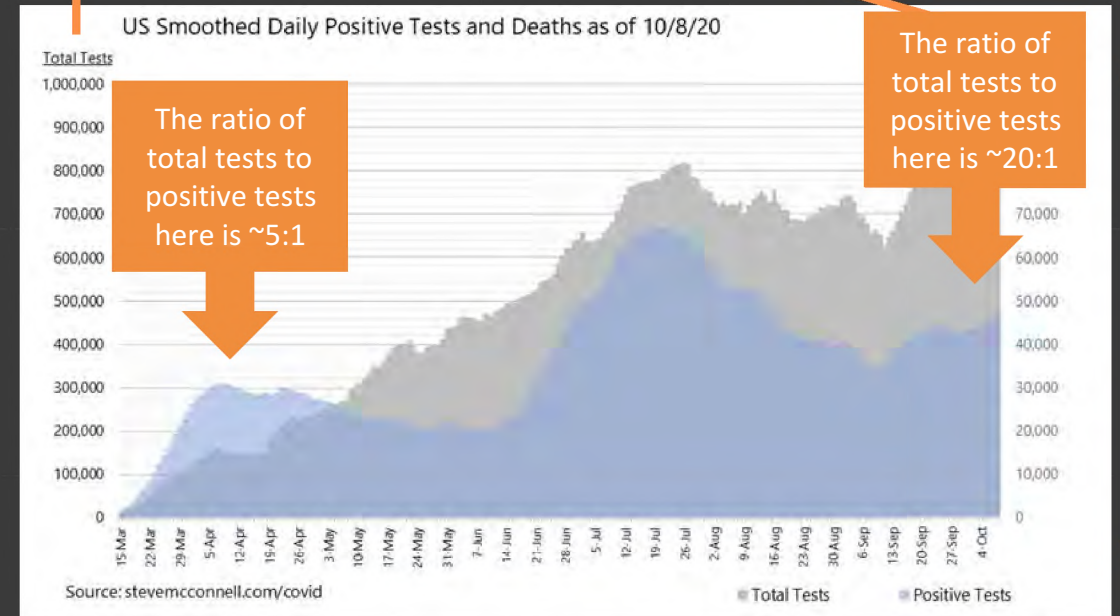
Data
Correction #2

Tests <> Cases

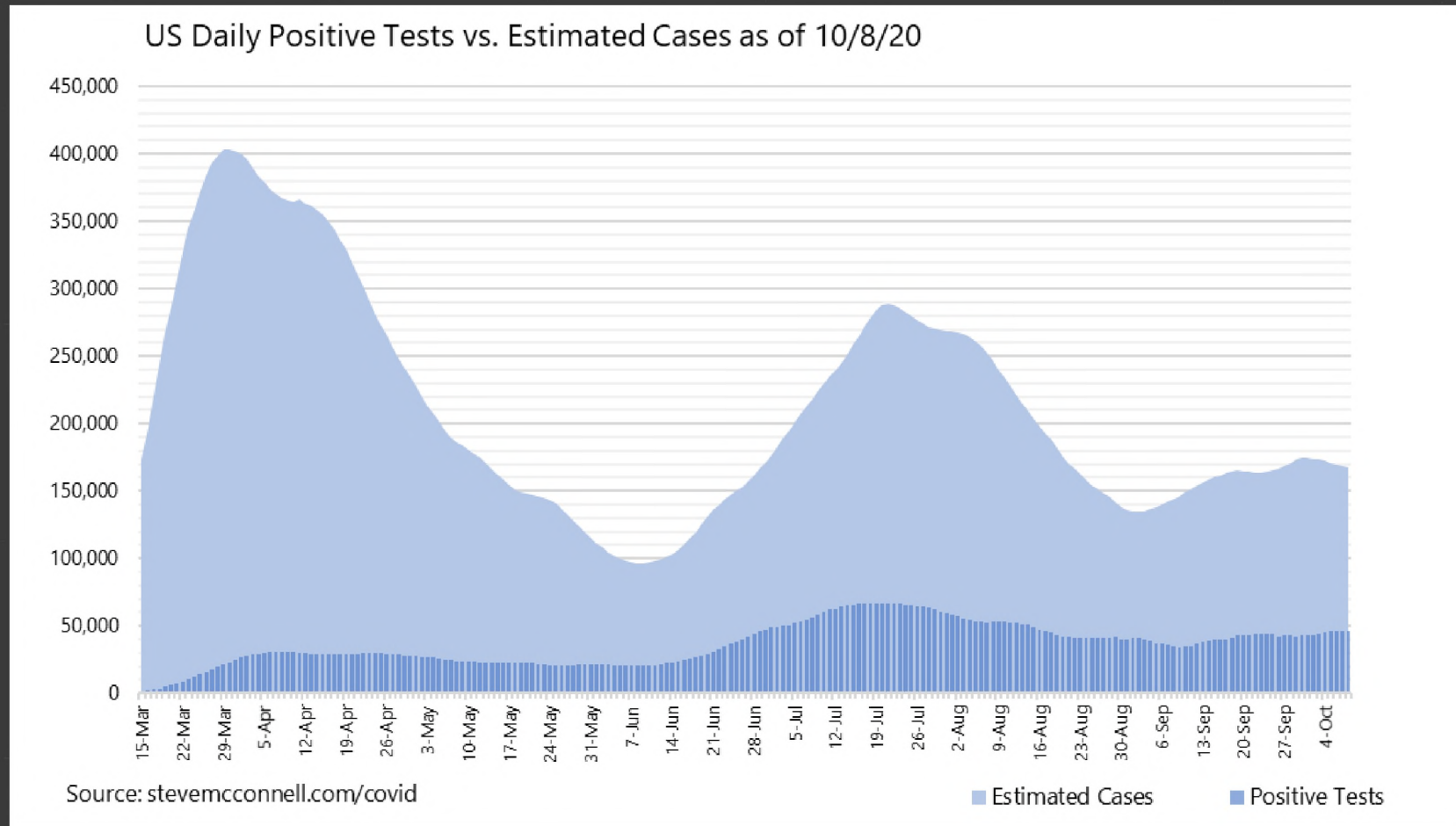
Relationship between positive tests and cases has changed

- The level of testing has steadily increased over the course of the pandemic
- Early in the pandemic, there were **10-20** cases per positive test
- More recently, there are **3-5** cases per positive test

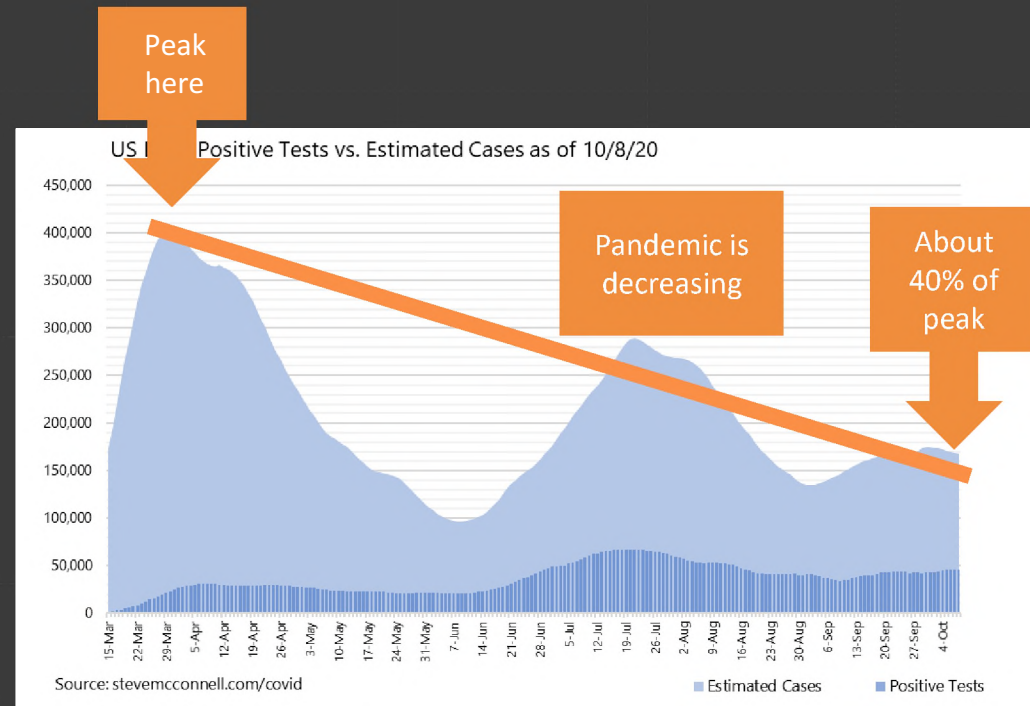
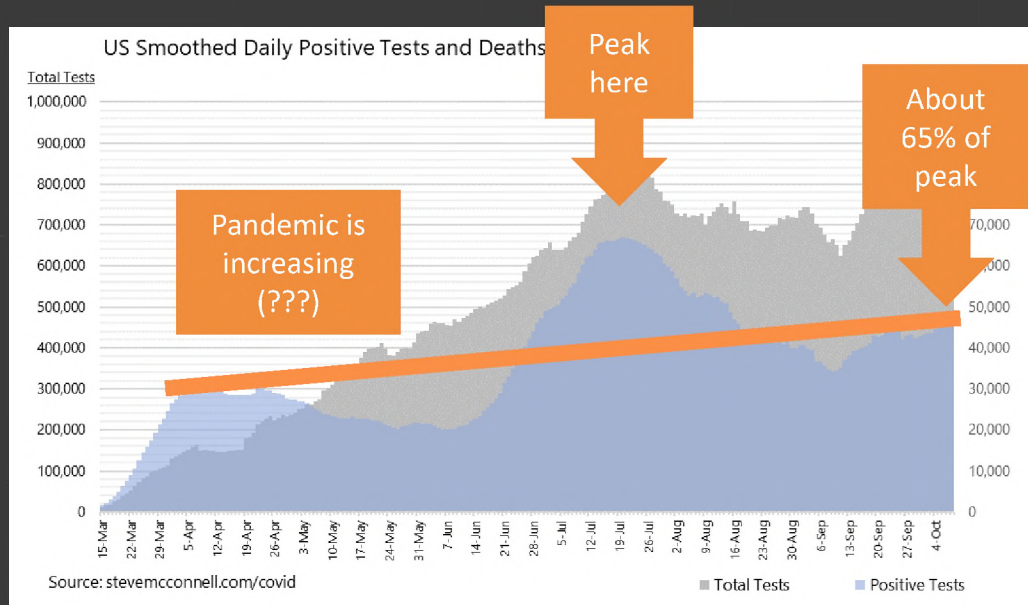
Note: the total tests axis is 10x the positive tests axis



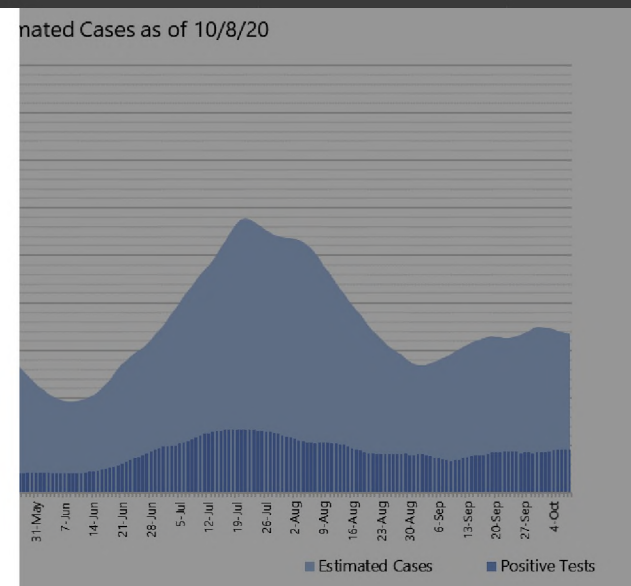
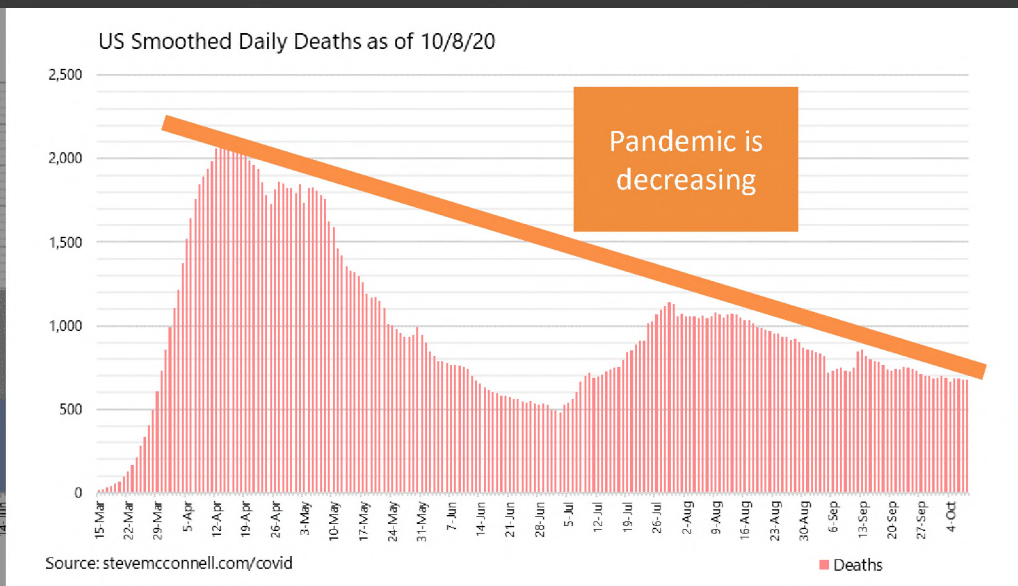
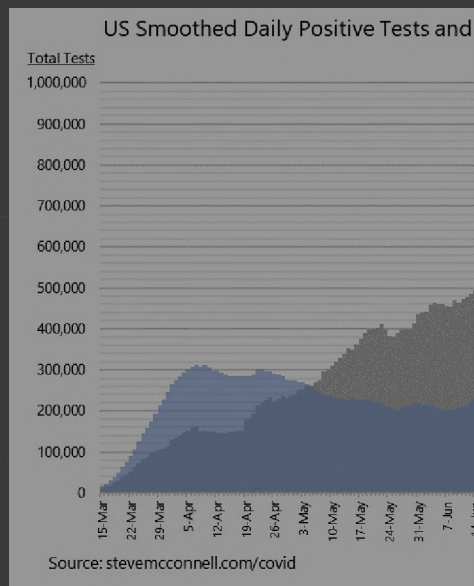
— PPositive Tests vs.. Estimated Cases



— More on Positive Tests vs. Cases



— The Estimated Cases Trend Aligns with Death Trend (The Positive Tests Trend Does not)



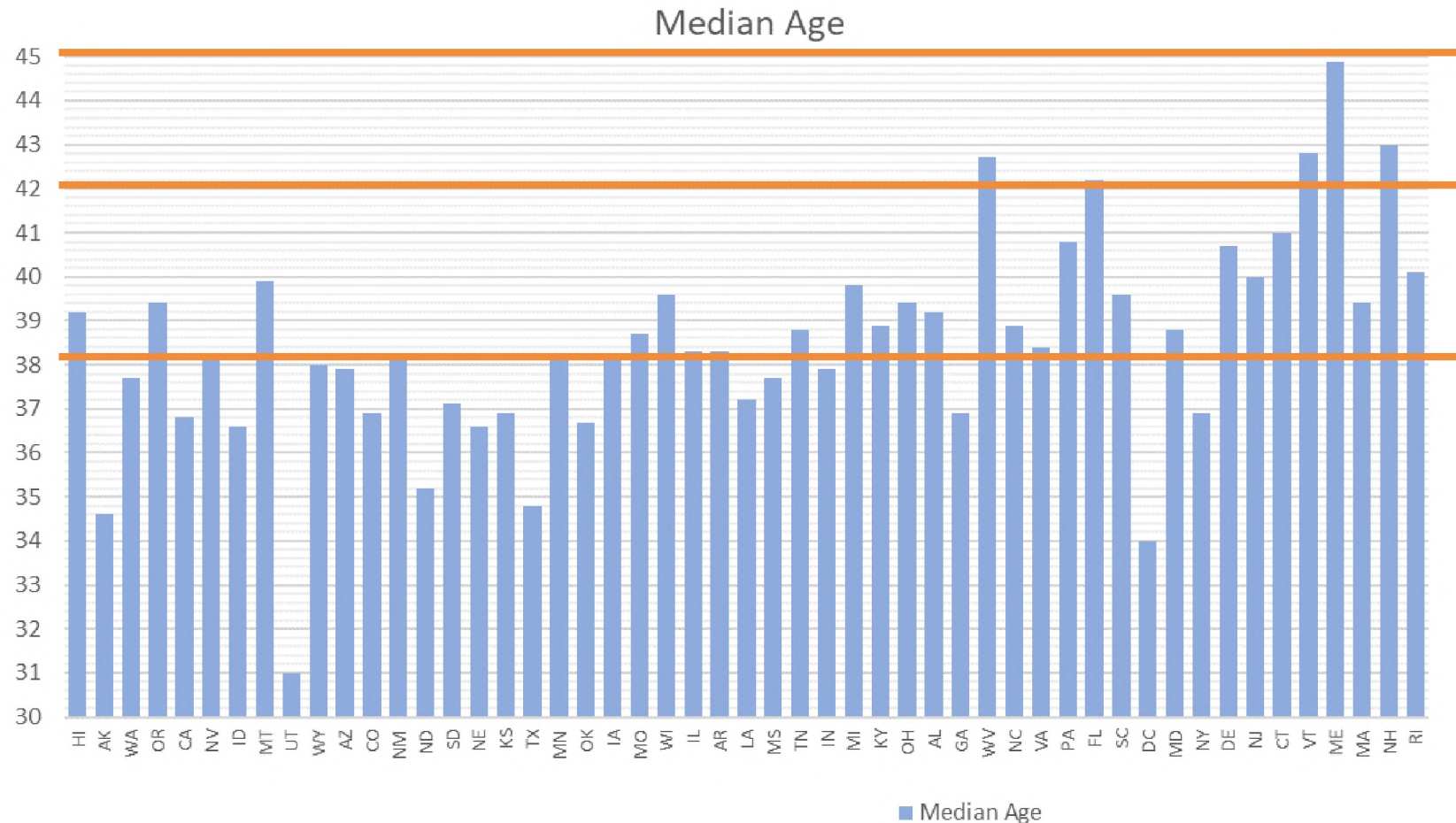
— Calculating Meaningful CFR

- ❑ Get our time periods for tests and deaths lined up correctly
- ❑ Get clear about the idea that tests $\neq$ infections
- ❑ Calculate a correct number for infections

Data
Correction #3

Account for Age

— Remaining CFR Issue: CFR is Age-Based



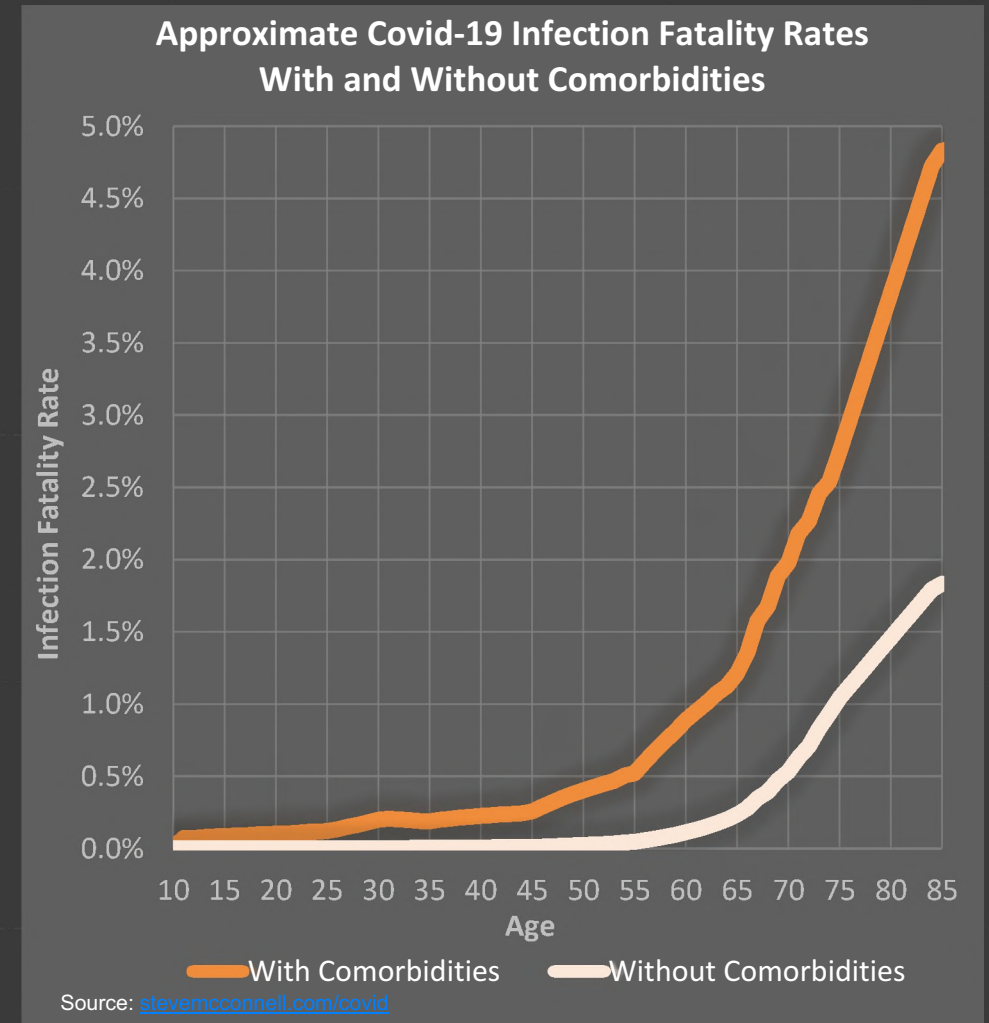
Italy's average is 45.1

Spain's average is 42.3

Average age in the US
is 38.2

IFRs and Co-Morbidity

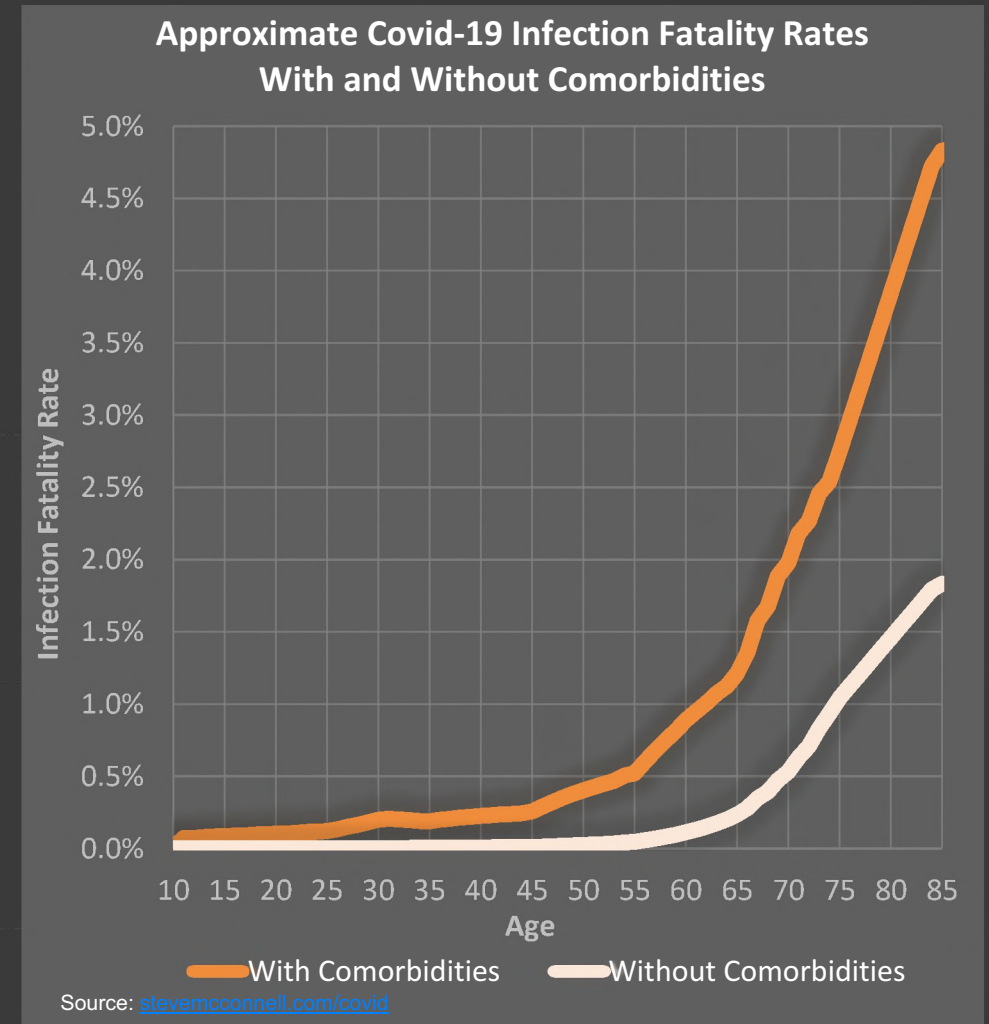
- 90% of deaths have involved at least 1 co-morbidity
- Risk of death with a co-morbidity is HIGHER than published; risk without is LOWER
- At younger ages this makes a HUGE difference in risk



Most serious comorbidities, according to CDC

- Serious heart conditions, such as heart failure, coronary artery disease, or cardiomyopathies
- Cancer
- Chronic kidney disease
- COPD
- Obesity (BMI > 30)
- Sickle cell disease
- Solid organ transplantation
- Type 2 diabetes mellitus

CDC list of risk levels of specific comorbidities and level of evidence for each:
<https://www.cdc.gov/coronavirus/2019-ncov/need-extra-precautions/evidence-table.html>



Summary of “Tools” we can use for Forecasting

- Corrections to raw data reporting (weekly cycle issues)
- Corrections to timeline (lag from positive tests to deaths)
- Understanding of what is needed to estimate infections (not tests)
- Potential to adjust for the key factor of age

– CDC

Covid-19 Forecasting

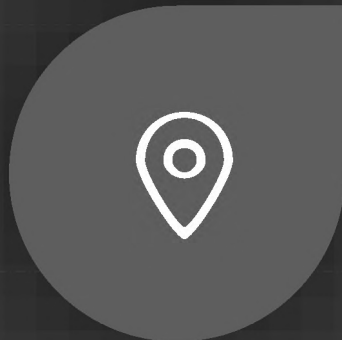
[CDC Forecasting Website](#)

— CDC Covid-19 Forecast Process

CDC work is overseen
by a team at University
of Massachusetts
Amherst



Modeling groups
submit forecasts
to the CDC



Specific organization is
Reich Lab based in the
Department of
Biostatistics and
Epidemiology



Forecasts that meet certain
criteria are combined into
the “Ensemble” model,
which is the forecast model
of record for the CDC

— Modeling submissions to the CDC

Forecast are submitted weekly, on Mondays

Forecasts are submitted via github

All forecast data is public and can be accessed by anyone

Forecasts are submitted in csv files with specific formatting requirements

Submitting a forecast to the Ensemble model requires generation of about 10,000 forecast records / week / team to participate

CDC Covid-19 Forecasts

Types

“Point” forecast
“Prediction interval”
(commonly known
as 95% confidence
ranges)

Quantities

“Cases”
(positive tests)
Deaths

Regions

51+ states and
territories
US National

Time Period

Week 1
Week 2
Week 3
Week 4

— Forecast Teams

Universities

Carnegie Mellon
Columbia
Columbia/UNC
Georgia Tech
Harvard
Iowa State
Johns Hopkins
London School of
Hygiene and Tropical
Medicine
MIT
Northeastern
Notre Dame

Universities

RPI
Texas Tech
UCLA
USC
UW/IHME
University of AZ
U Cal Merced
UCSD
University of Geneva
University of Georgia
U Mass Amherst
U Michigan
U Texas Austin

Research Labs

Los Alamos National Lab
US Army Research and
Development Center
Walmart Labs Data
Science Team
Covid19 Simulator
Consortium (MassGen,
Harvard, Georgia Tech,
Boston Medical Center)

Individuals and Firms

Discrete Dynamical
Systems
Institute for Business
Forecasting
IQVIA
John Burant
Karlen Working Group
LockNQuay
Oliver Wyman
Predictive Science Inc
Qi-Jun Hong
Robert Walraven
Steve McConnell
Youyang Gu

Forecast Teams and Coordination

The better forecast teams and the worst are not what you'd expect them to be, considering some of the institutions involved

There's a weekly forecast call Tuesdays at noon Pacific time to review the week's forecasts

The university teams can be quite academic ("once you know how the sausage is made, you'll never eat sausage again")

Some well-known, highly respected institutions are submitting **terrible** forecasts

Groups use Various Methods

- Massive data sets, e.g., 500 million records
- Esoteric data, e.g., use of mobility data from cell phone records
- Machine learning
- AI
- Bayesian analysis, Monte Carlo simulations, etc.
- Pre-existing infectious diseases models

Model descriptions are available from the [CDC website](#) (many just link to github)



One of the most enduring and useful conclusions from research on forecasting is that simple methods are generally as accurate as complex methods.

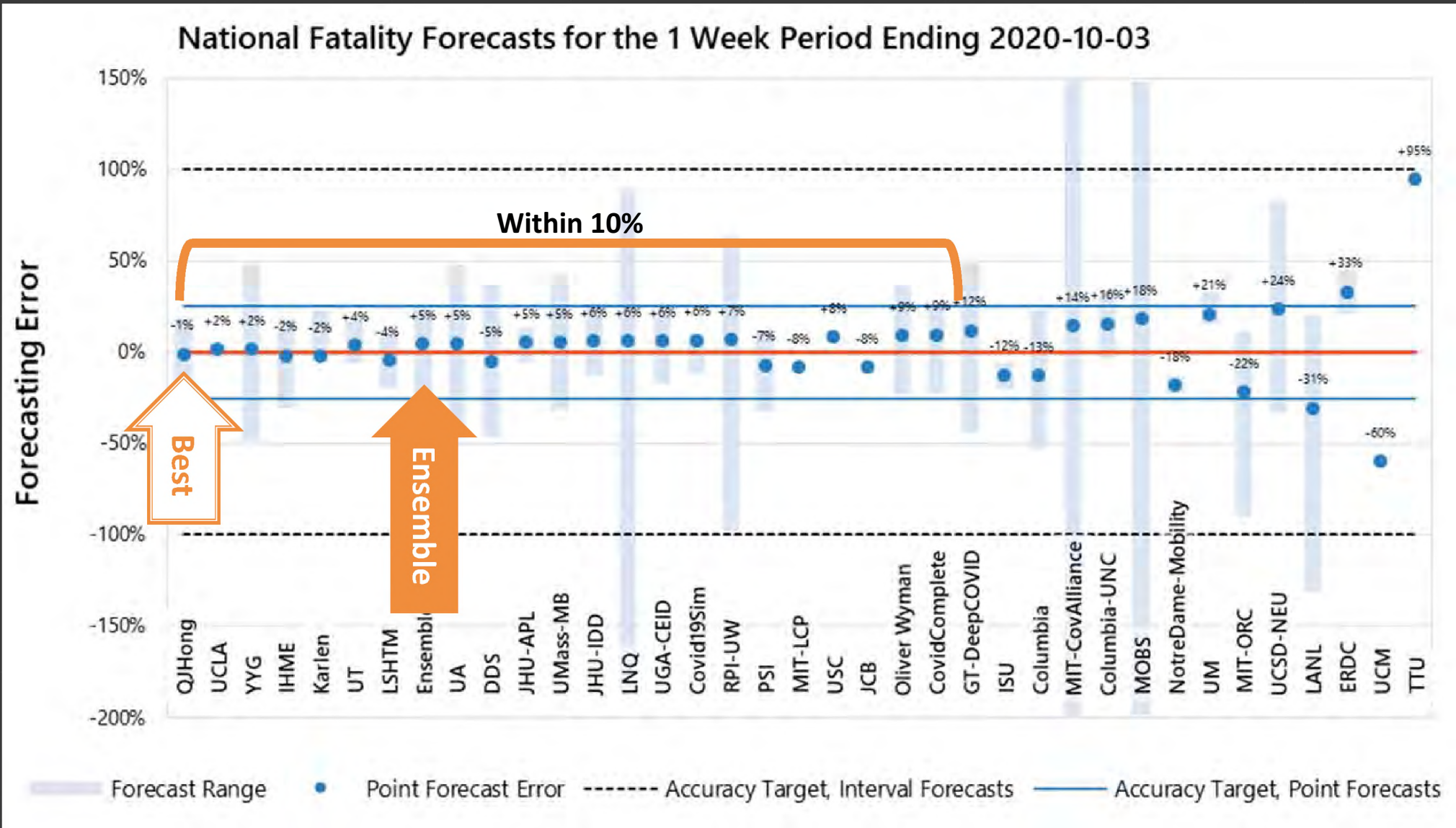
— *J. Scott Armstrong,*
Principles of Forecasting



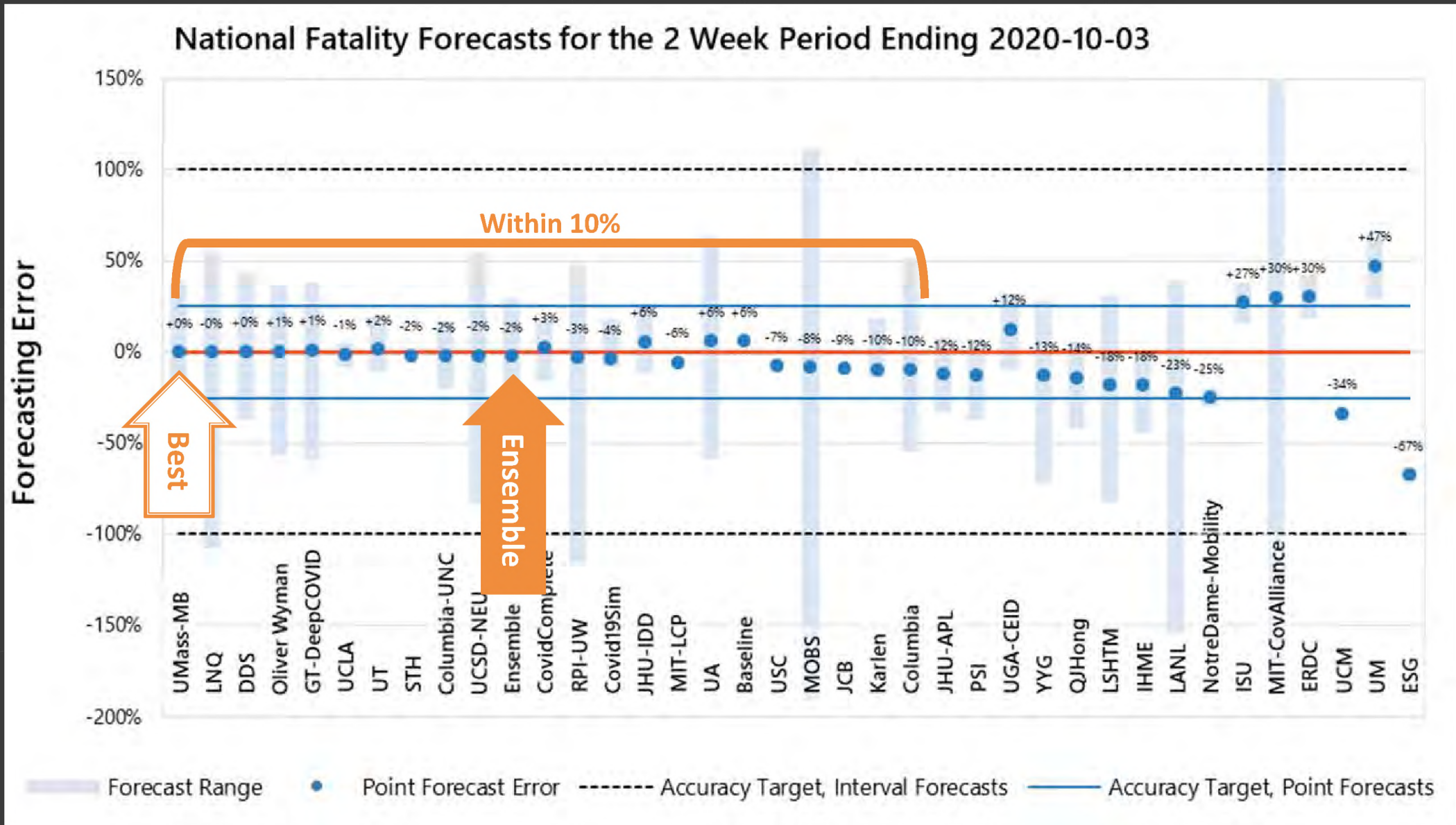


What's Possible with Forecasting Nationally?

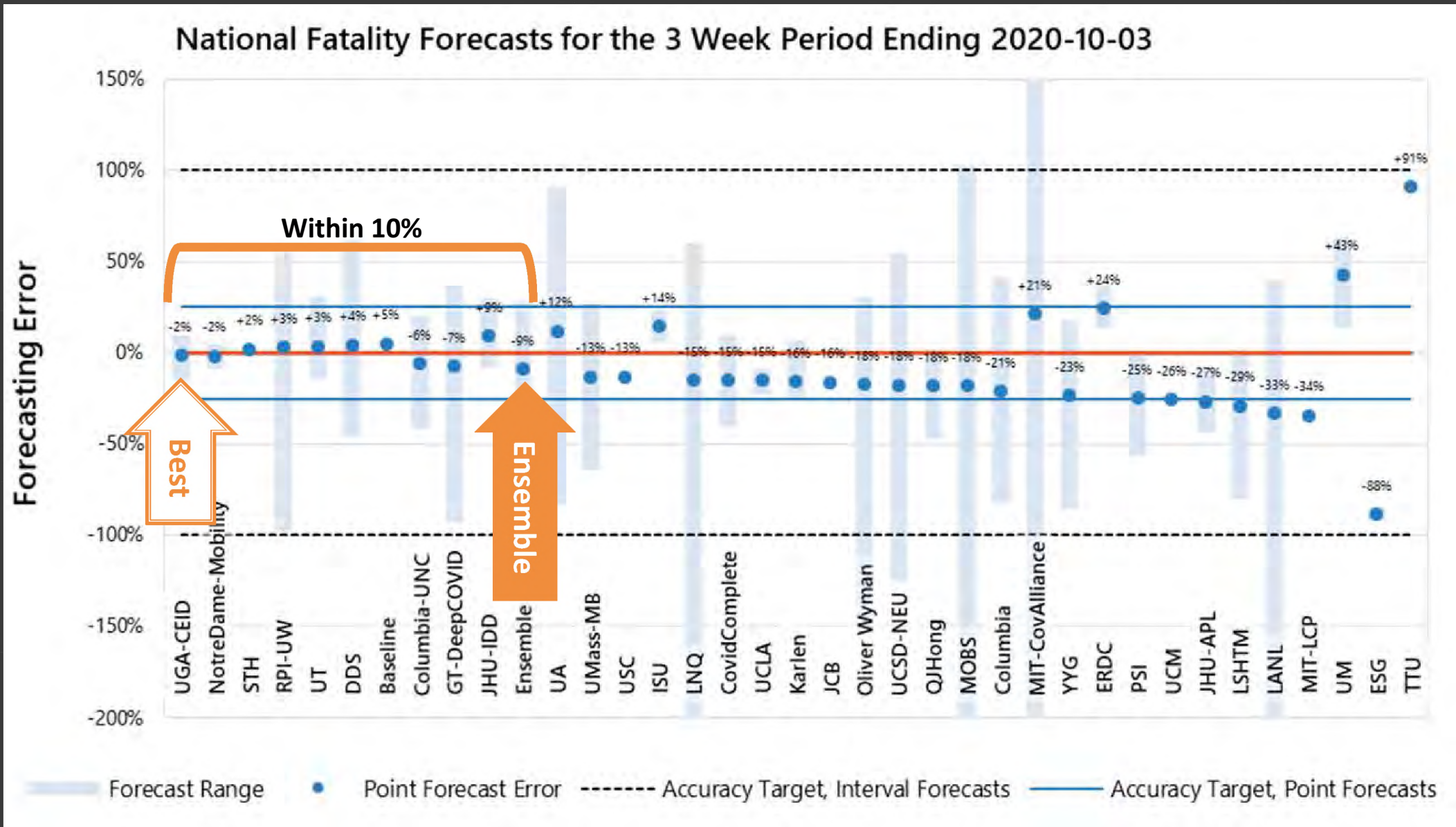
Recent National Forecasts



Recent National Forecasts

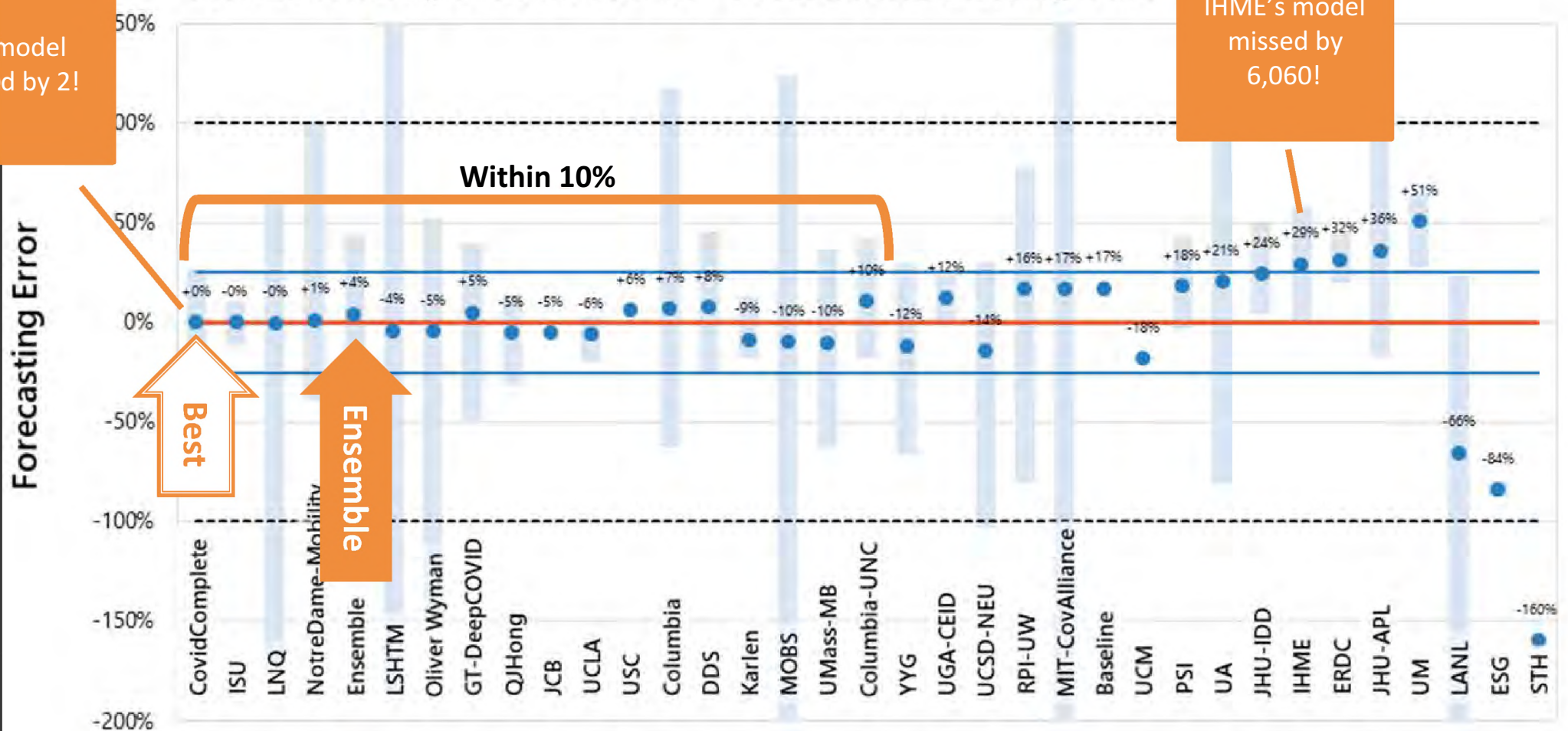


Recent National Forecasts



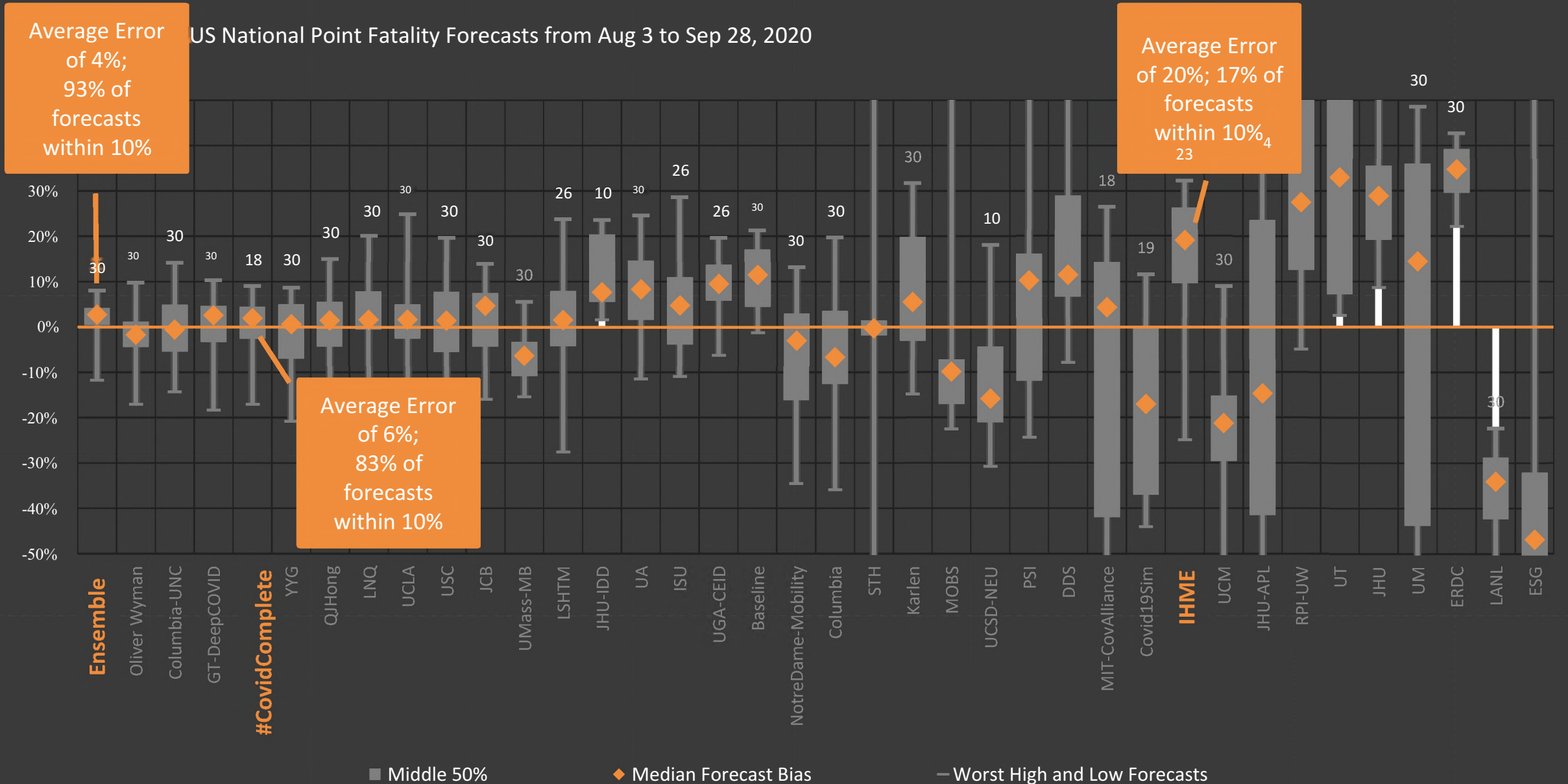
Recent National Forecasts

National Fatality Forecasts for the 4 Week Period Ending 2020-



Forecast Range Point Forecast Error Accuracy Target, Interval Forecasts Accuracy Target, Point Forecasts

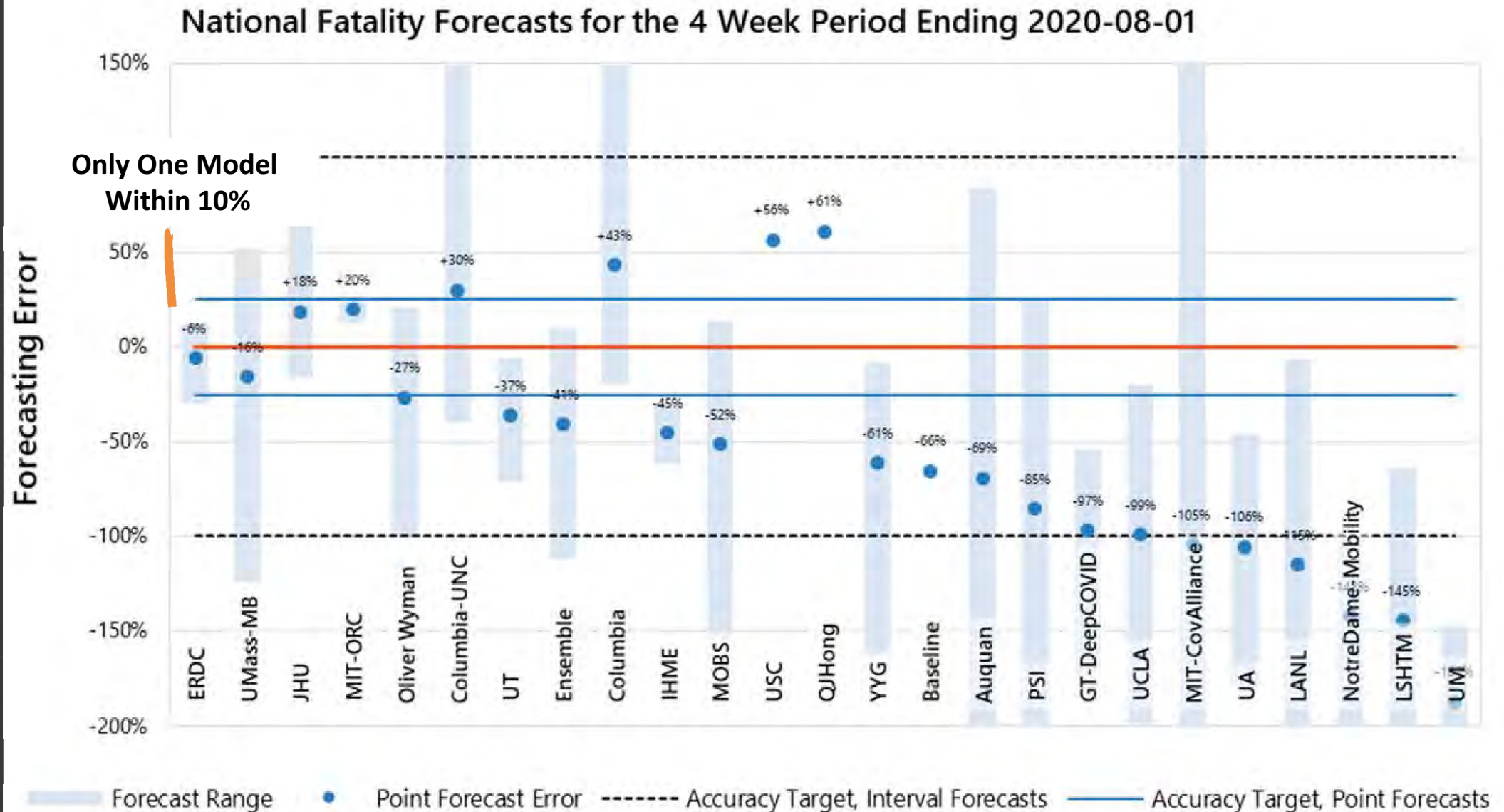
Accuracy over time for 1, 2, 3, and 4-week forecasts



Covid-19

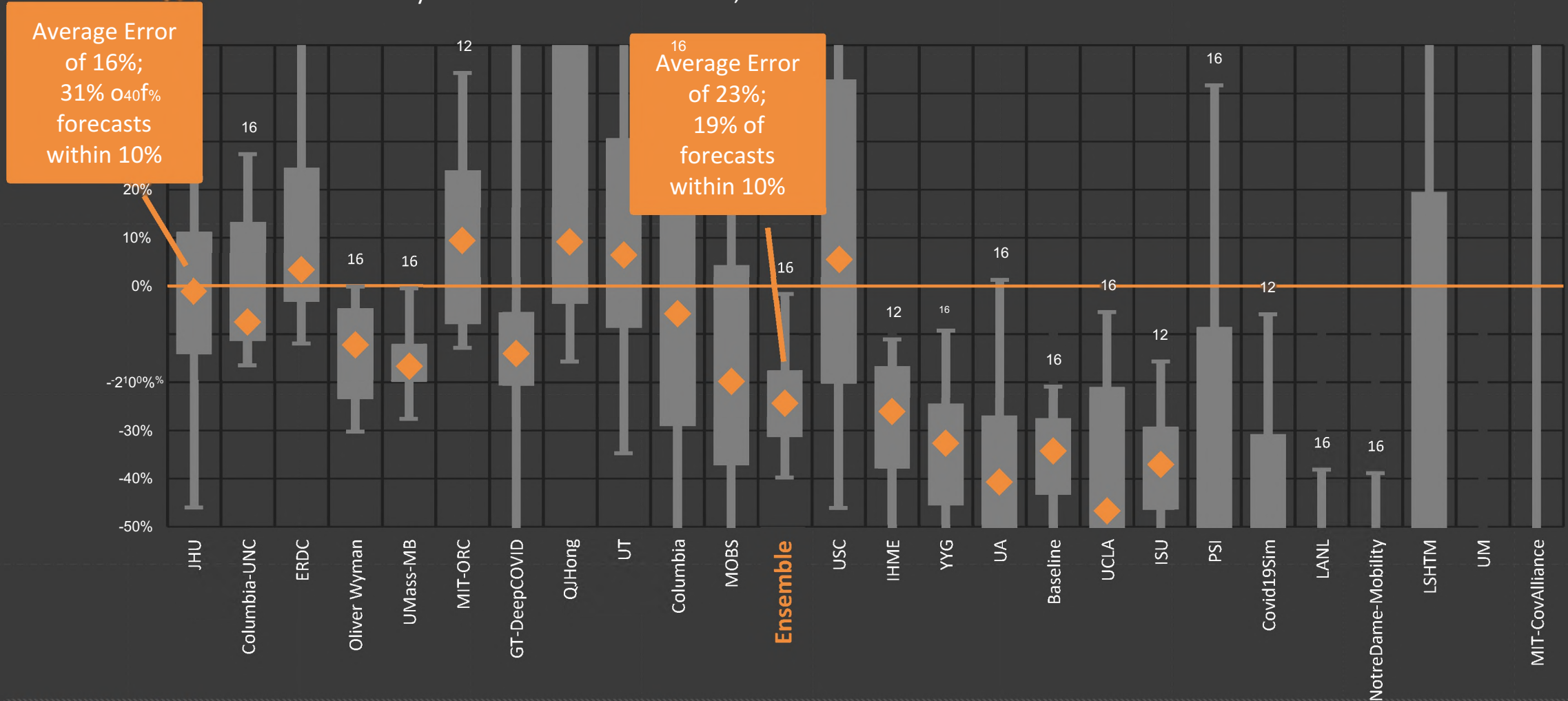
National Forecasts Have
Improved Markedly
Since July

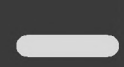
Earlier National Forecasts



Longer-term history of earlier national forecasts

US National Point Fatality Forecasts from Jul 6 to Jul 27, 2020





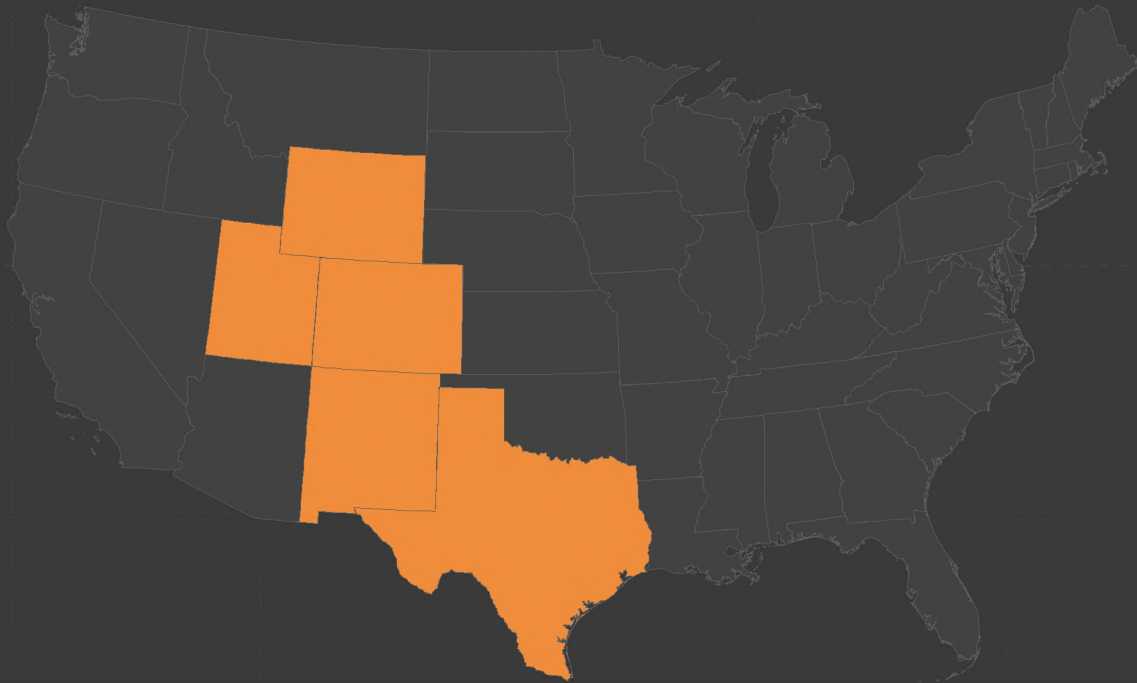
MMoorree NNaattiionnaall FFoorreeccaasstt EEvvaalluaa



www.ssteevemmcccoonnneelll.ccoomm//ccoovviidd



What's Possible with Forecasting at the State Level?

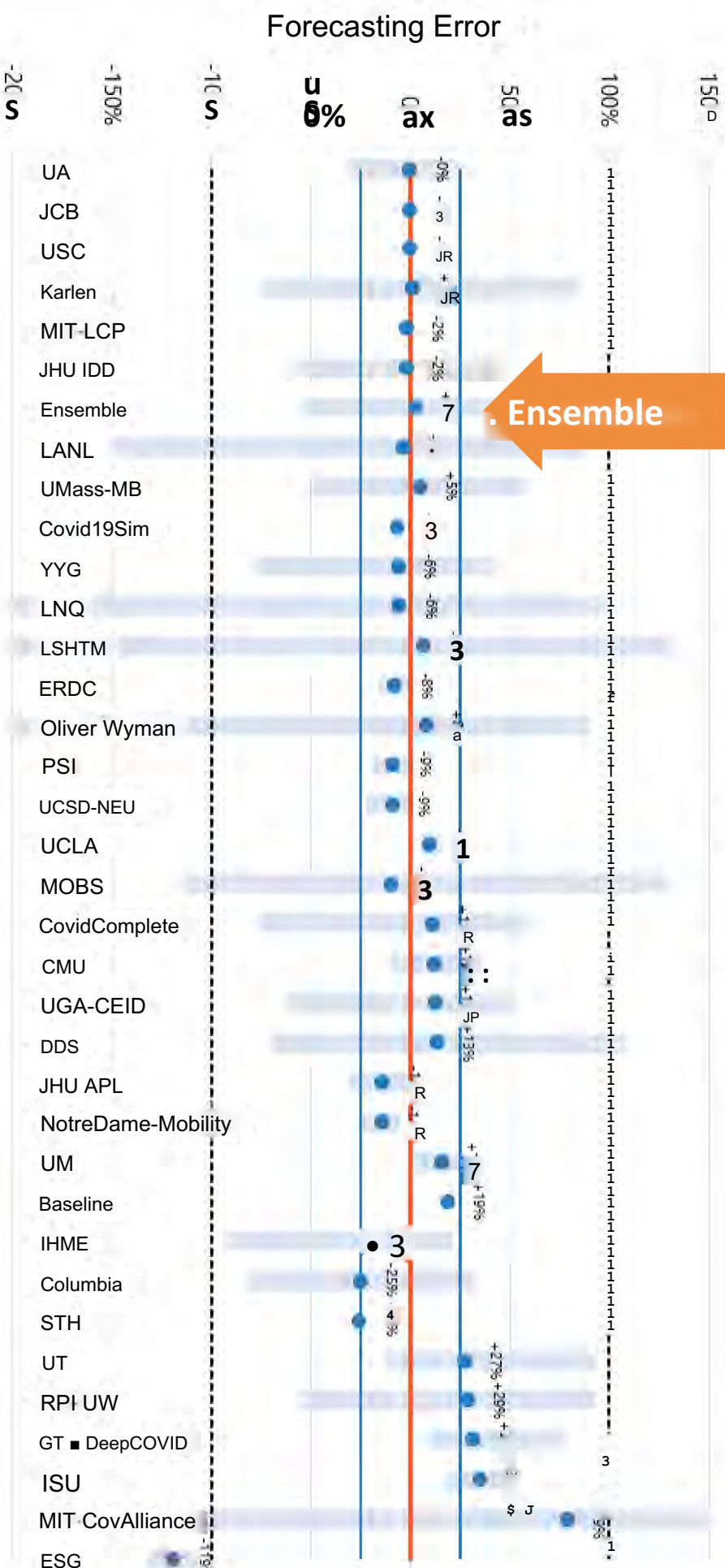


State Level Forecasts

- Numbers of total tests, positive tests, cases, and deaths are smaller
- Data is rougher
- Models tend to do better on states with bigger numbers
- Accurate forecasting is arguably more important than national forecasts

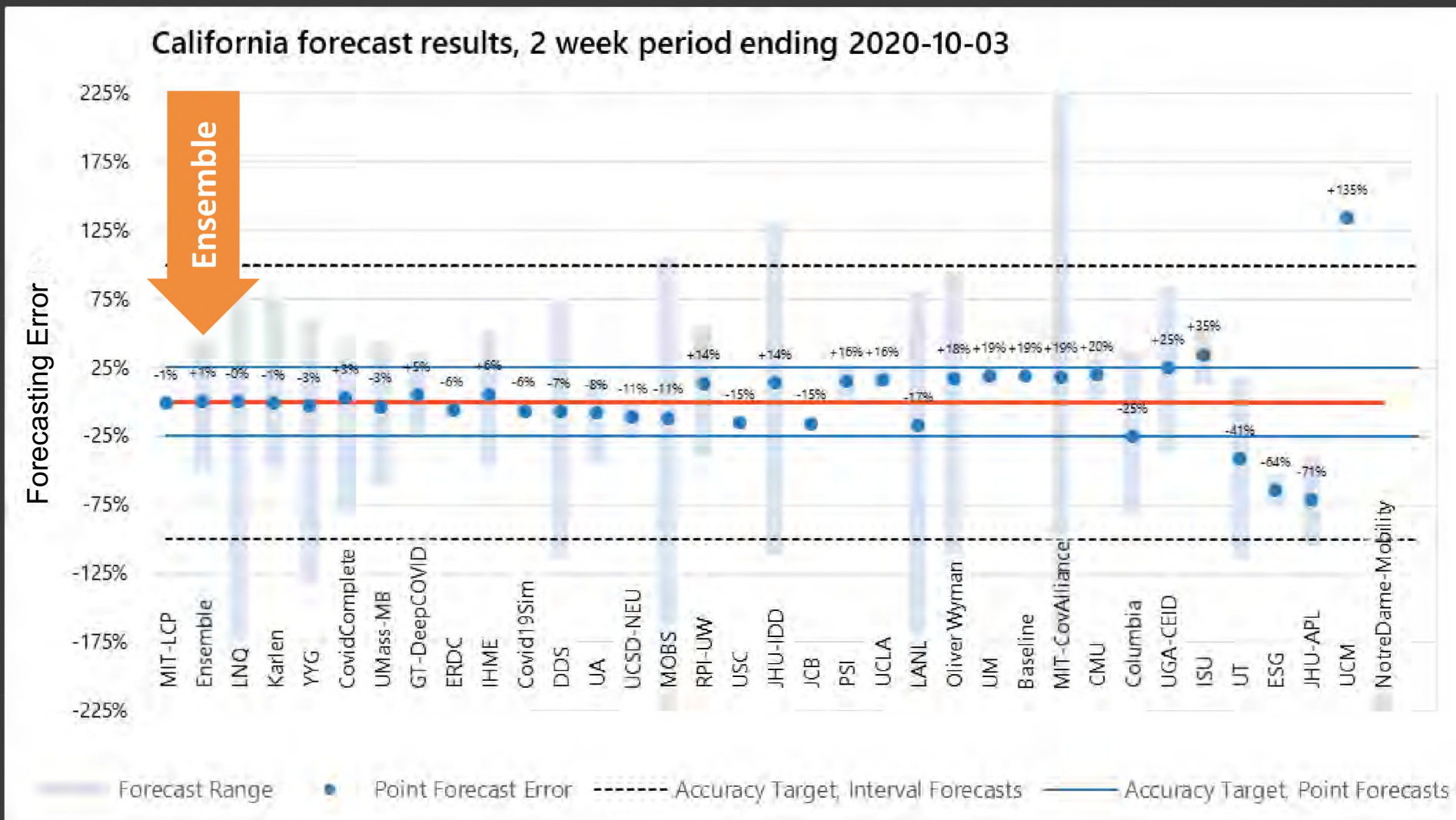
Statog Forecast Example

Example forecast results, 2 week period ending 2020-10-08

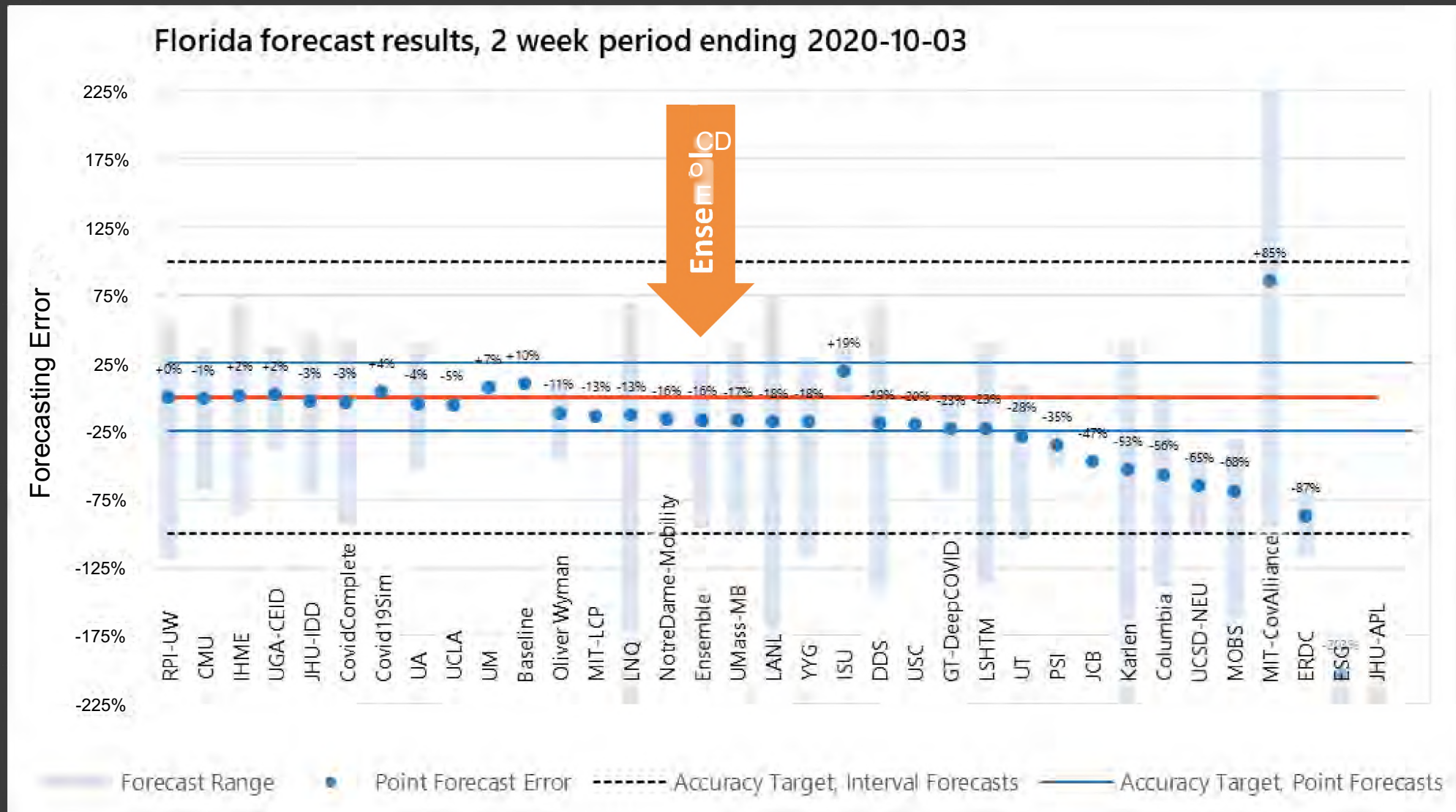


Forecast Range · Point Forecast Error · Accuracy Target Interval Forecasts · Accuracy Target Point Forecasts

State Forecast Examples

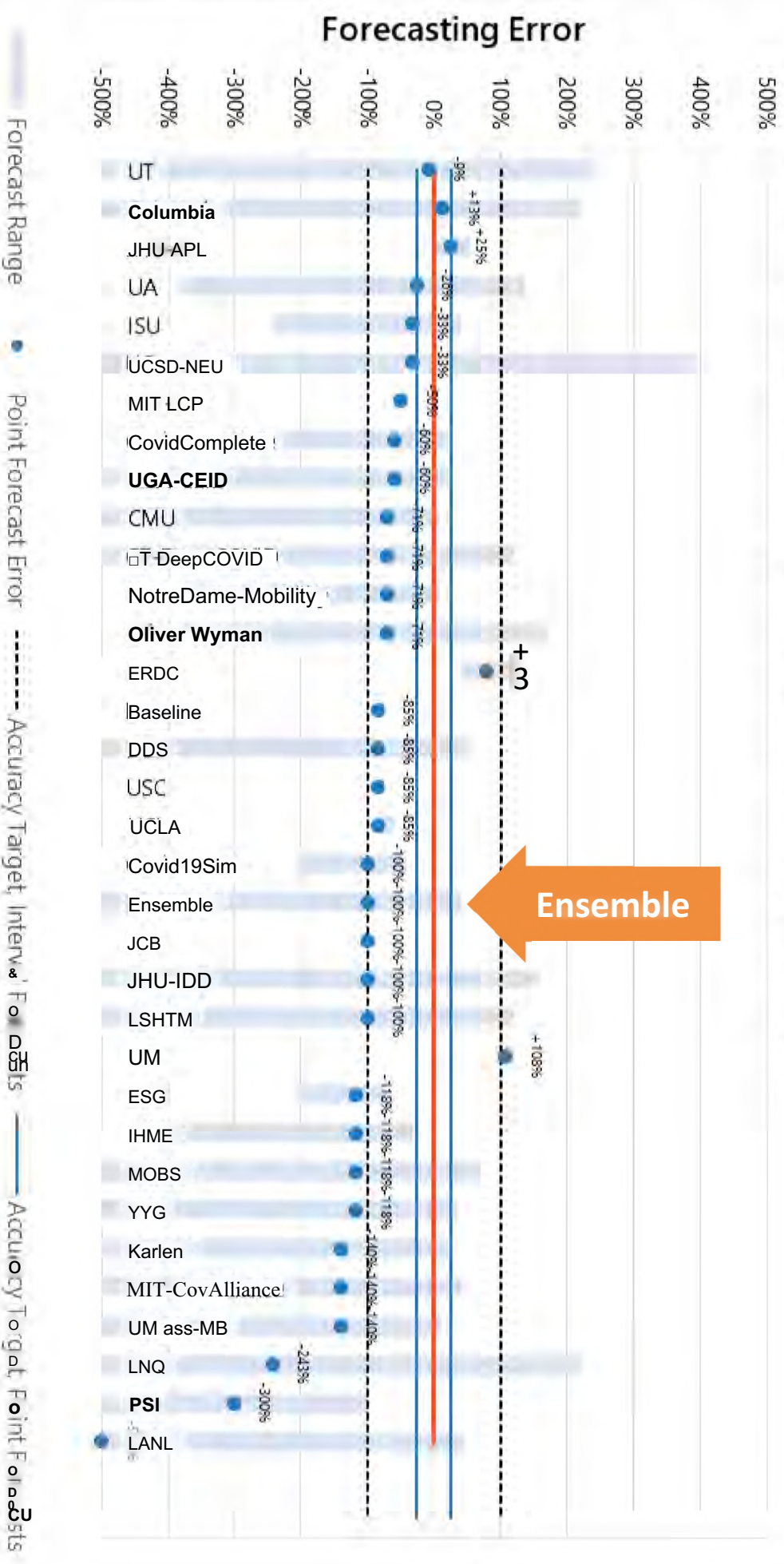


State Forecast Examples



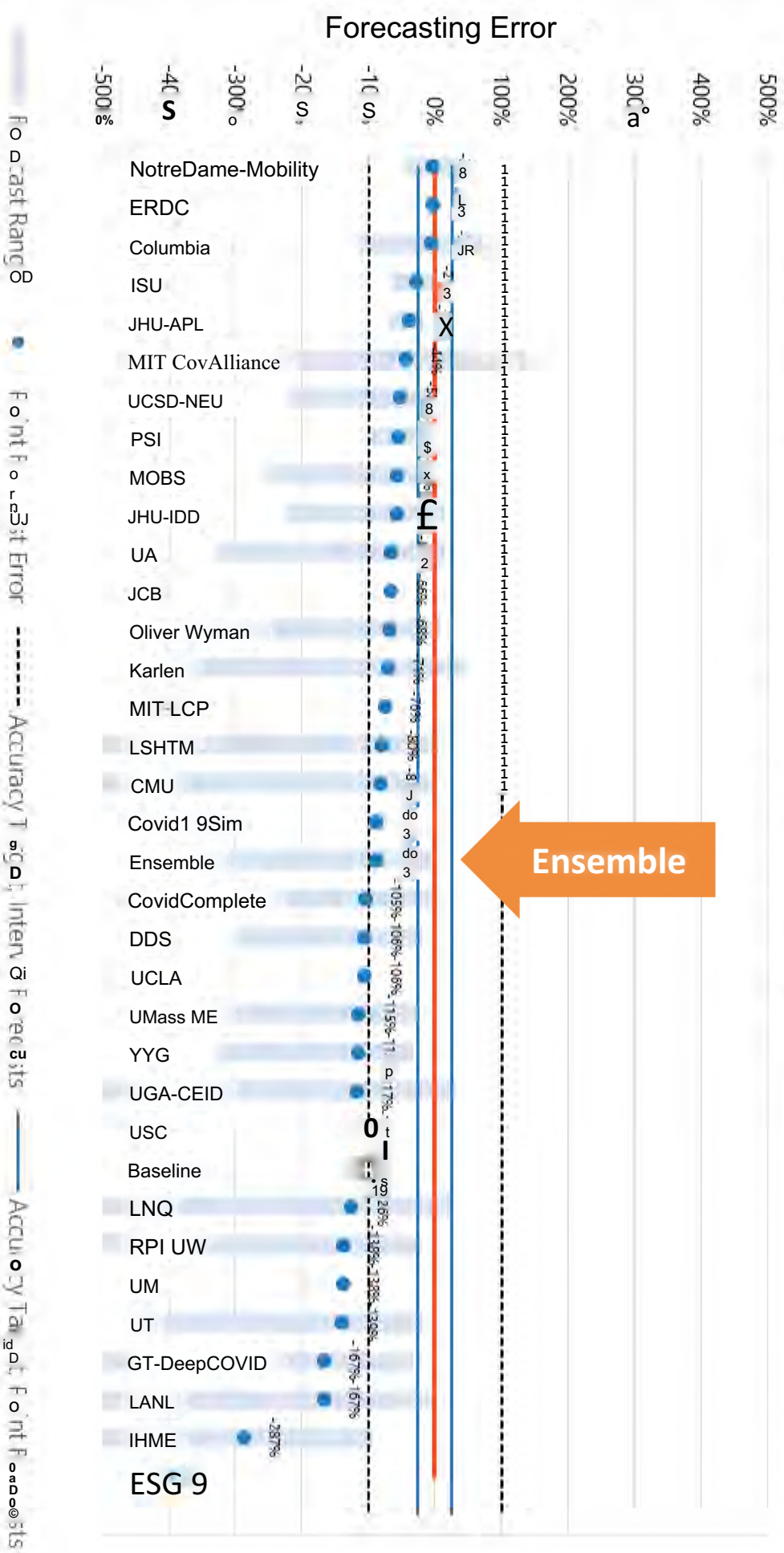
State Forecast Examples

Delaware forecast results, 2 week period ending 2020-10-02

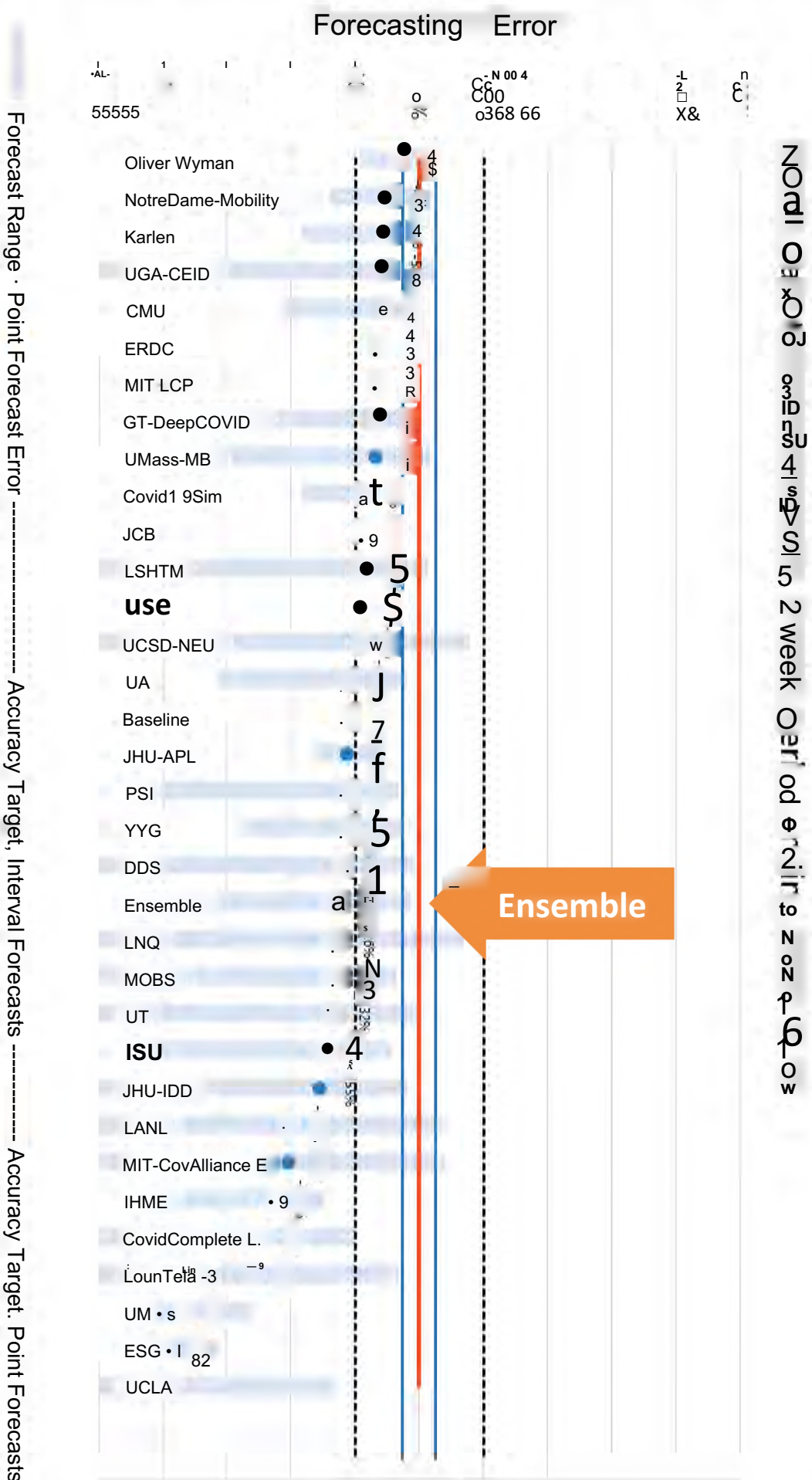


State Forecast Example

Monday forecast results, 2 week period ending 2020-10-03

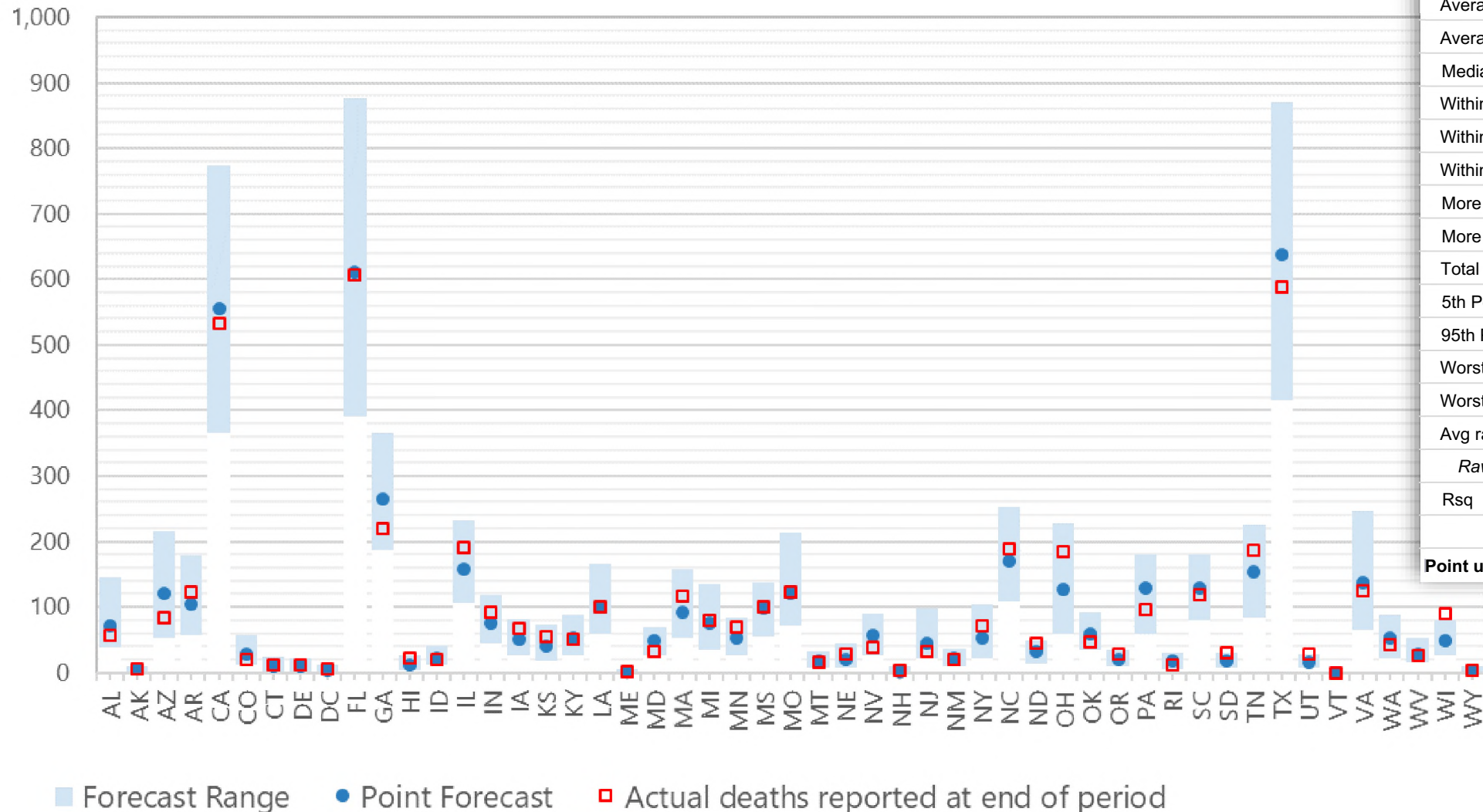


Forecast Examples



State Forecasting at its Best: 1 Week Horizon

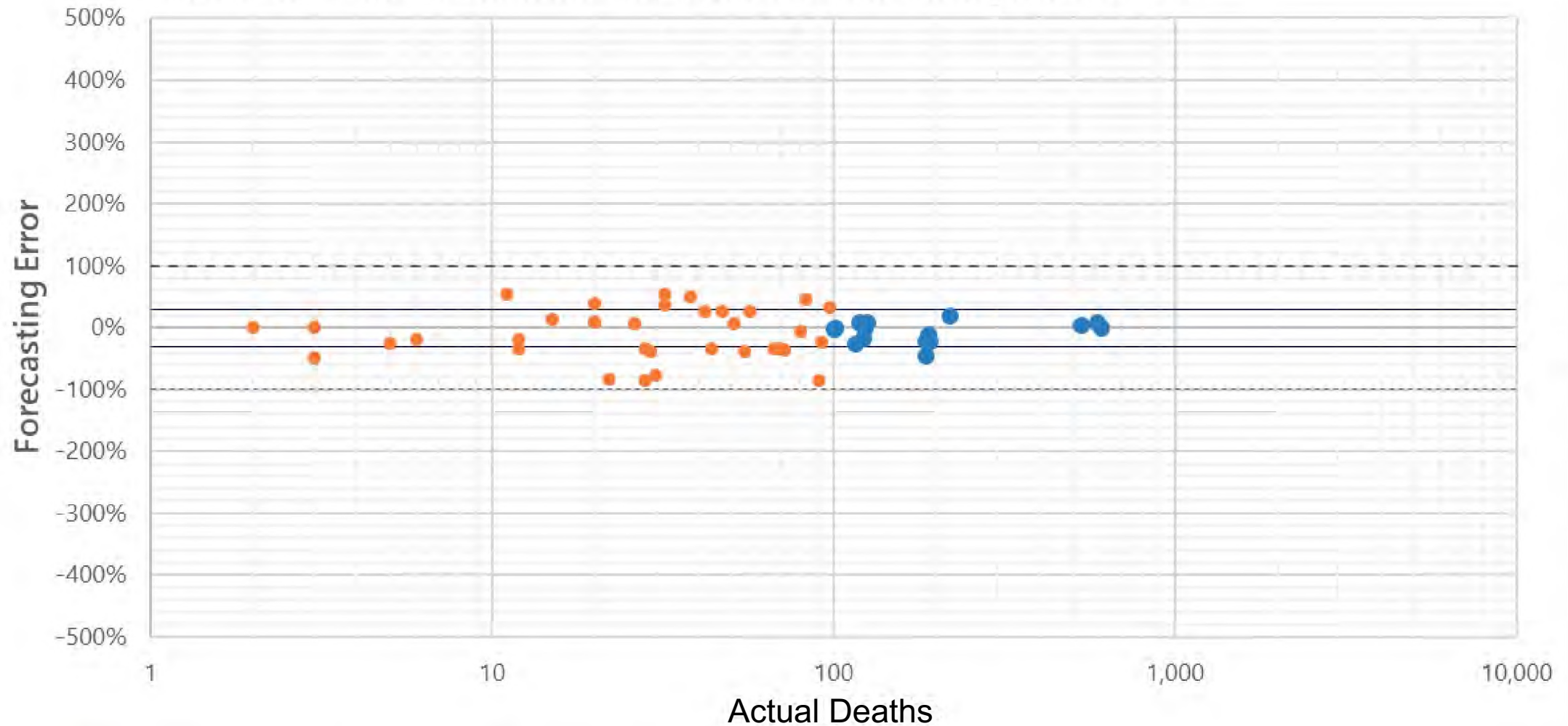
Ensemble Model Performance for 1 Week Period Ending 2020-10-03



Point Forecasts	
Median error magnitude:	25%
Average error magnitude	27%
Average bias:	-8%
Median error bias:	-1%
Within 10%:	31%
Within 25% of actual:	53%
Within 50% of actual:	88%
More than 100% high	0%
More than 100% low	0%
Total Missed by more than 100%	0%
5th Percentile:	-80%
95th Percentile:	+47%
Worst overestimate:	+55%
Worst underestimate:	-87%
Avg raw error	14
Raw error as %	15%
Rsqr	0.98
Point usefulness	
71%	

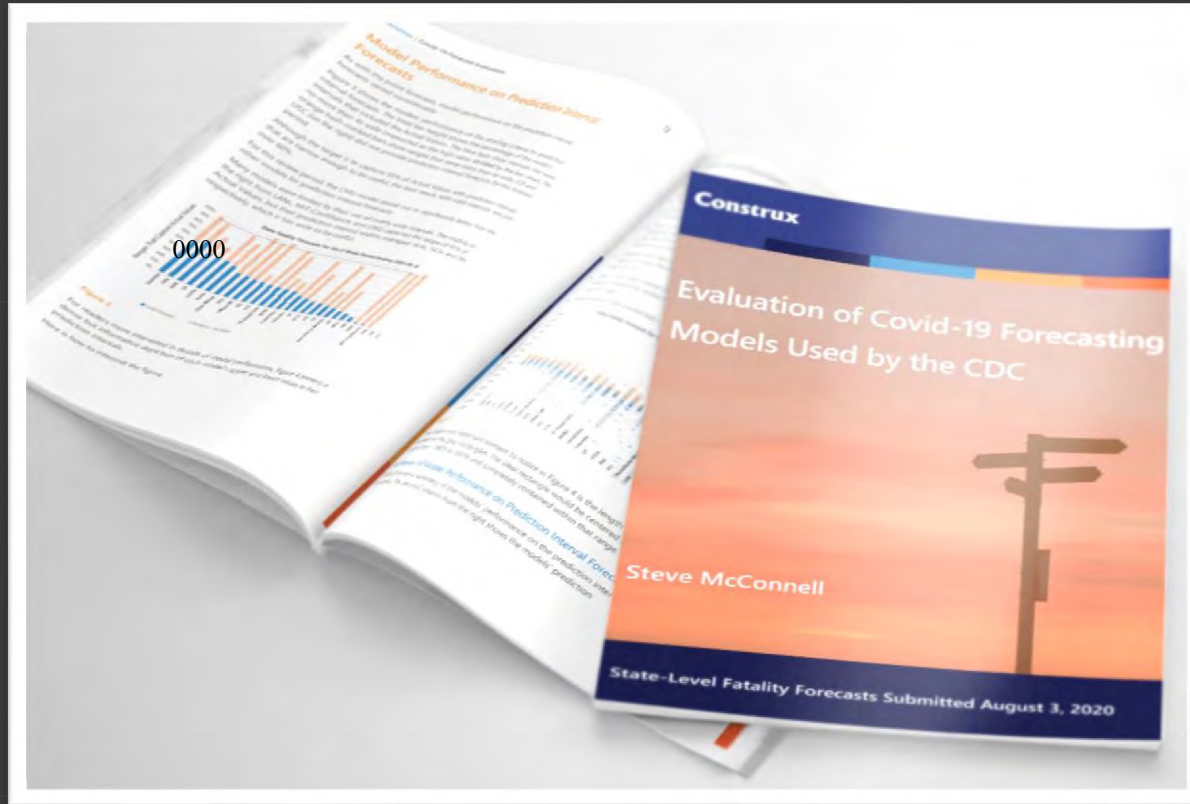
— State Forecasting at 11 Weeks

Ensemble Model Performance for 1 Week Period Ending 2020-10-03



- States with <100 deaths
- States with ≥ 100 deaths

— MMoorree SSttaattee FFoorreeccaasstt EEvvaalluaattiioon



www.steveemccoonneel.com/coovid

— More Overall Forecast Evaluations



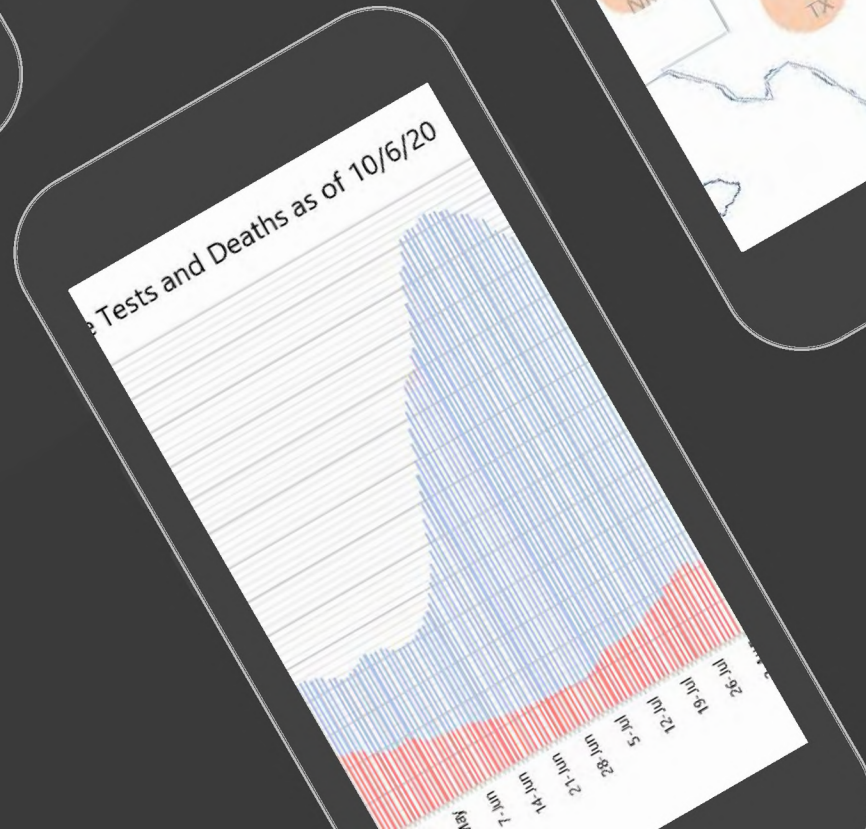
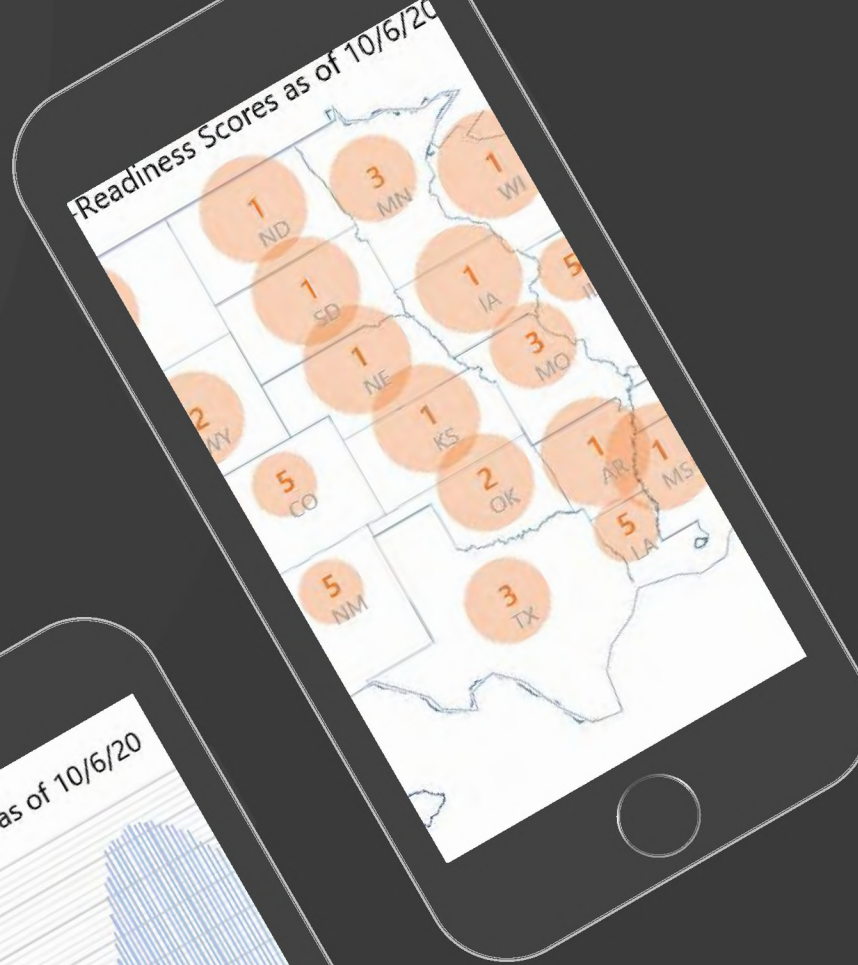
www.stevemcconnell.com/covid



Summary

Summary

- Accurate forecasts are possible on 1-4 week horizons
- National forecasts were not accurate into early summer, but have become quite accurate in late summer and fall
- State forecasts are not as accurate as national forecasts; forecasts for bigger states tend to be more accurate, but not always
- The most accurate forecasts come from private individuals as often as they come from universities; the most prestigious universities are not producing accurate forecasts
- The CDC's Ensemble model is consistently among the most accurate forecast models for nearly all periods and types of forecasts



Covid-19 Spin-Free Data Center

stevemcconnell.com/covid

— Useful Links

- SteveM's Covid-19 Data Center
<https://www.stevemccconnell.com/covid>
- SteveM's article on detailed age and comorbidity risk
<https://medium.com/@stevemcc/the-one-graph-you-need-to-understand-covid-19-comorbidities-5c7c8b64254f>
- CDC list of specific comorbidity risks
<https://www.cdc.gov/coronavirus/2019-ncov/need-extra-precautions/evidence-table.html>
- CDC forecast hub
<https://viz.covid19forecasthub.org/>
- Washington state data dashboard
<https://coronavirus.wa.gov/what-you-need-know/covid-19-risk-assessment-dashboard>
- Tableau international comparison data
https://public.tableau.com/profile/jonas.nart#!/vizhome/COVID19_15844962693420/COVID19-TrendTracker

— Discussion