

Replace *conventional certitude* with strategic communication of uncertainty

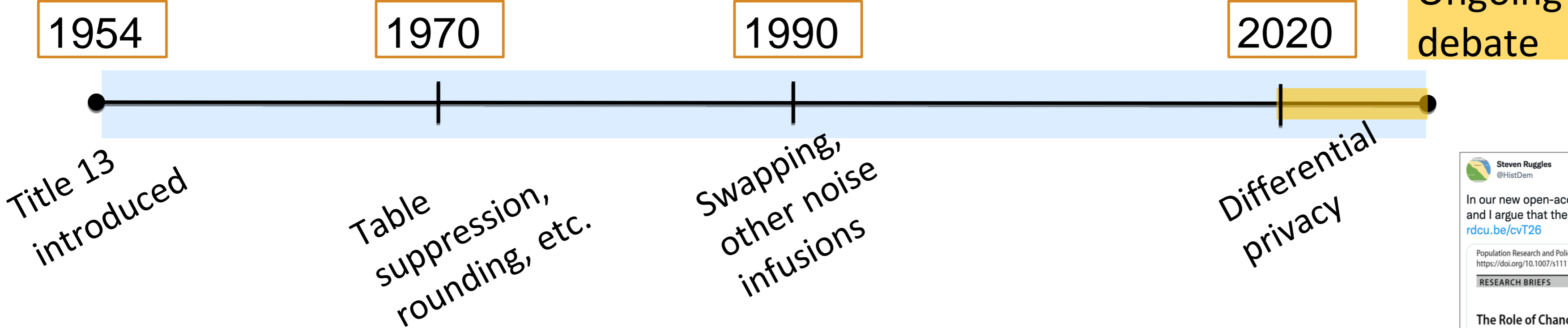
Jessica Hullman

Ginni Rometty Associate Professor of Computer Science
Northwestern University

*Term coined by Manski 2011

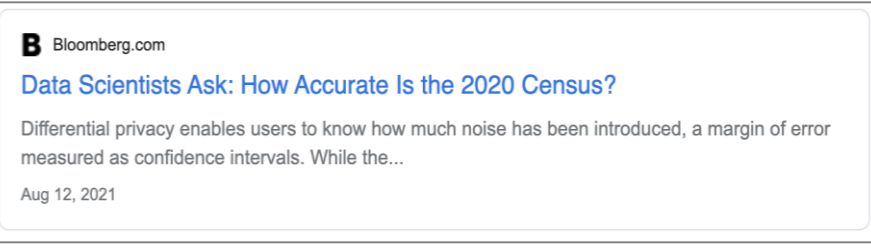
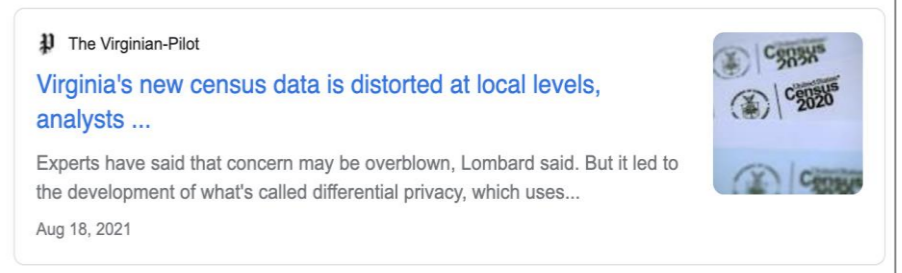
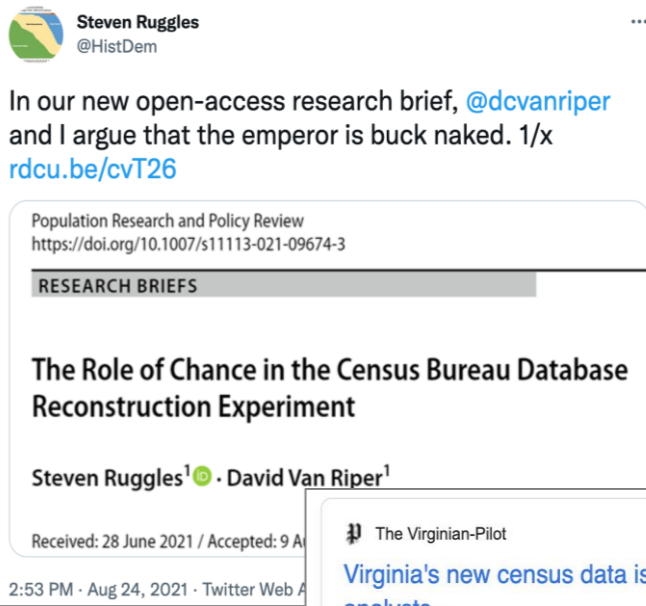
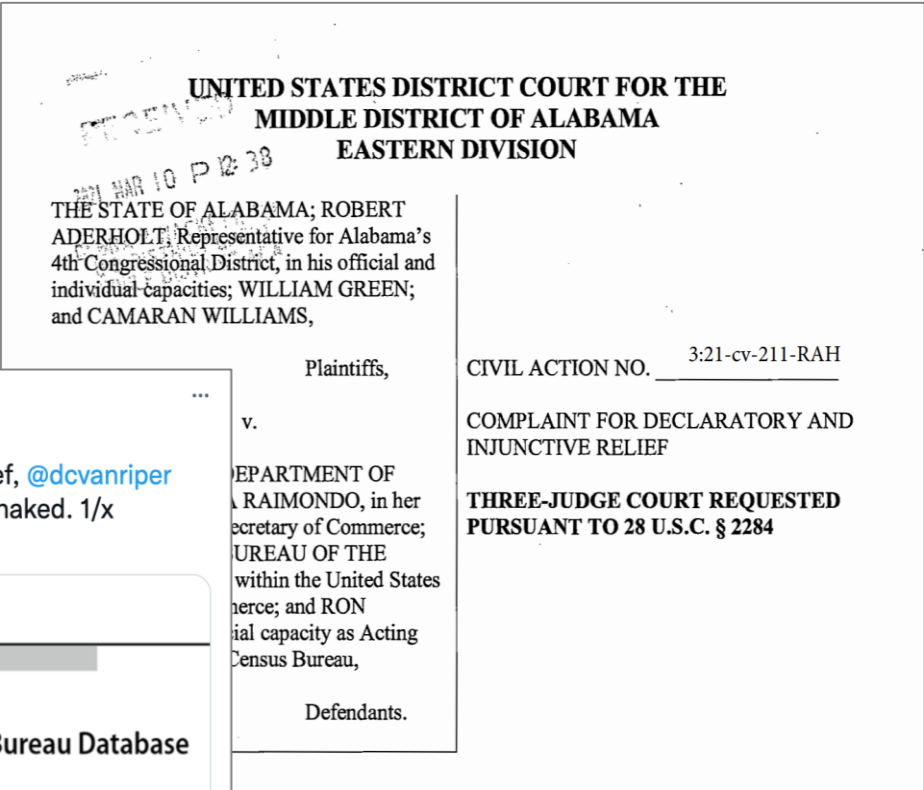
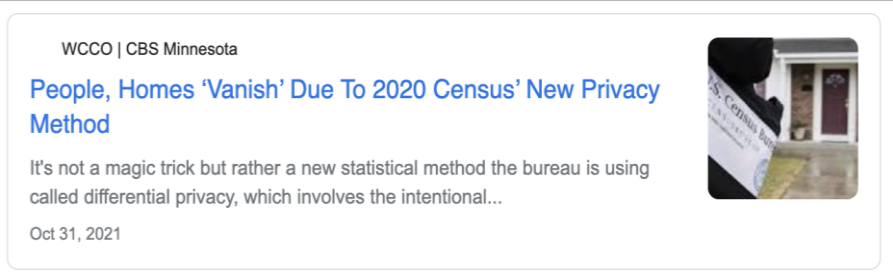


Legitimacy of the US Census at risk



New DAS is not a “sea change,” but is interpreted as such in light of pervasive **conventional certitude**

U.S. Population: 332,574,351
 332,574,350?
 332,574,400?
 332,574,000?
 332,570,000?



Conventional certitude pervades reporting

332,574,351 in US

Census Bureau

\$3995 bil baseline budget

Congressional Budget Office

888k on temporary layoff

Bureau of Labor Statistics

79,571,321 new cases

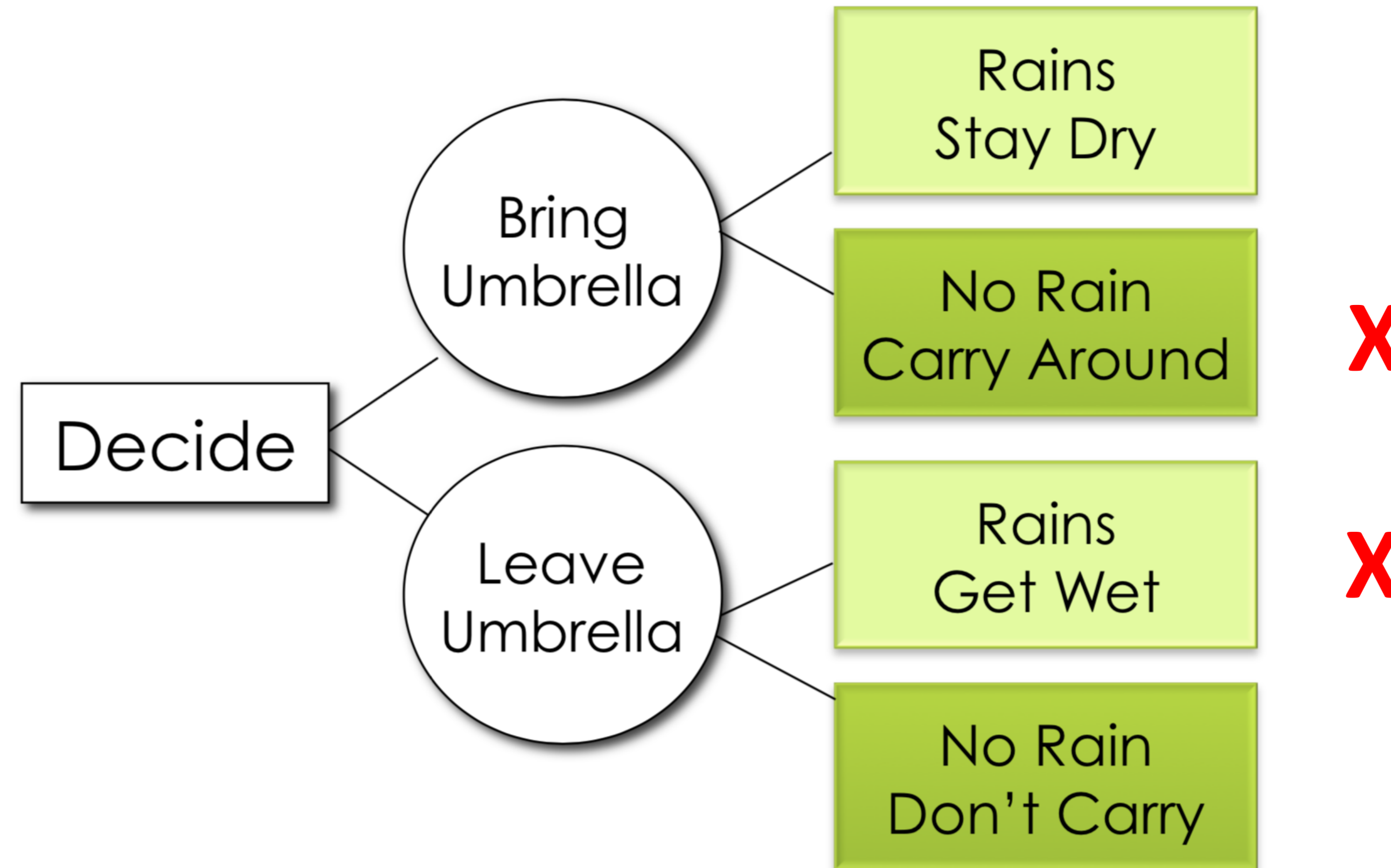
Centers for Disease and Control

975,513 total deaths

\$22,997.5 bil GDP

Bureau of Economic Analysis

Communicating uncertainty improves trust, decisions



Joslyn & LeClerc 2011, Johnson & Slovic 1995, Johnson 2003.

Reducing data to point estimates cannot yield diversification or information acquisition → ***Express uncertainty with all point estimates.***

Implement strategic uncertainty communication

Quantify and visualize uncertainty to imply importance

Use frequency framing to make probability concrete

Suppress “as-if optimization” at all levels of the info hierarchy

Quantify and/or acknowledge transitory uncertainty

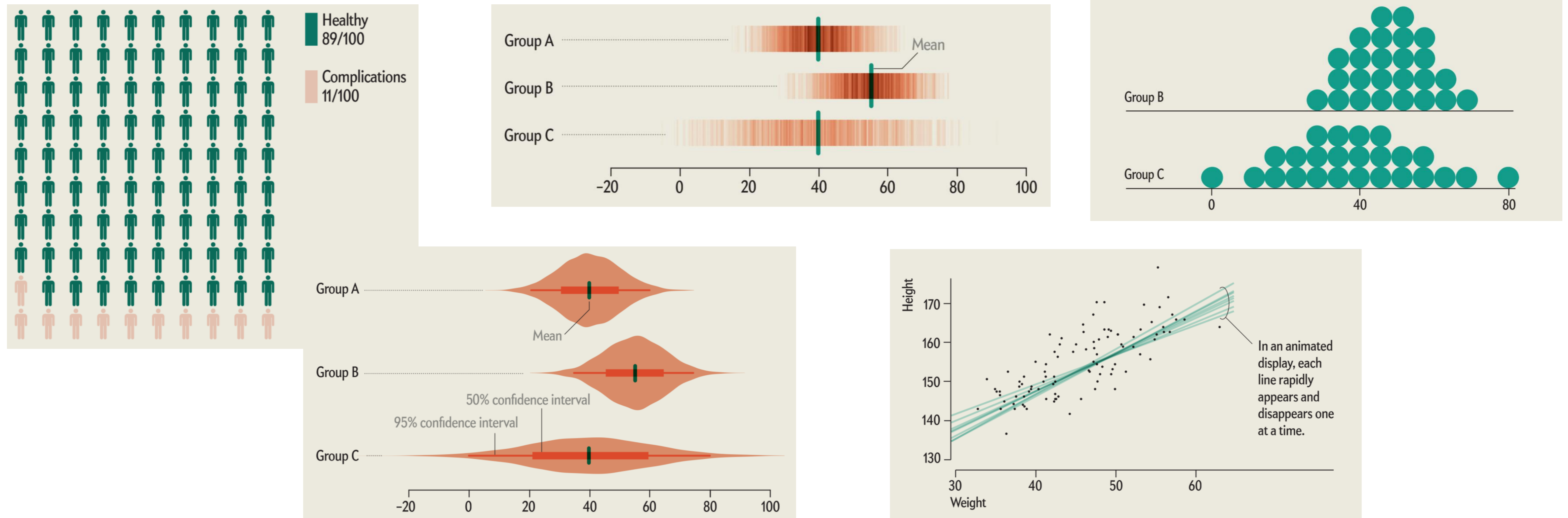
Provide prior forecast error by default

Label forecasts as hypothetical outcomes from statistical experiments

Label partial uncertainty expressions as incomplete

Quantify and visualize uncertainty

Graphs get attention, makes comparisons salient, and convey importance.



Frame probability as frequency to ease reasoning

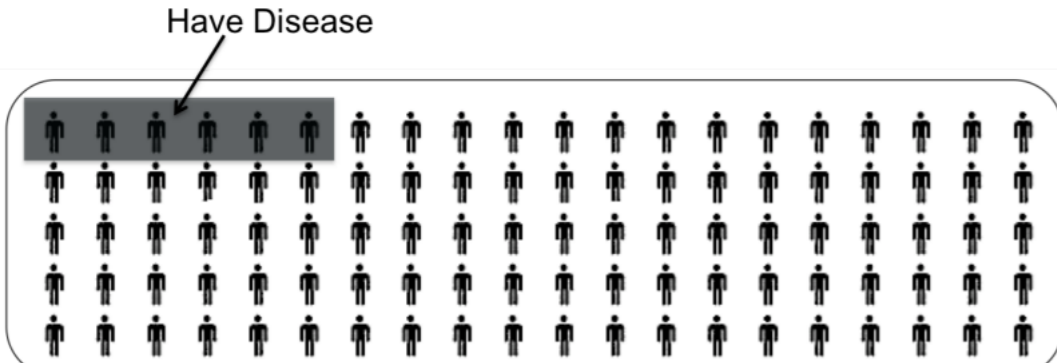
Gigerenzer and Hoffrage, 1995.

There is a total of 100 people in the population.



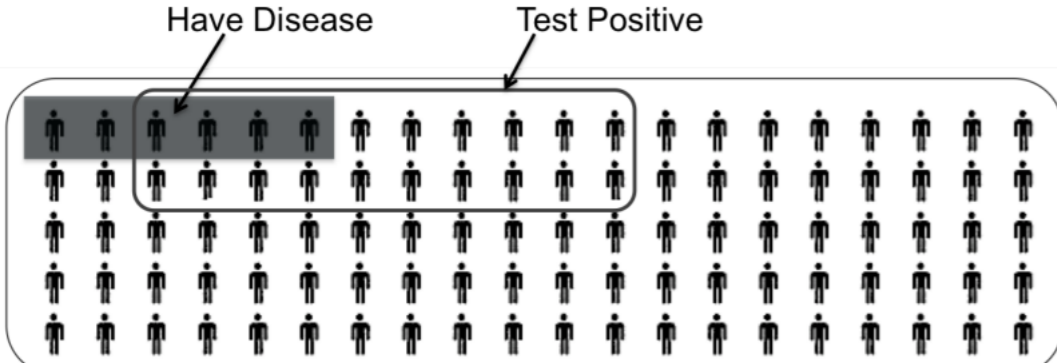
Population

Out of the 100 people in the population, 6 people actually have the disease.



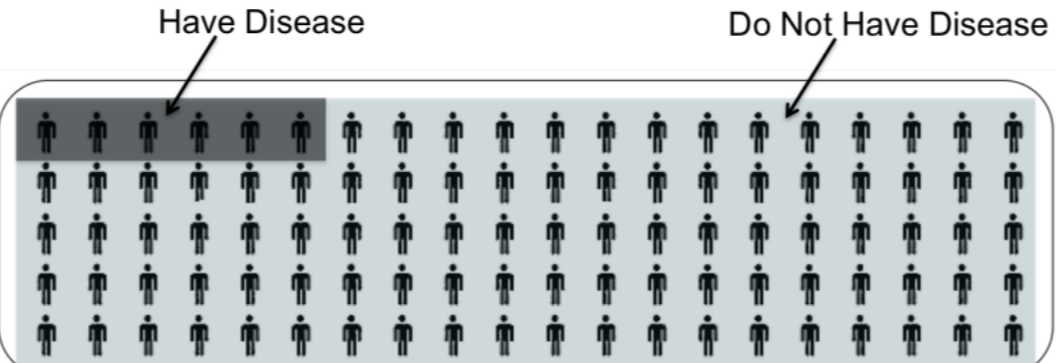
Population

Out of these 6 people, 4 will receive a positive test result and 2 will receive a negative test result.



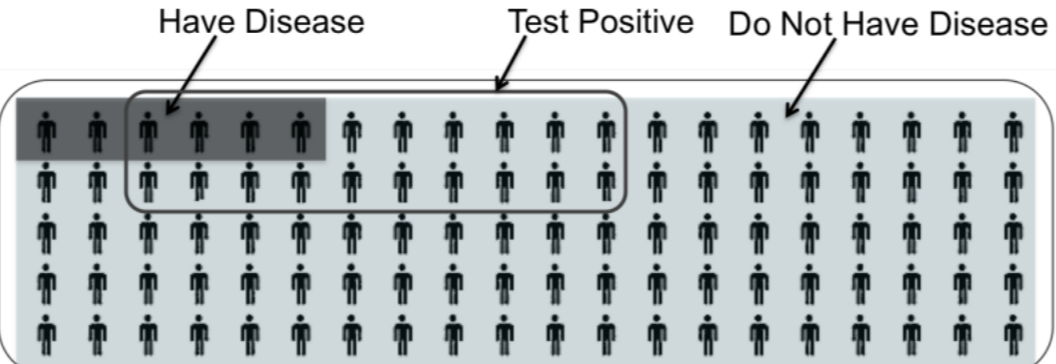
Population

On the other hand, 94 people do not have the disease (i.e., they are perfectly healthy).



Population

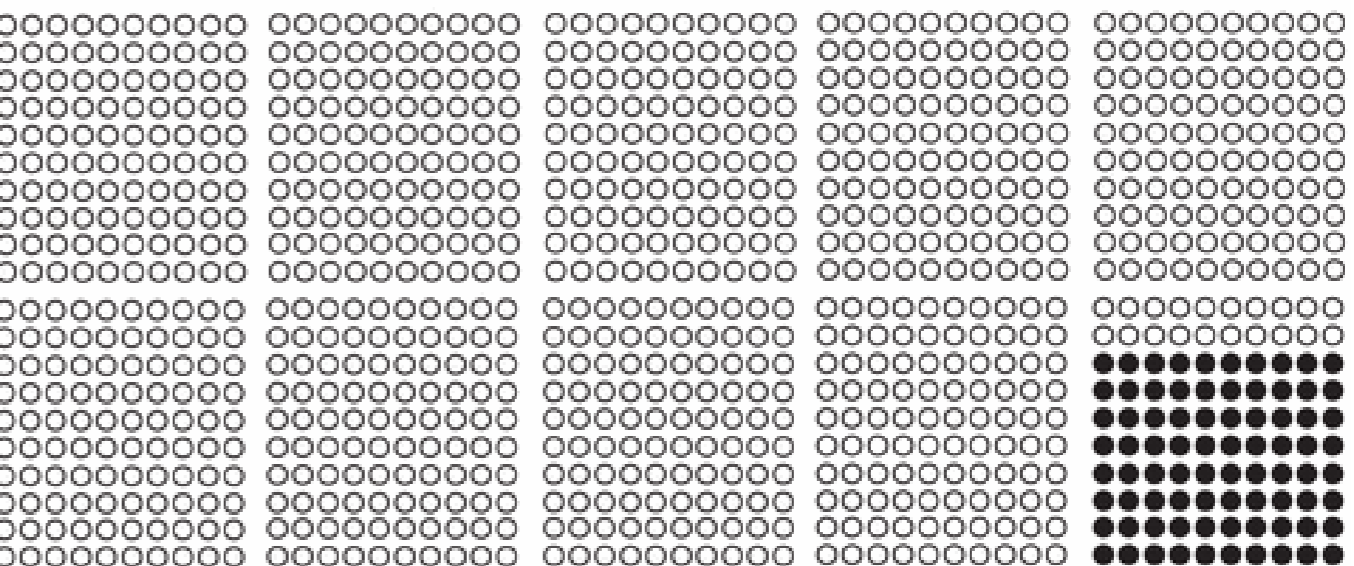
Out of these 94 people, 16 will receive a positive test result and 78 will receive a negative test result.



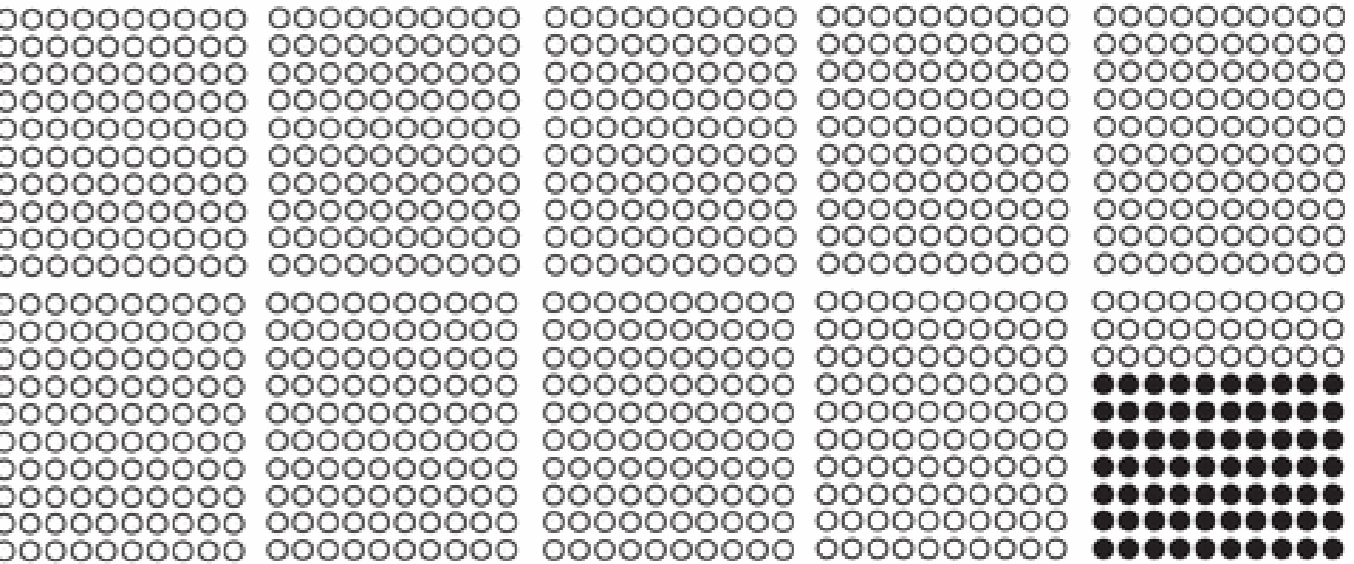
Population

For people with symptoms of arterial disease, aspirin can reduce the risk of having a stroke or heart attack by 13%.

Without aspirin



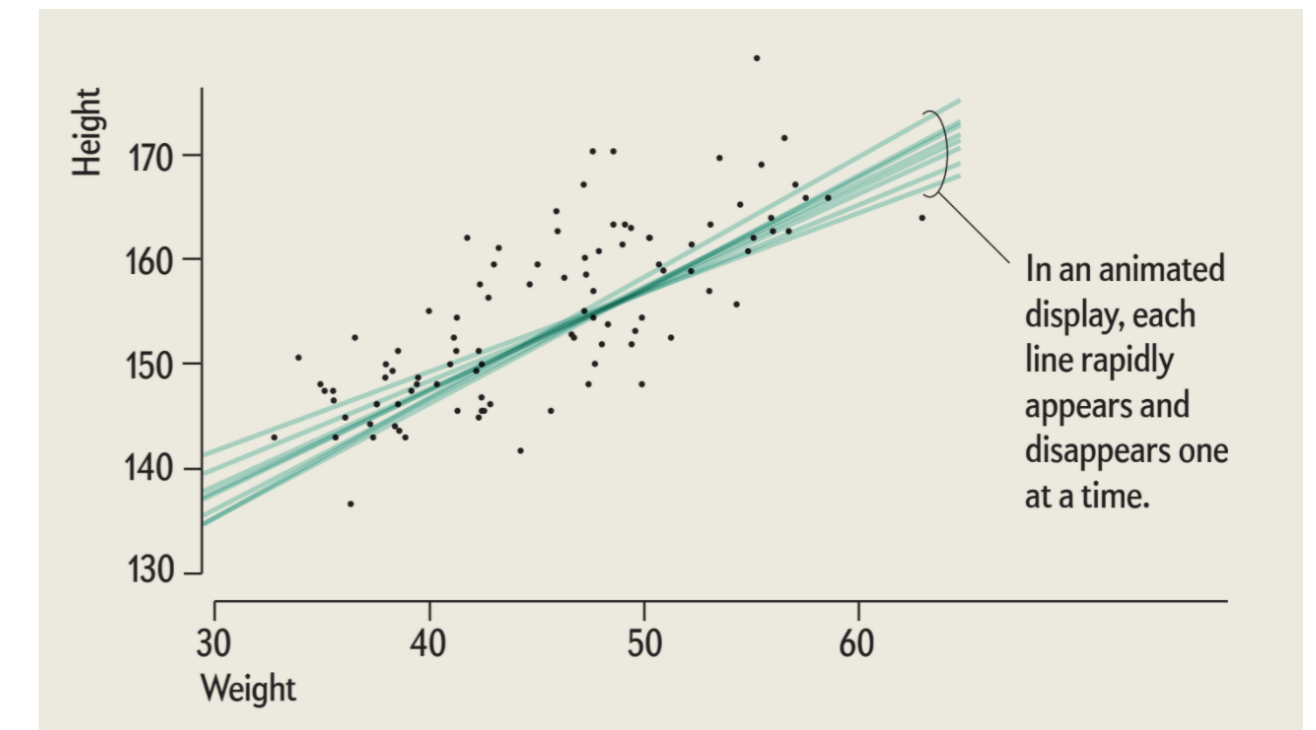
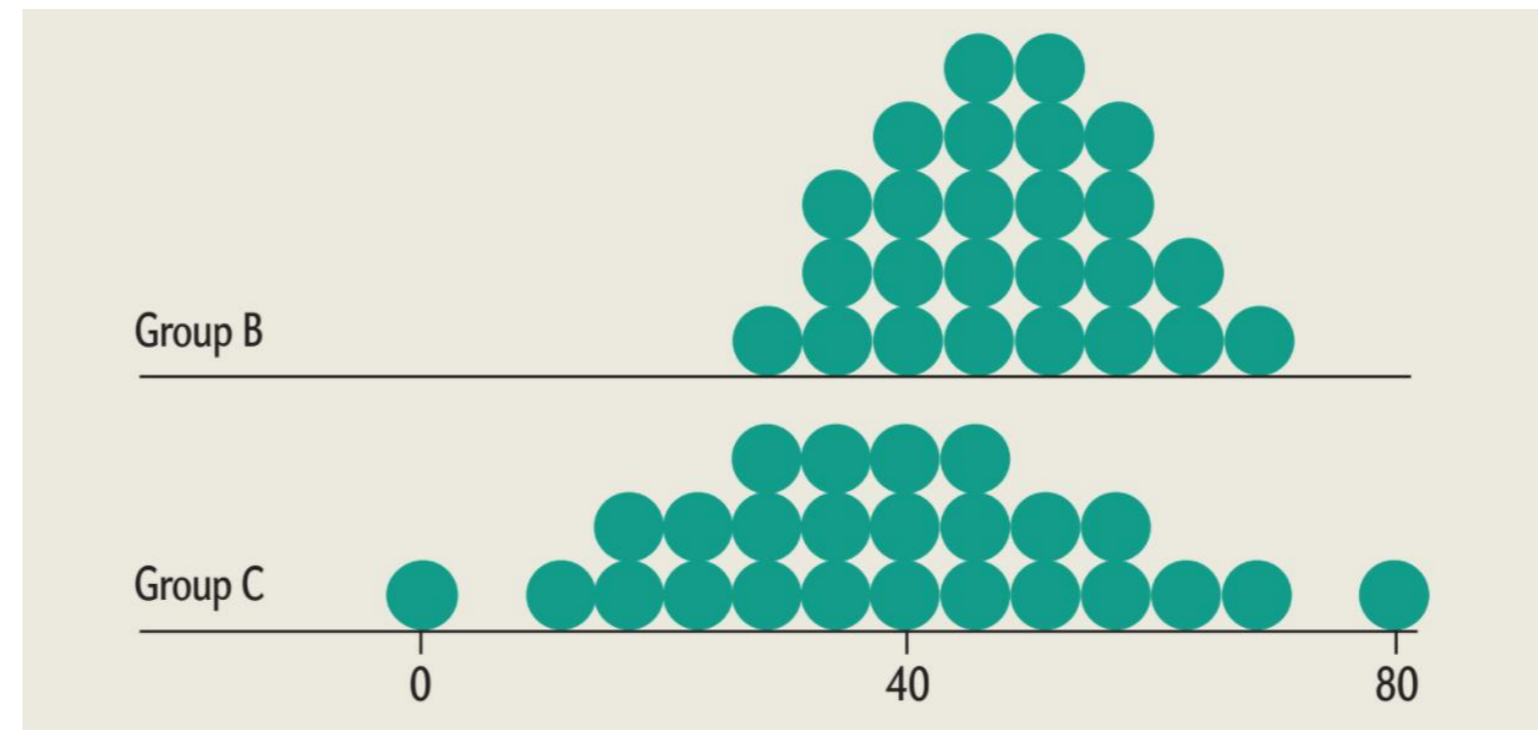
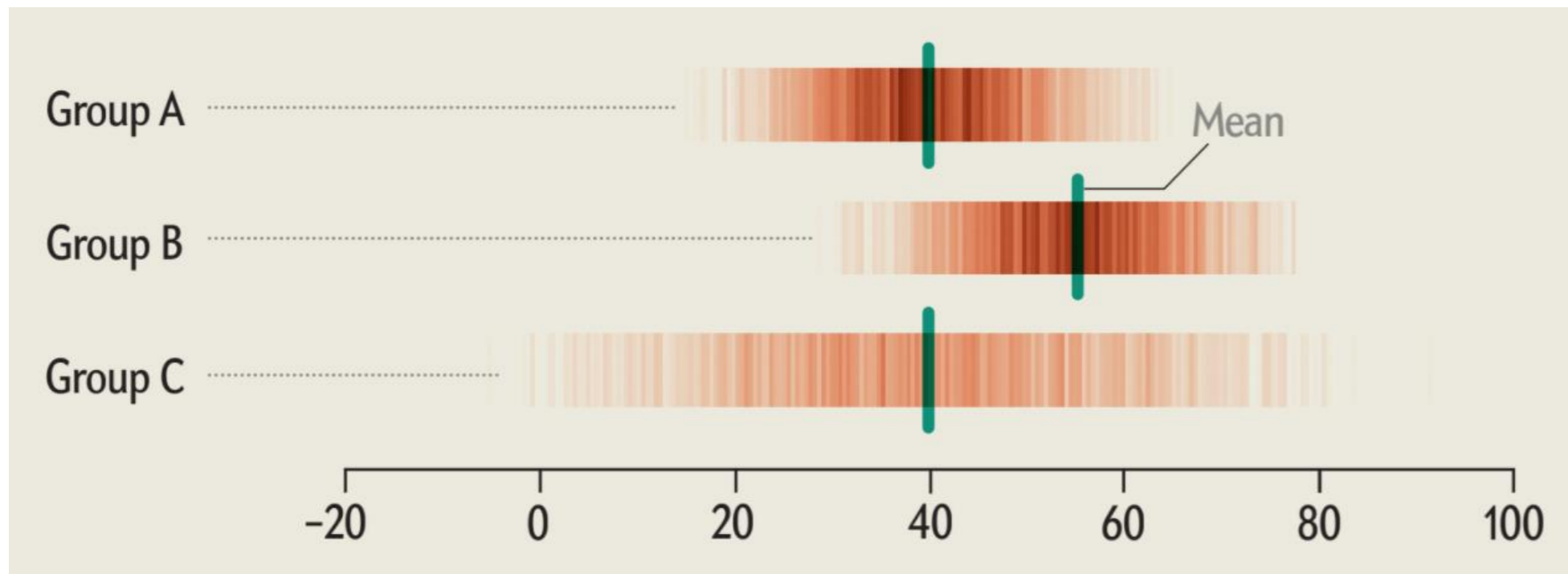
With aspirin



Ottley et al. 2015.

Galesic et al. 2009.

Use continuous frequency formats over error bars



Curate sets of scenarios to convey distribution

“This is something we see a lot”

“This is something we sometimes see”

“This is something we see rarely (but with large implications for decisions)”

Tailor to information needs but anticipate heuristics

Not all users are alike in their information needs/aptitude → Provide info at different levels of granularity with different emphasis.

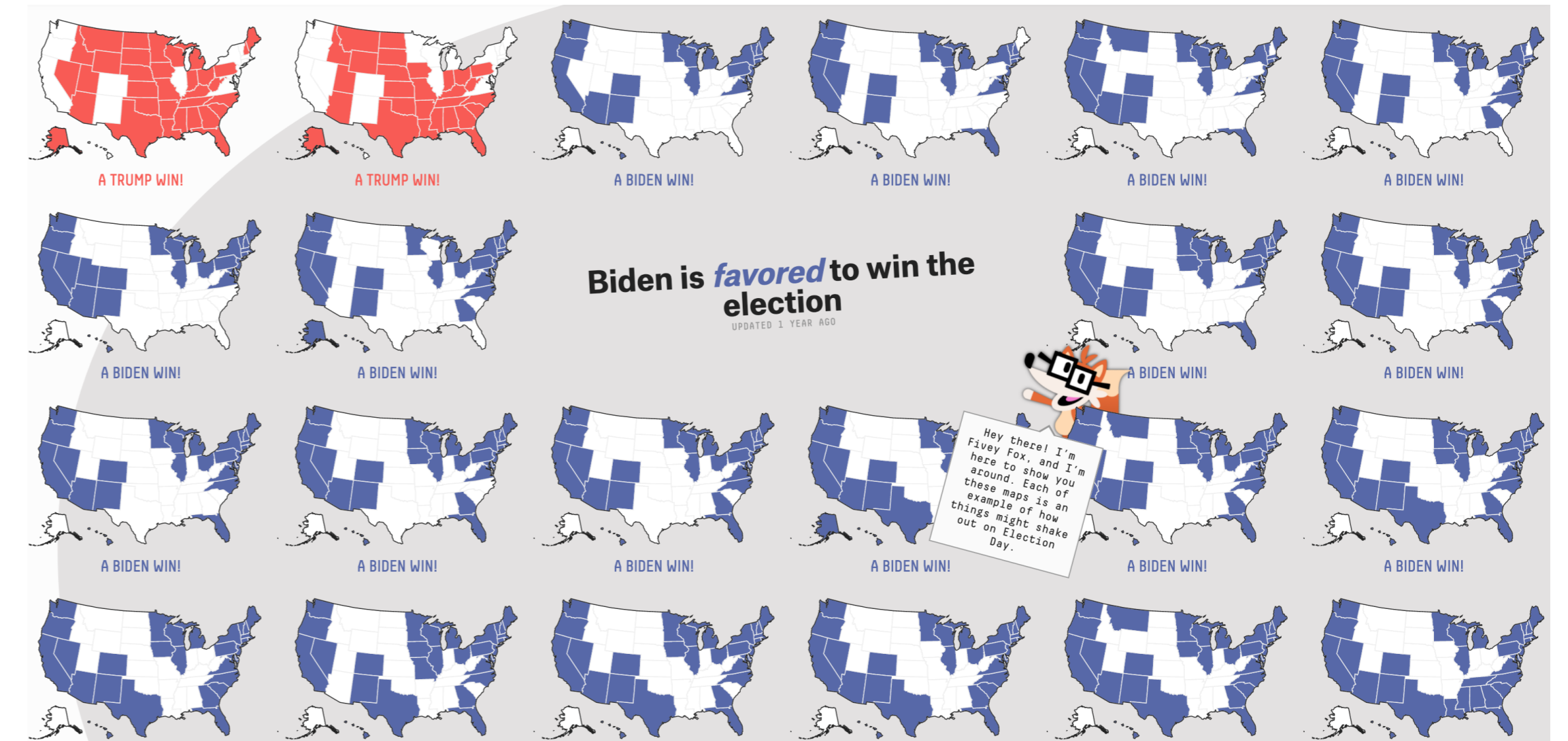
But: Expect users to tend to suppress uncertainty information at all levels.

Who will win the presidency?

Chance of winning



2016



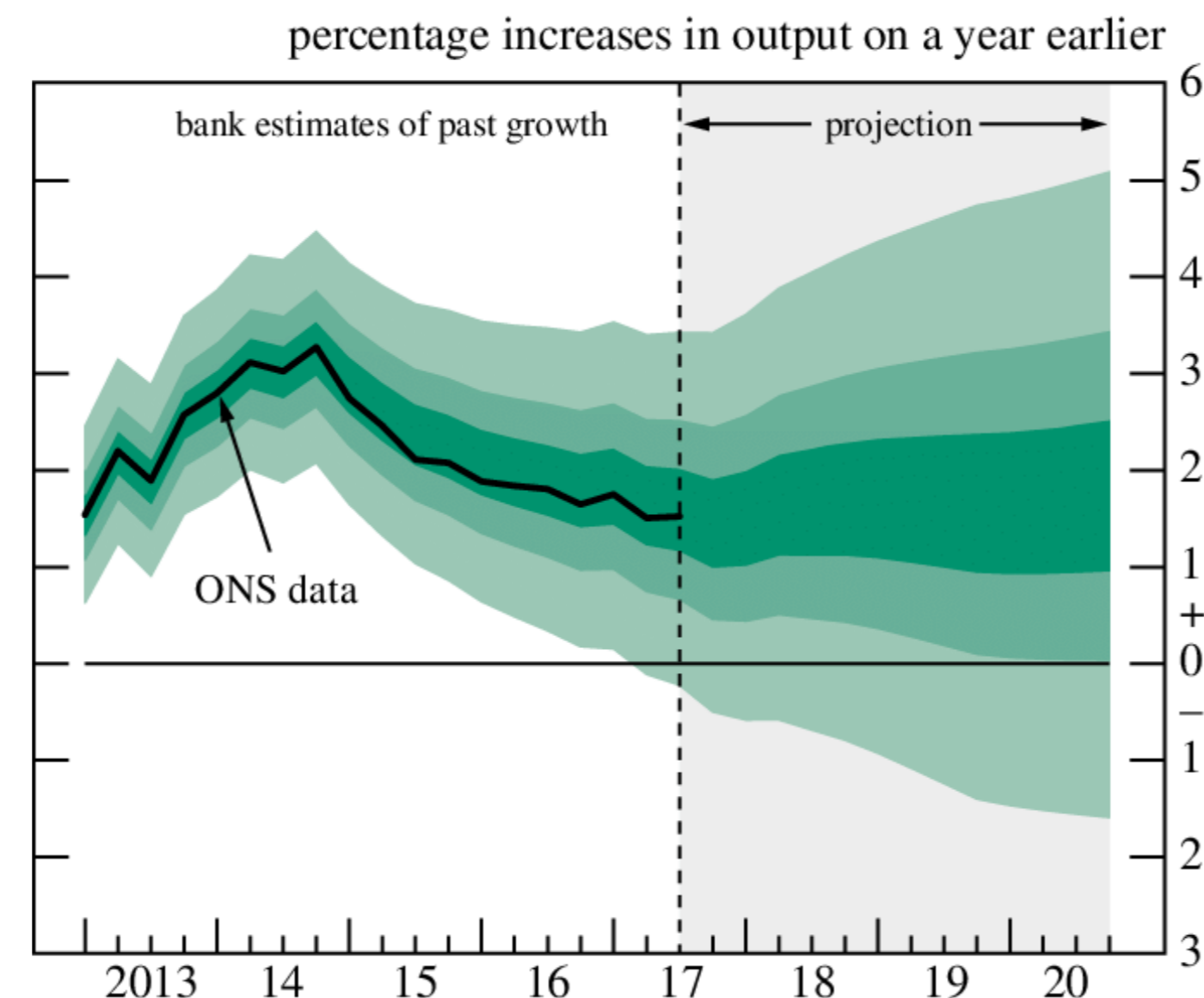
2020

FiveThirtyEight top-level forecasts

Explicitly acknowledge transitory uncertainty

Project uncertainty to past data (e.g., BEA), propagate past uncertainty to future forecasts.

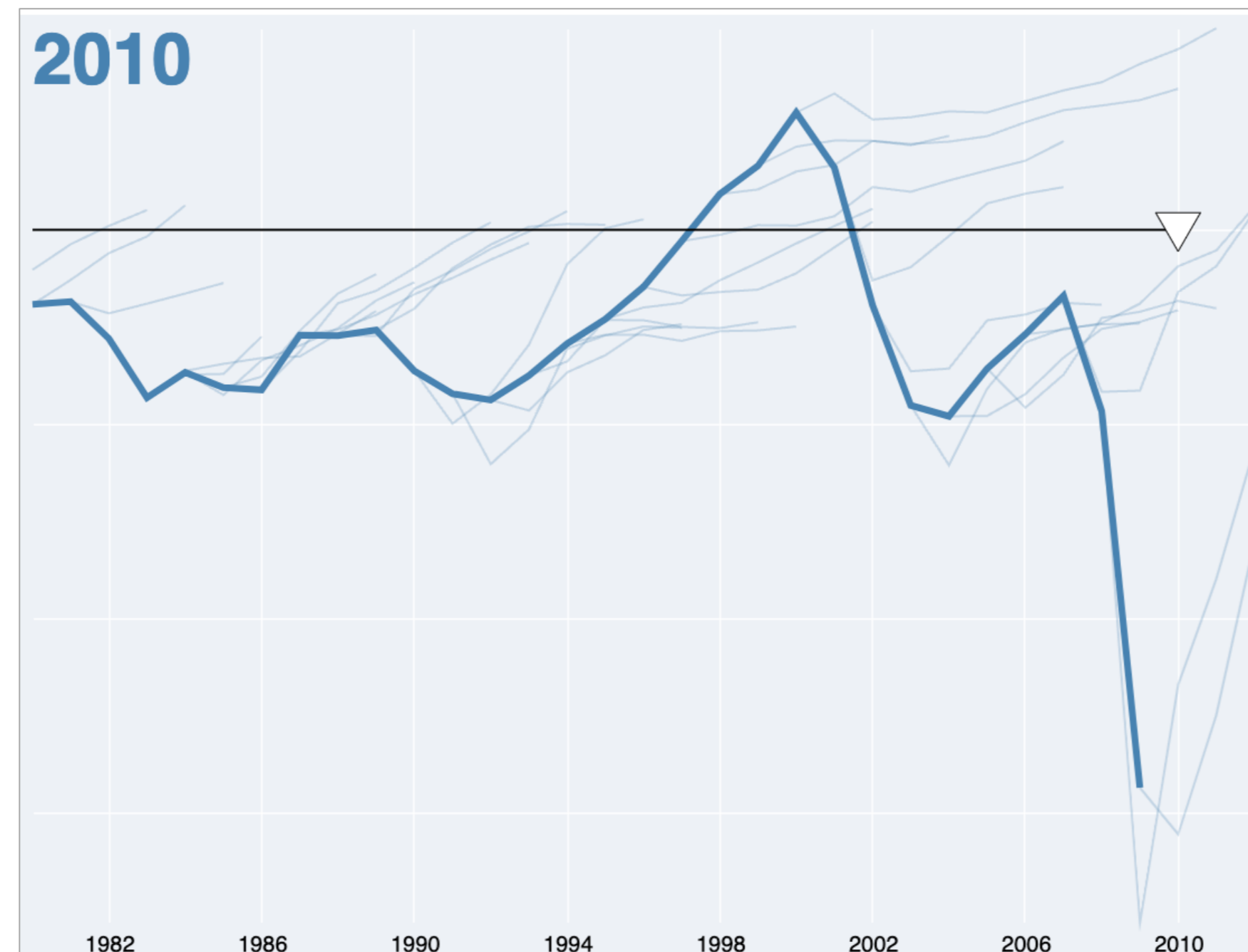
Standardize a data quality or completeness scoring system



Bank of England Fan Chart, 2017.

Provide past error & calibration info by default

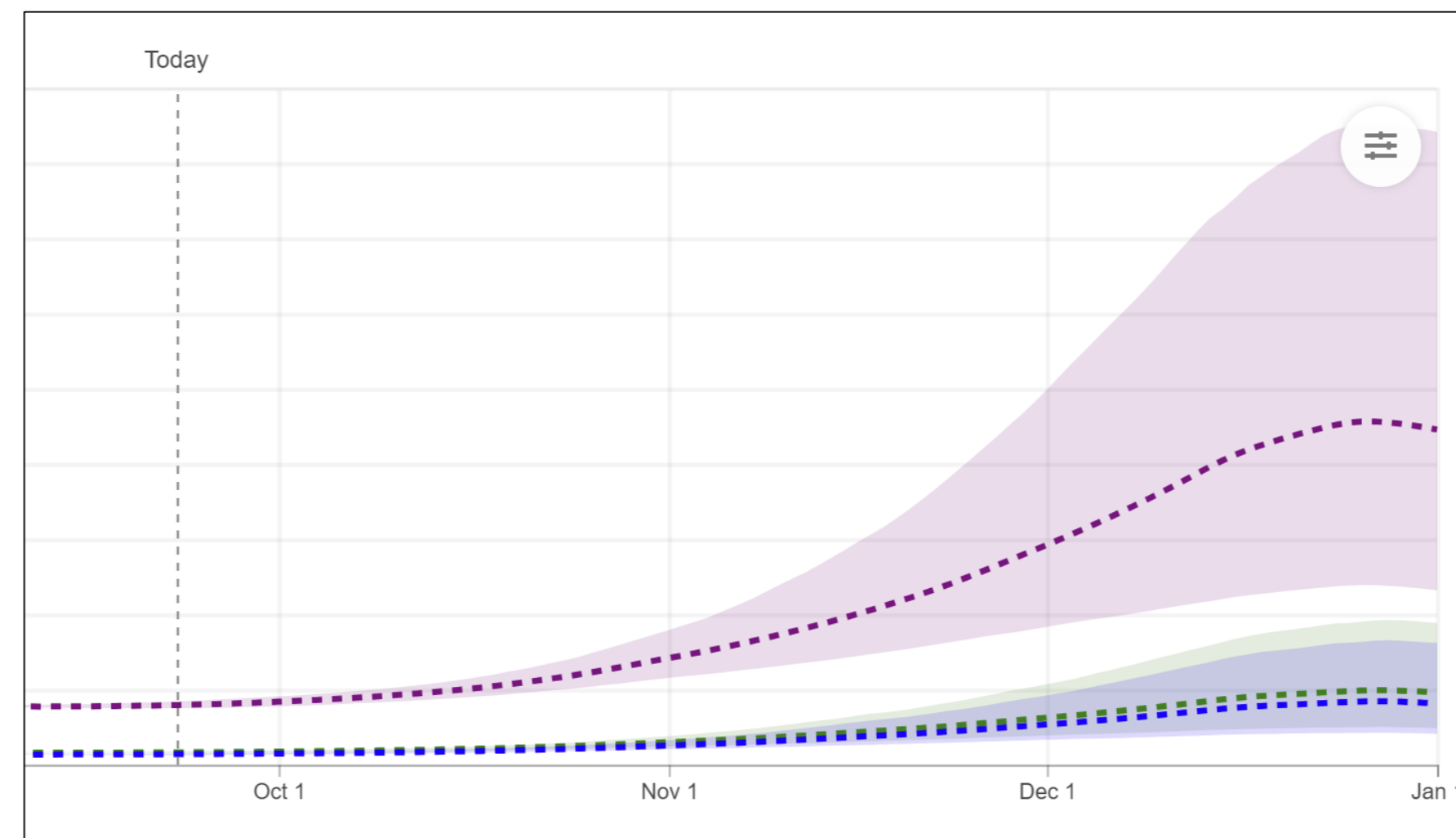
Accompany all new forecasts with prior prediction error in the same units/outcome spaces as the predictions.



Adaption of Cox (2012) NYTimes graphic.

Label partial expressions (i.e., of risk) incomplete

E.g., SEIR models that drove COVID-19 policy ignore behavioral, economic outcomes →
Label forecasts as hypothetical experiment results



UWashington IHME 2020

Report and/or acknowledge all expected forms of error (e.g., non-response and other non-sampling error), not just one type.

Back to Census: Transition to releasing “noisy” counts

What if government data looked imprecise?

Providing obviously imperfect measurements by default avoids the veneer of certitude, normalizing error, and will enable analysts to account for noise in inference.

There's a simple solution to the latest census fight

Trust is important for ensuring participation in all future censuses, and confidentiality of responses is essential to maintaining trust.

By [Cynthia Dwork](#), [Ruth Greenwood](#), and [Gary King](#) Updated July 26, 2021, 3:00 a.m.



What is at stake: Trust in science

For trust in data-driven estimates and science, government communication must convey sources of uncertainty.

The legitimacy of government institutions is at stake.

See also:

Manski. The lure of incredible certitude.

Manski. Communicating uncertainty in policy analysis

Manski. Forming covid-19 policy under uncertainty.

boyd and Sarathy. Differential perspectives: Epistemic disconnects surrounding the
US Census Bureau's use of differential privacy

Jessica Hullman

 jhullman@northwestern.edu

users.eecs.northwestern.edu/~jhullman

mucollective.northwestern.edu

 @JessicaHullman