



Institute
and Faculty
of Actuaries

IFoA Life Conference

**Avoiding pitfalls in setting longevity assumptions:
going beyond technical proficiency**

Agenda

01 Why discuss these issues?

02 Behavioural biases

03 Data considerations

04 Default model calibrations

05 Achieving better outcomes

Introduction

“It’s tough to make predictions, especially about the future”

“It’s especially tough if you’re a human with all sorts of biases”

“If you don’t know where you’re going, you’ll end up someplace else”

Behavioural biases

Instinctive reactions can lead to erroneous conclusions

A famous example



A bat and ball cost \$1.10 in total

The bat costs \$1 more than the ball

How much does the ball cost?

Instinctive answer: 10 cents

Correct answer: 5 cents

Availability heuristic

Decision makers assess the frequency of a class or the probability of an event by the ease with which instances or occurrences can be brought to mind.



*“Will this nightmare
ever end”*

“NHS in crisis”

*“You must stay
at home”*

“Lockdown Britain”

Seeing a high mortality event in the news may make us more pessimistic about future mortality.

Herd mentality

When people conform to the actions, beliefs, or decisions of a larger group.



Busier restaurant

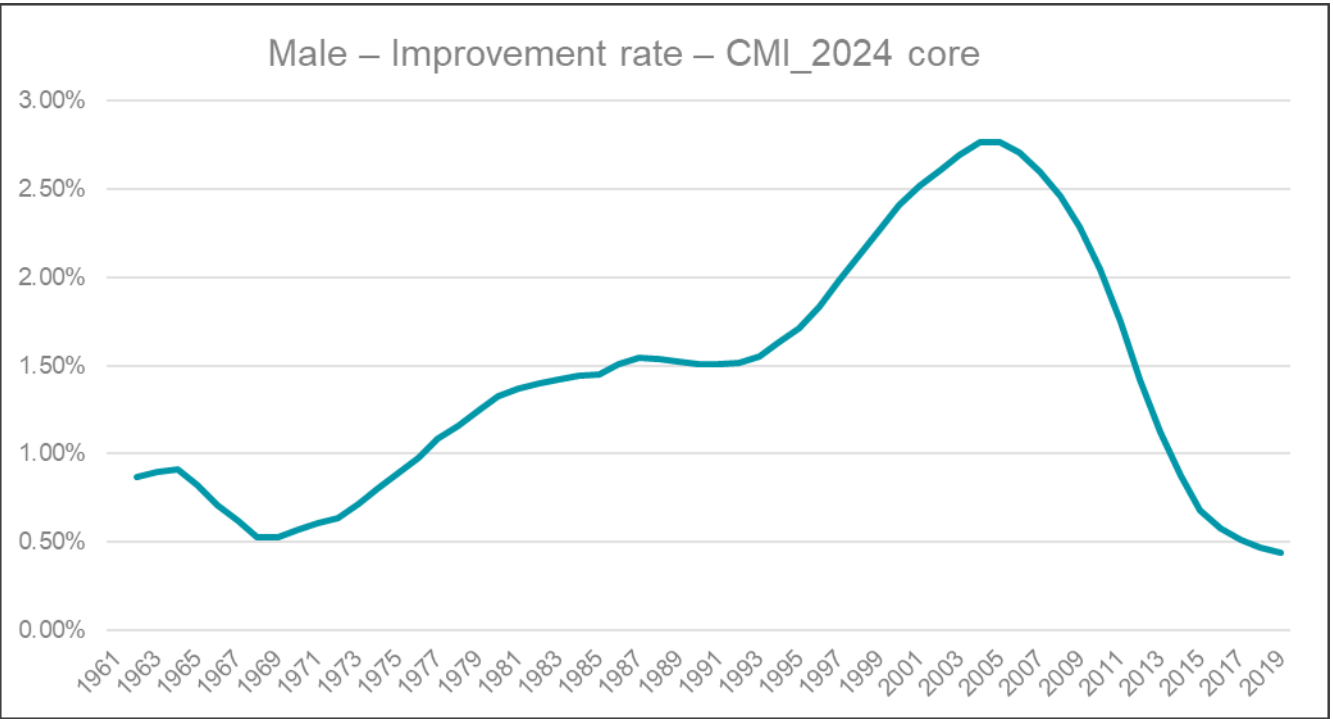
OR



Quieter restaurant

Hindsight bias

Hindsight bias is an implicit cognitive bias that occurs when people overestimate their ability to predict outcomes, after the fact.



Fitted age standardised mortality rates taken from CMI_2024 fitted between 1961 and 2024, with all other parameters set to the core parameters

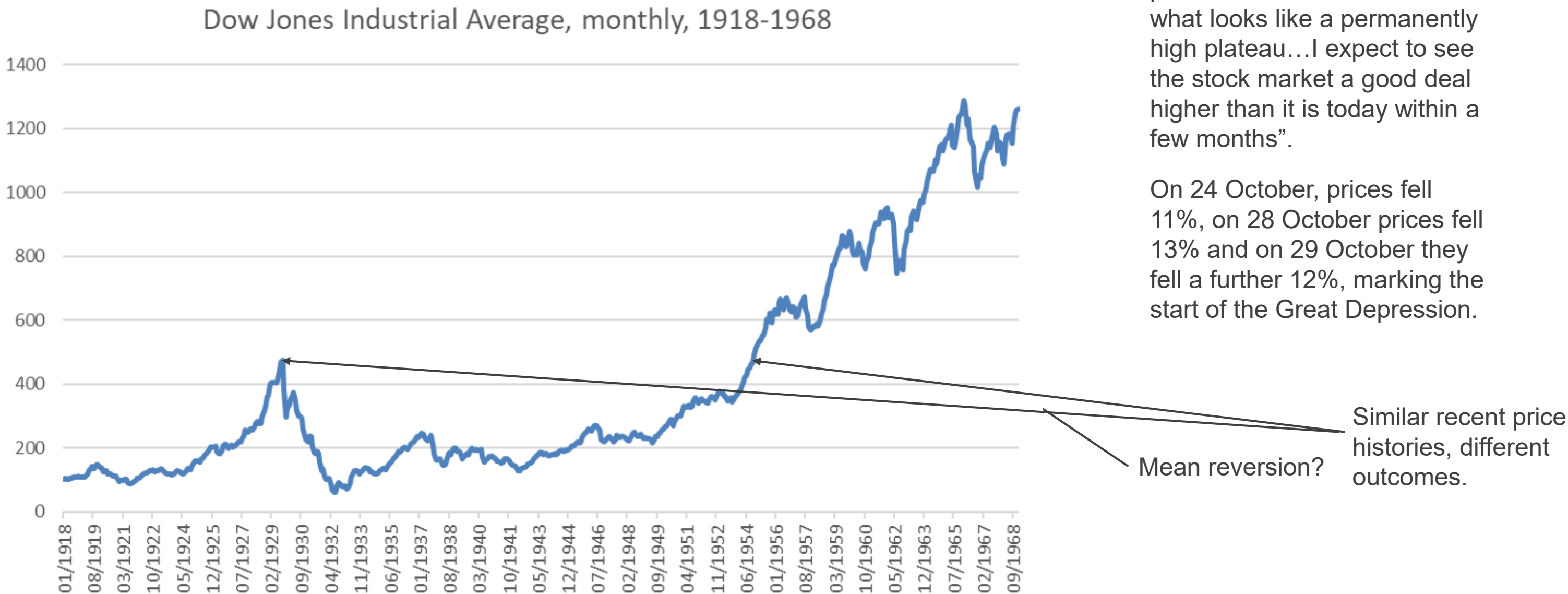
“The high improvements in the 2000s were bound to come to an end.”

CMI working paper 39, July 2009

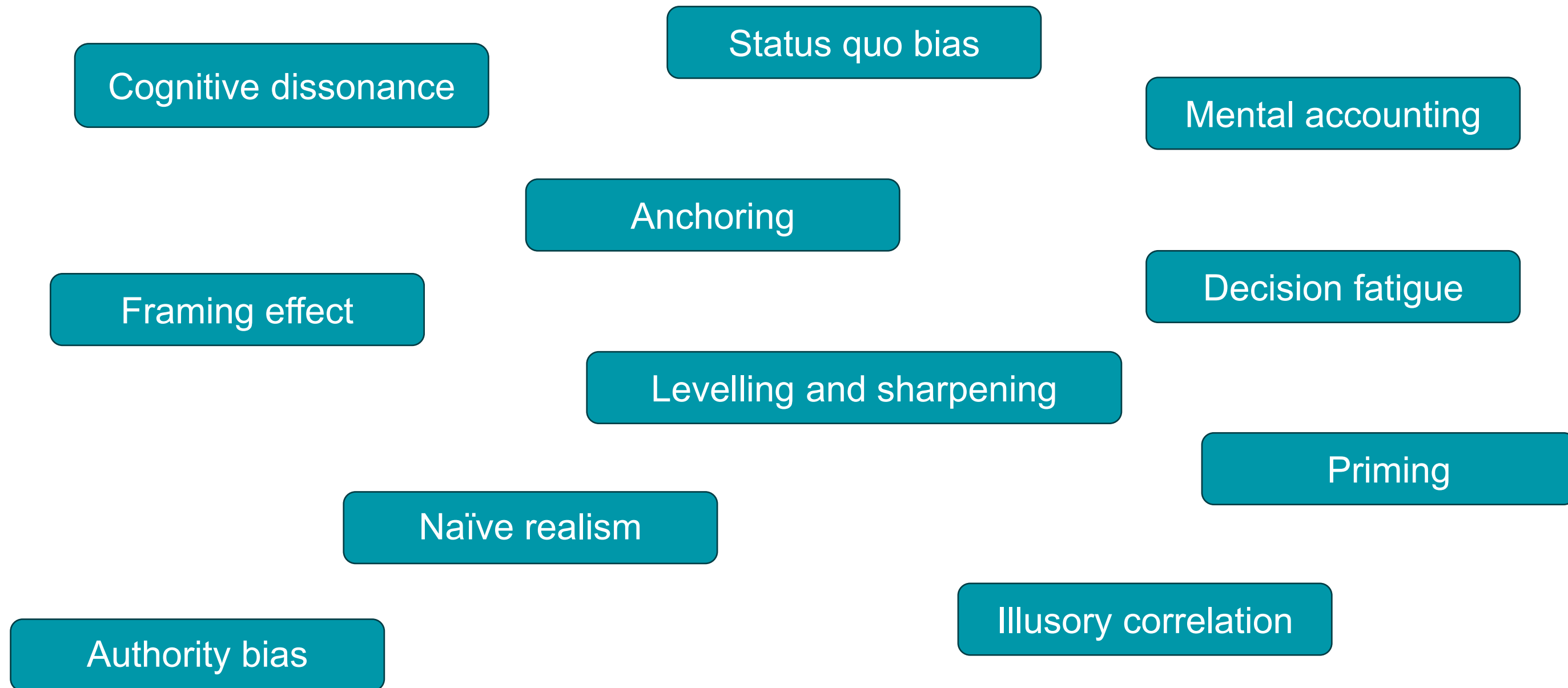
This recognises that Age/Period features in mortality improvements for younger adults have tended to change from decade to decade. However, for ages 60 to 80, where a few major causes of death dominate, there is greater scope for longer-running features and there appears to be no clear end in sight yet for the high rate of mortality improvements driven by advances in relation to circulatory diseases. For the oldest ages it seems appropriate to use a short Period of Convergence, in keeping with the general concept of mortality improvement rates running to zero relatively quickly after age 100.

Hindsight bias

It's easy when you already know the outcome

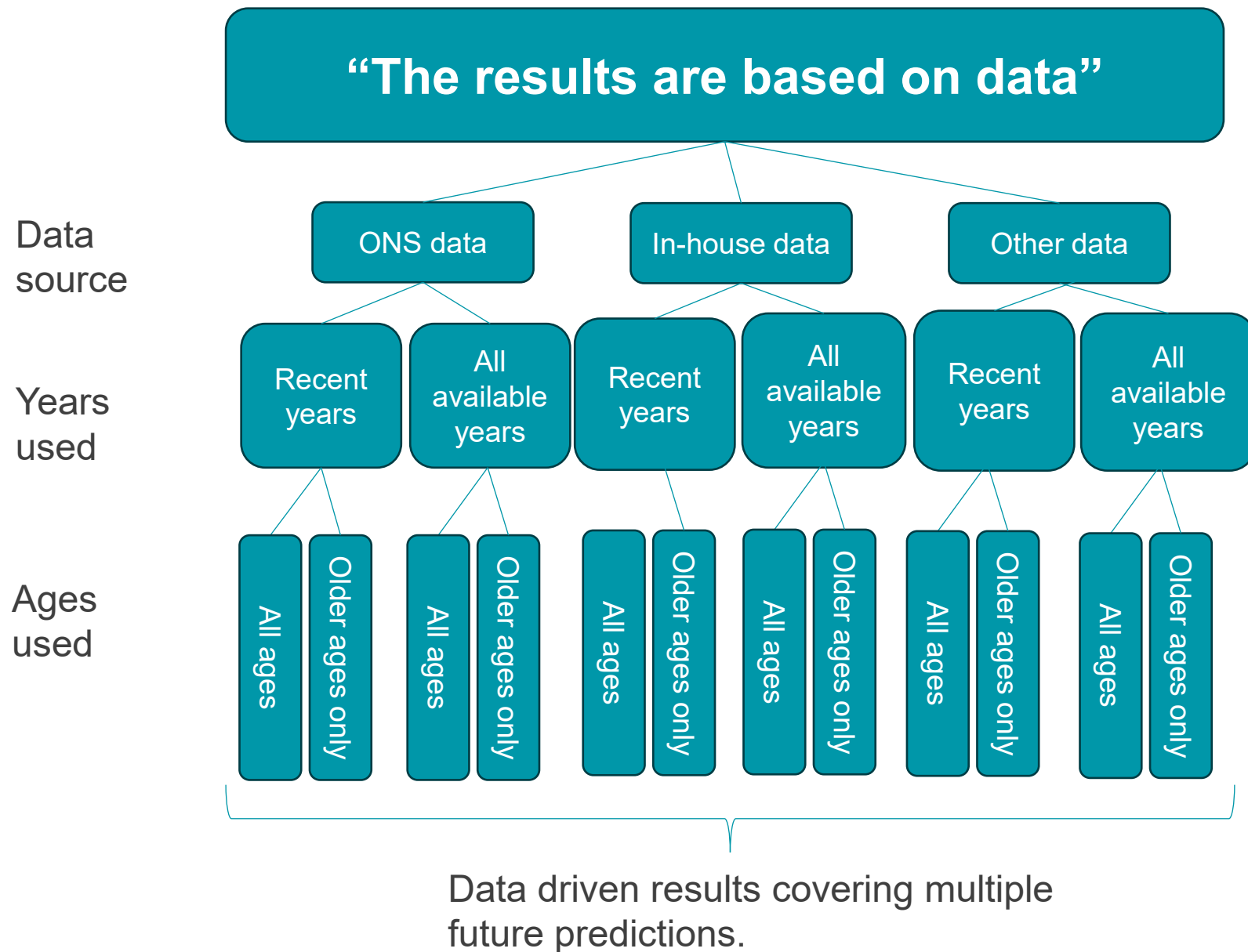


Other biases



Data is not truth

Data is fallible – approach with caution

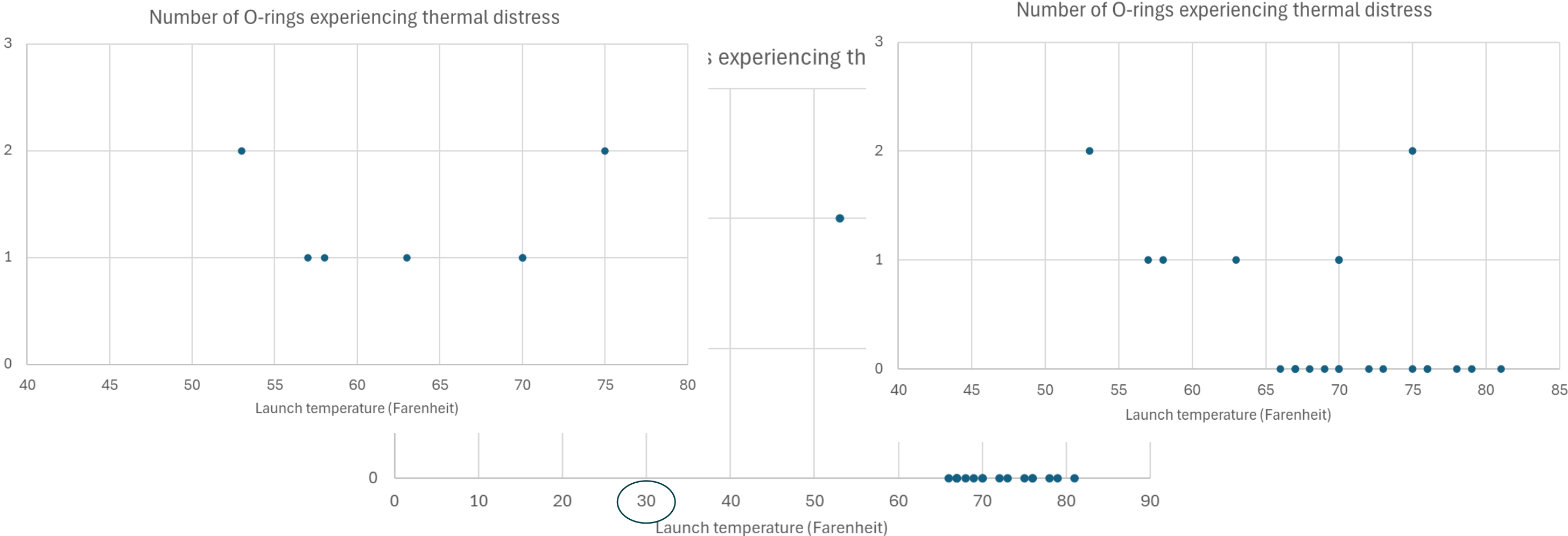


Is there really one ‘source of truth’ with data?

- There can be multiple population data sources (e.g. in the US.)
- A fixed window that rolls forward every year: what is added, what drops out?
- Cause of death data is only as reliable as the categorisation and diagnosis.

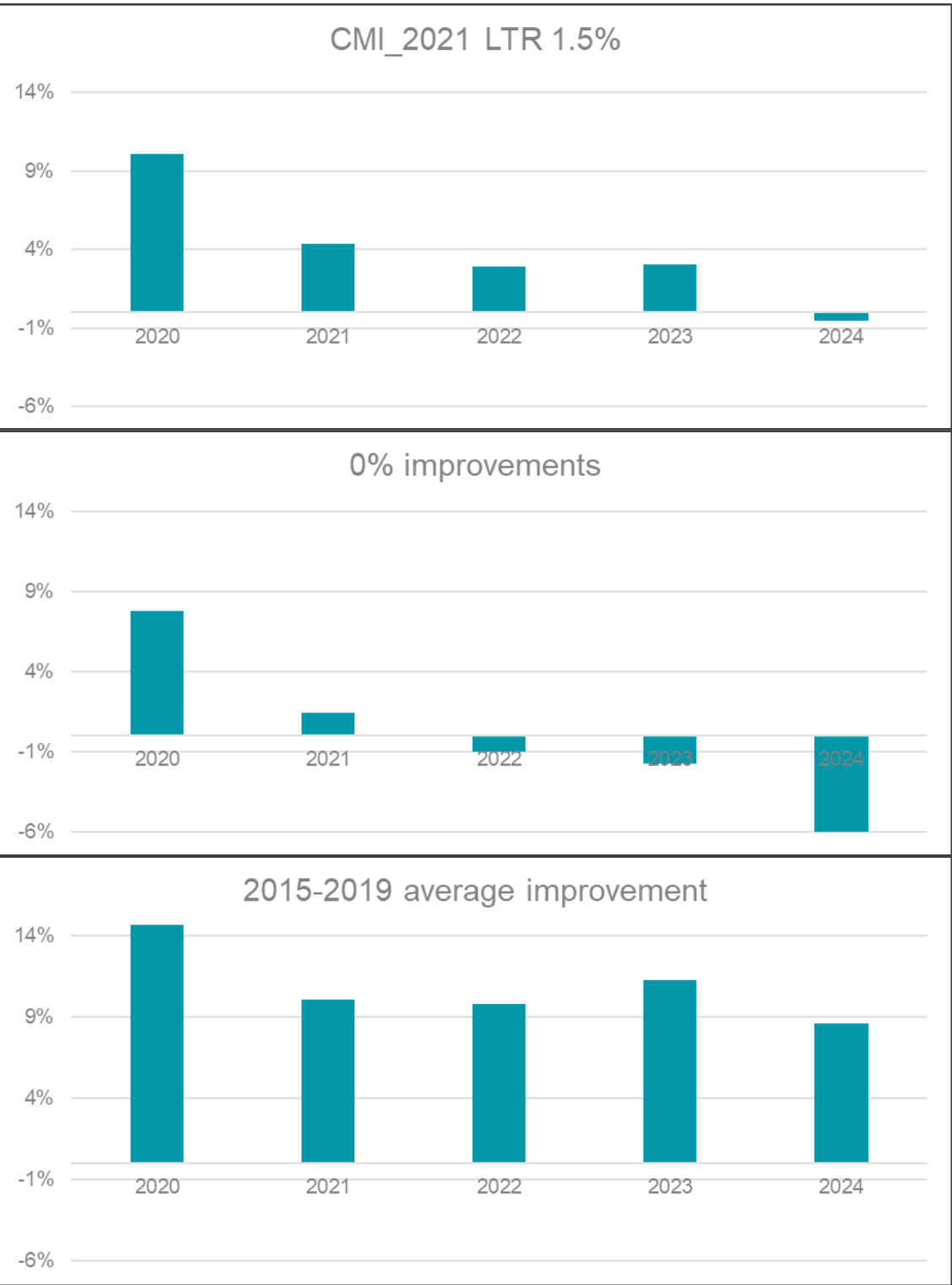
What you don't see can be fatal

What could be missing?



Deciding what information is relevant prior to looking at the whole picture should be avoided

Blurring the objective and subjective



UK female population data, base qx is the 2015-2019 average with a base year of 2017

“Excess deaths are **close to zero** in **2024** in England and Wales”

“Excess deaths are **negative** in **2024** in England and Wales”

“Excess deaths are **high** in **2024** in England and Wales”

Default model calibrations

Default models and industry engagement

Ensure someone else isn't making key decisions for you

We all know there are many advantages to having industry models, benchmarking surveys and default calibrations.

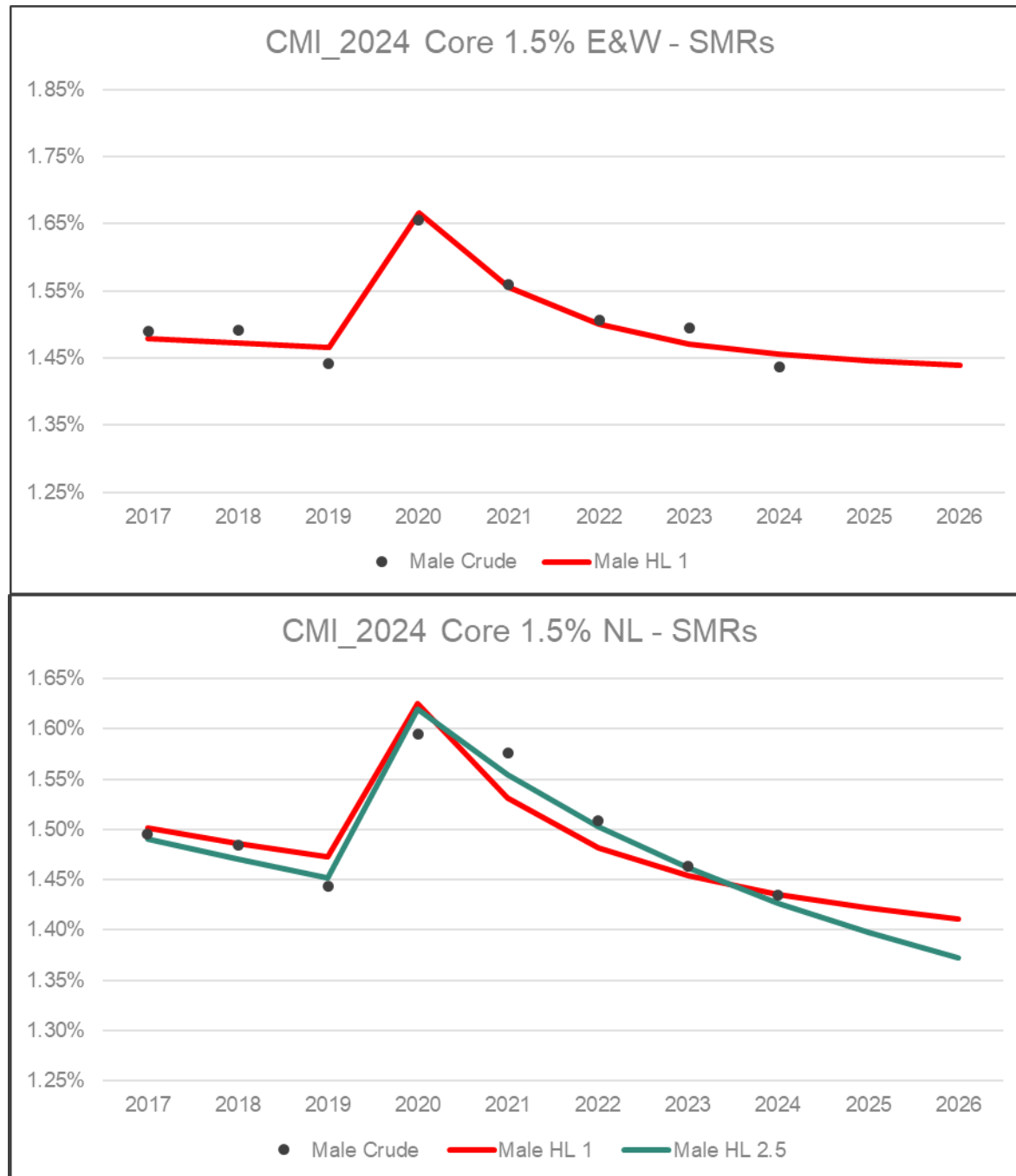
- Source of external benchmarking
- Common language when describing improvements
- External view to challenge internal viewpoints
- Any many more

But do we spend enough time thinking about the risks this can introduce:

- Status quo bias
- Herding
- Anchoring
- Decision fatigue

Default model calibrations

Ensure someone else isn't making key decisions for you



The appropriateness of assumptions can change:

- Over time
- When they are applied to different input data
- When the model output is applied to different lives

How to combat these issues

Taking steps to allow more considered thinking

1. Training and awareness is not sufficient
2. Sensitivities and implications
3. Pre-mortem analyses - try to falsify assumptions
4. Leveraging historical examples
5. Ensuring robust challenge
6. List out key judgements, rationale, alternatives
7. Share rationale and methodology in advance of the results

In the same way we employ reviews and checks to prevent numerical and human errors, we need systems and structures in place to mitigate biases

Q & A



Institute
and Faculty
of Actuaries

Thank you

For more information, contact:

Michael Anderson - michael.anderson@rgare.com

Tom Honeywell - thomas.honeywell@rgare.com

References

- Centraal Bureau voor de Statistiek. Annual death counts by single age (0-107, 108+), year of birth, and sex, various years, Netherlands. (Data obtained through the Human Mortality Database, www.mortality.org, on 10/03/2025)
- Centraal Bureau voor de Statistiek. Weekly death counts by age group, year of birth, and sex, various years, Netherlands. (Data obtained through the Human Mortality Database, www.mortality.org, on 10/03/2025)
- Cognitive Biases, The Decision Lab. Available at: <https://thedecisionlab.com/biases>, last visited 03/10/2025
- Continuous Mortality Investigation (CMI), CMI_2024 (working paper 201), 14/08/2025. Available at: <https://www.actuaries.org.uk/learn-and-develop/continuous-mortality-investigation/cmi-working-papers/mortality-projections/cmi-working-paper-201>, Last visited 01/10/2025
- Continuous Mortality Investigation (CMI), Working paper 39, p63, 06/07/2009. Available at: <https://www.actuaries.org.uk/learn-and-develop/continuous-mortality-investigation/cmi-working-papers/mortality-projections/cmi-wp-38-39>, Last visited 03/10/2025
- Continuous Mortality Investigation (CMI), CMI_2021 (working paper 160), 22/11/2022. Available at: <https://www.actuaries.org.uk/learn-and-develop/continuous-mortality-investigation/cmi-working-papers/mortality-projections/cmi-working-paper-160>, Last visited 03/10/2025
- Draper, D. (1993). Challenger USA Space Shuttle O-Ring Dataset. University of California, Irvine Machine Learning Repository. <https://doi.org/10.24432/C5PW2T>.
- Kahneman, D (2011). Thinking fast and slow. Penguin books.
- National Bureau of Economic Research, Dow-Jones Industrial Stock Price Index for United States [M1109BUSM293NNBR], retrieved from FRED, Federal Reserve Bank of St. Louis; <https://fred.stlouisfed.org/series/M1109BUSM293NNBR>, Last visited 03/10/2025.
- Novel Investor, “A Cautionary Tale of Forecasting“, 21/08/2024. Available at: <https://novelinvestor.com/a-cautionary-tale-of-forecasting/>. Last visited 03/10/2025.
- Office for National Statistics, 2024, Deaths and population estimates for England and Wales, 1961 to 2023, by single year of age, Available at: <https://www.ons.gov.uk/peoplepopulationandcommunity/birthsdeathsandmarriages/lifeexpectancies/adhocs/2517deathsandpopulationestimateforenglandandwales1961to2023bysingleyearofage>, (Data obtained via CMI, in CMI_2024, subject to some transformation and estimation).
- Richardson, G., Komai, A., Gou, M. and Park, D. Stock Market Crash of 1929, Federal Reserve History. 22/11/2013. Available at: <https://www.federalreservehistory.org/essays/stock-market-crash-of-1929#:~:text=After%20prices%20peaked%2C%20economist%20Irving,Dow%20declined%20nearly%2013%20percent..> Last visited: 03/10/2025.