

# Explaining Systematic Bias and Nontransparency in US Social Security Administration Forecasts<sup>1</sup>

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(Hellenic American Bankers Association, 1/28/2015)

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<sup>1</sup>Based on joint work with Konstantin Kashin and Samir Soneji

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# References

# References

- Systematic Bias and Nontransparency in US Social Security Administration Forecasts  
*Journal of Economic Perspectives*

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- Articles, data, code available at [GaryKing.org/ss](http://GaryKing.org/ss)

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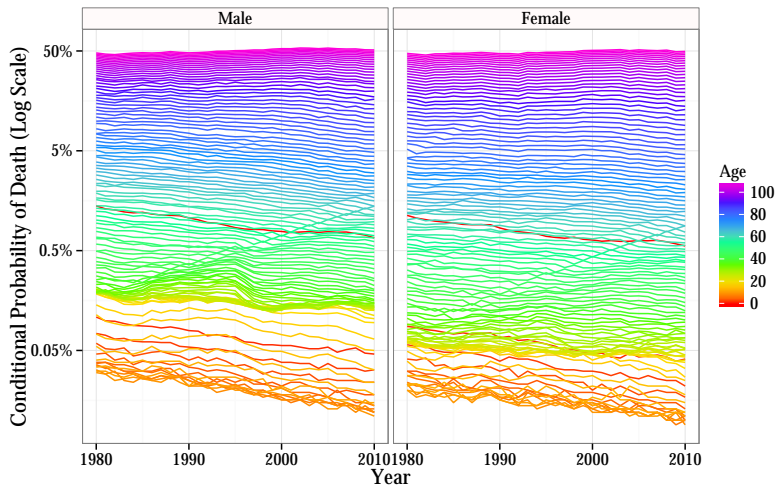
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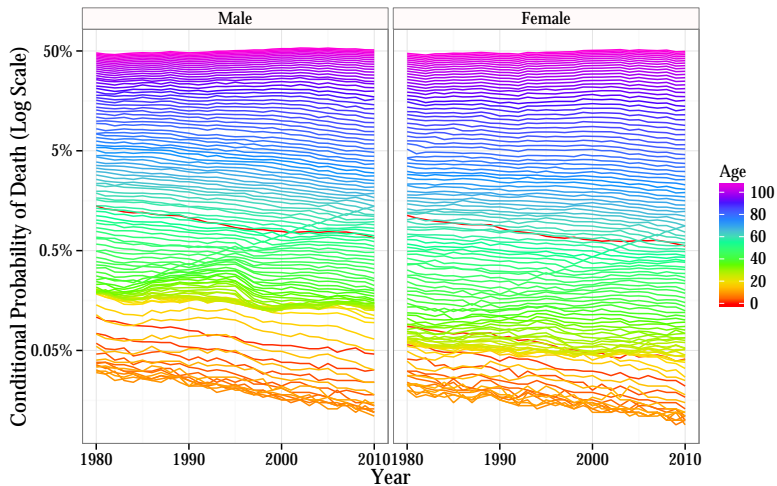
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# Actual Mortality Time Profiles are Complex

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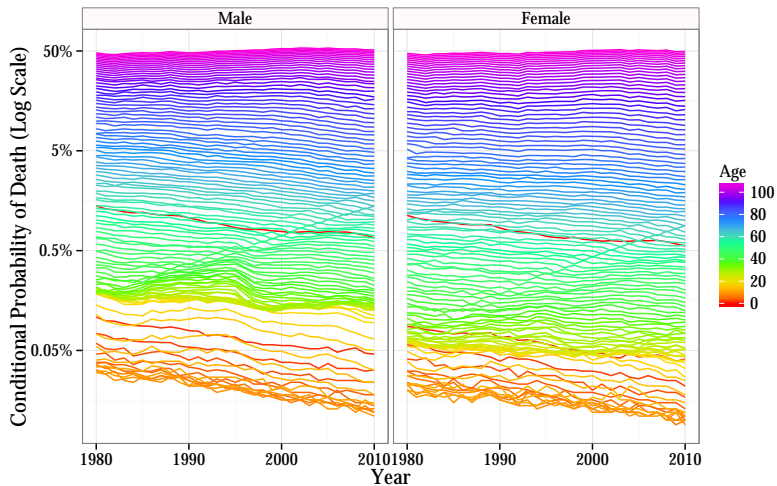


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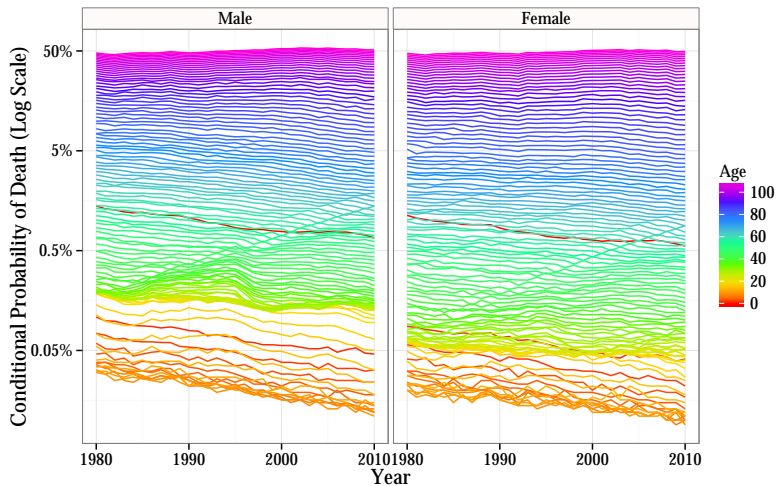
Patterns:

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Patterns:  $\approx$  linear,

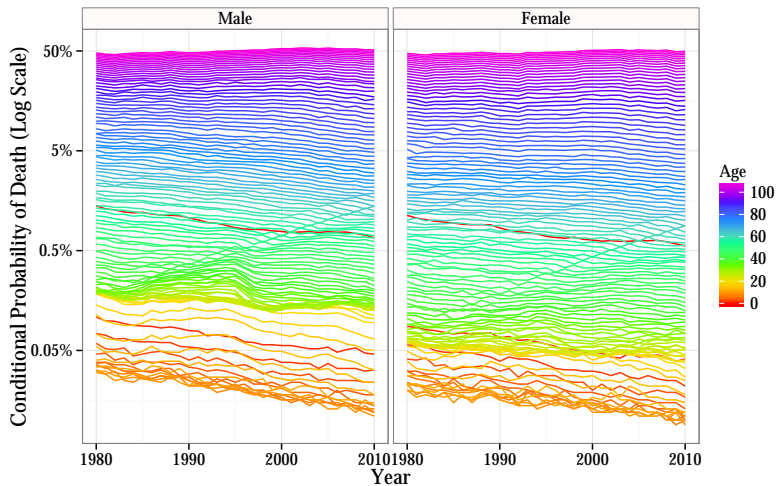
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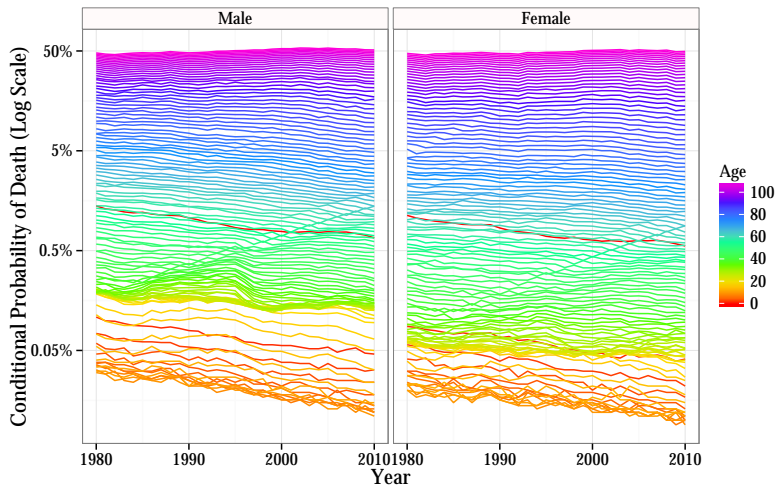


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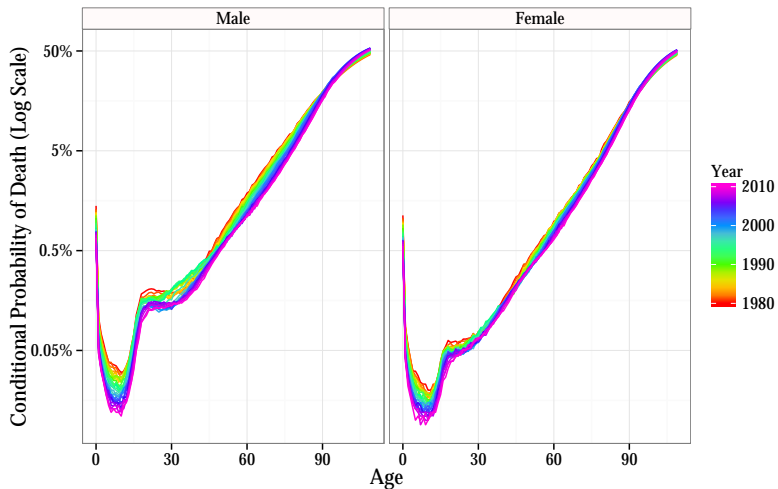
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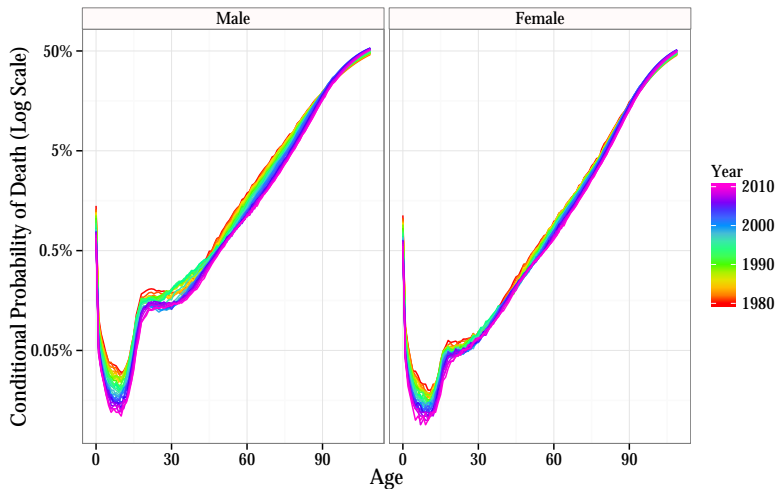
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## Actual Mortality Age Profiles are also Complex

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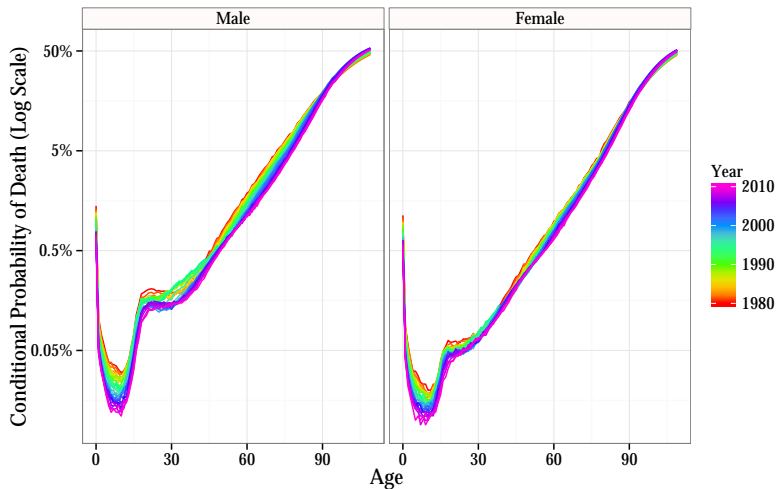


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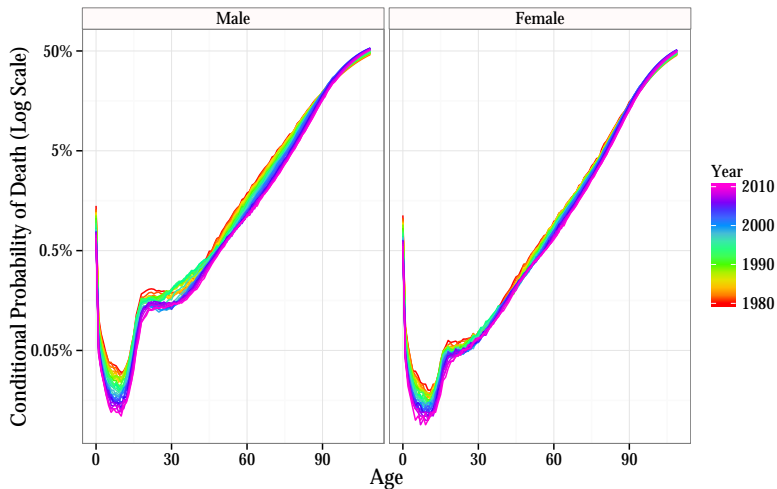
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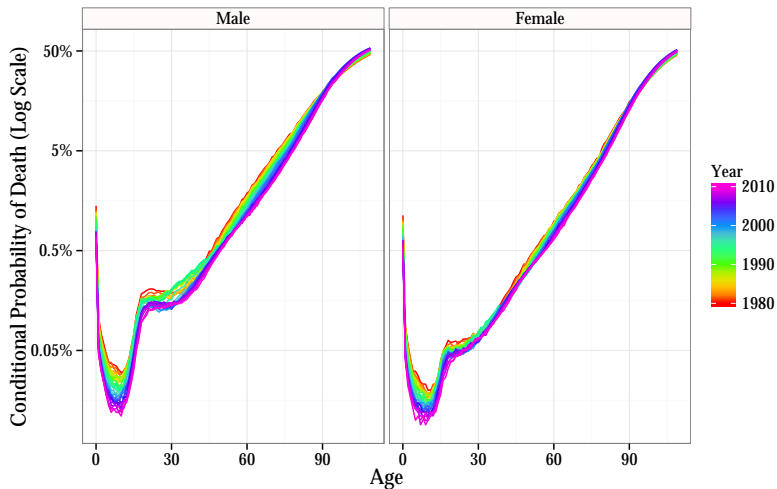
Patterns: Characteristic shape,

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Patterns: Characteristic shape, partly linear,

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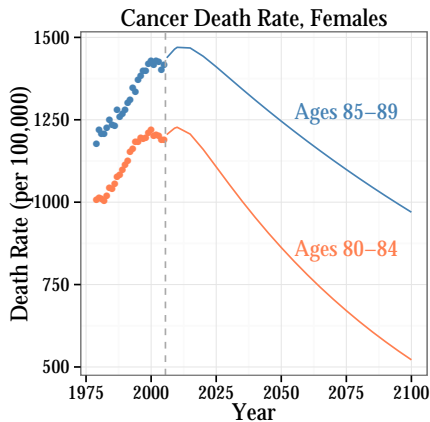
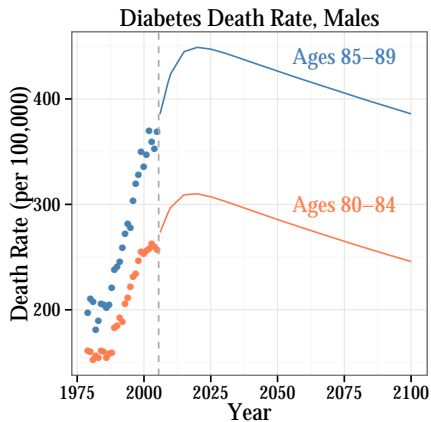
Patterns: Characteristic shape, partly linear, different time  $\times$  age trends



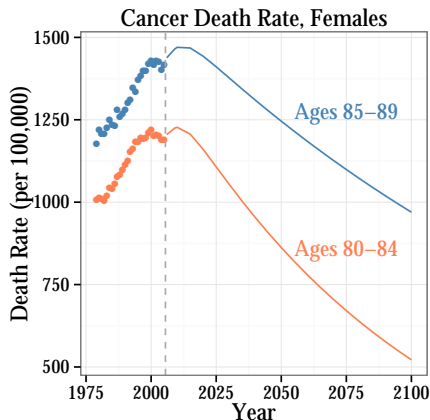
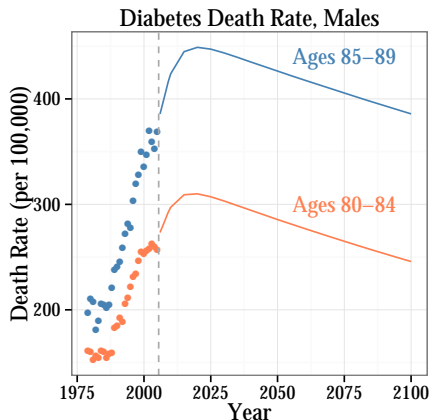
# OCACT Qualitative Choices

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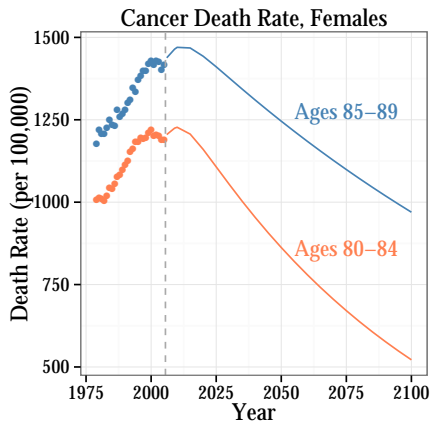
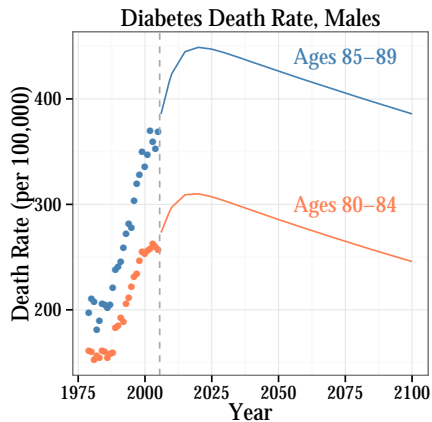


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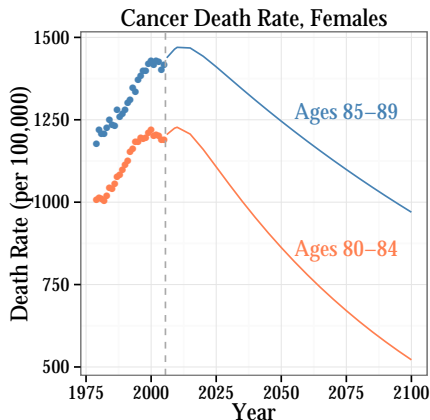
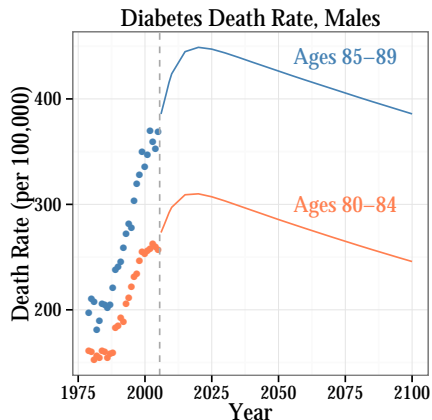
Unrealistic patterns:

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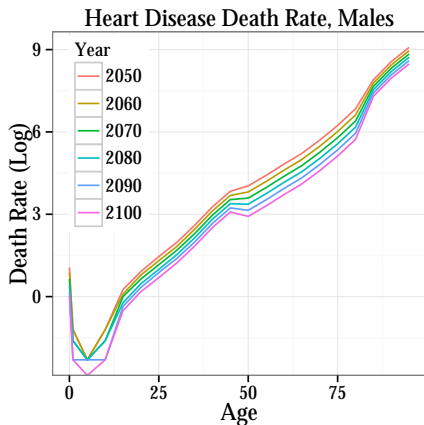
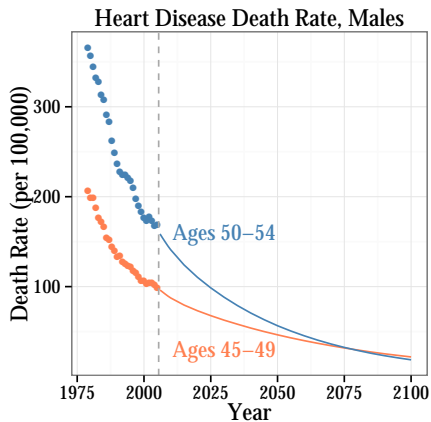
Unrealistic patterns: change of directions,

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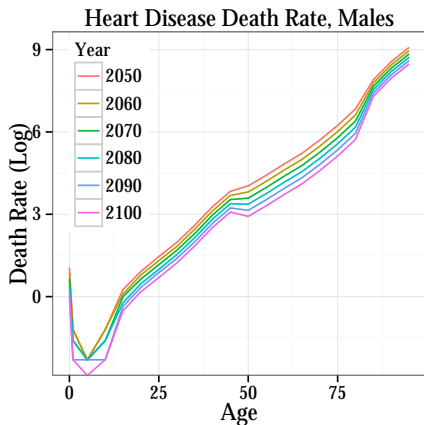
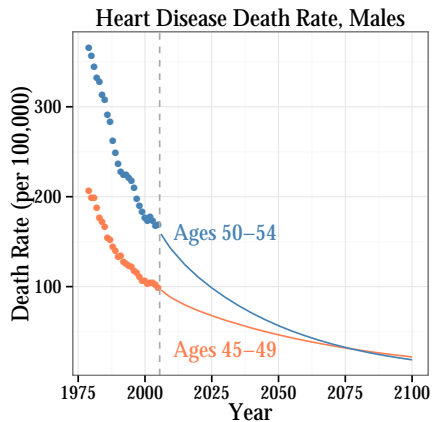


Unrealistic patterns: change of directions, change of differences

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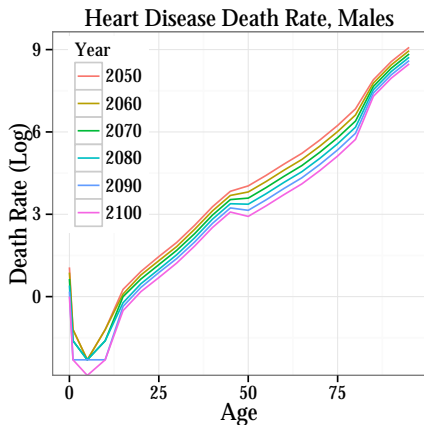
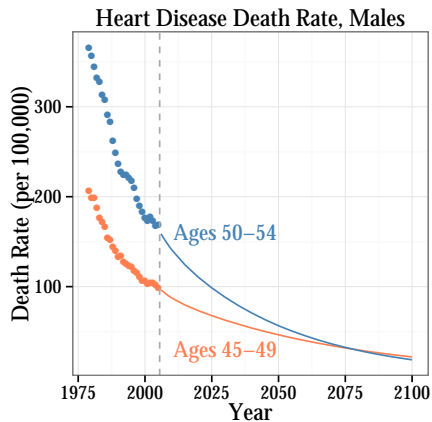
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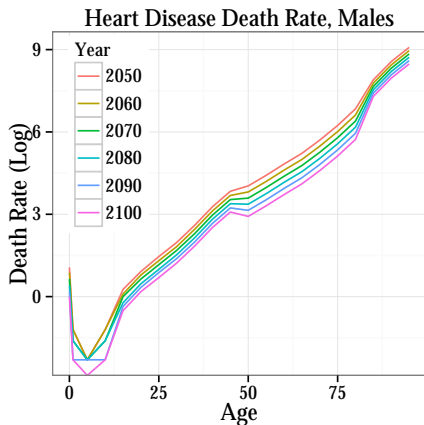
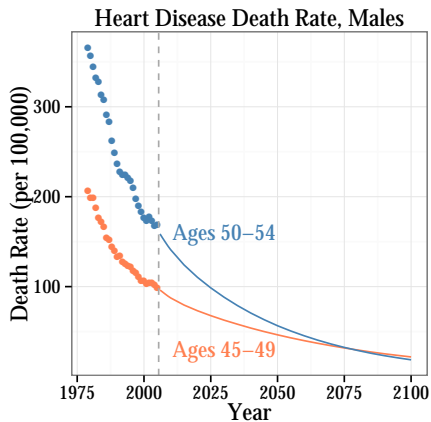


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Unrealistic patterns: Crossing age plots,

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Unrealistic patterns: Crossing age plots, notch for 50-75 year olds

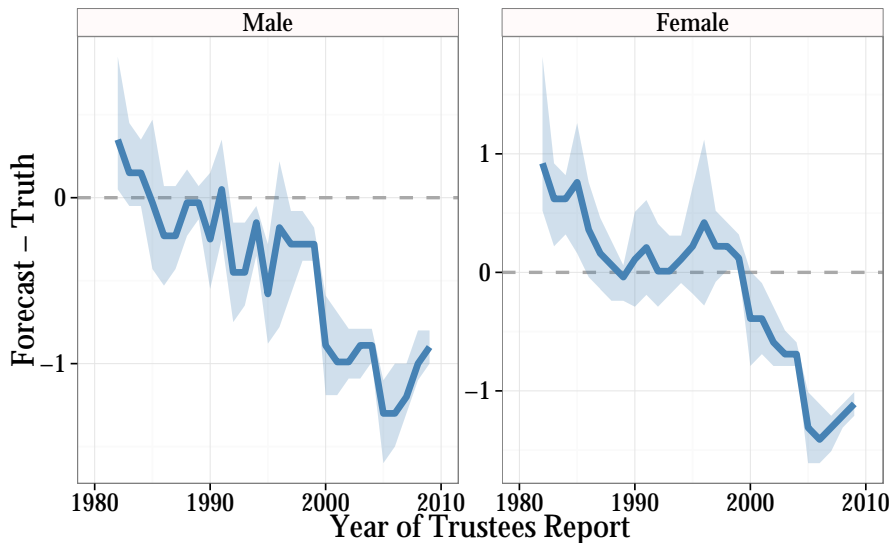
# SSA Life Expectancy Forecasts: Increasing Bias Since 2000

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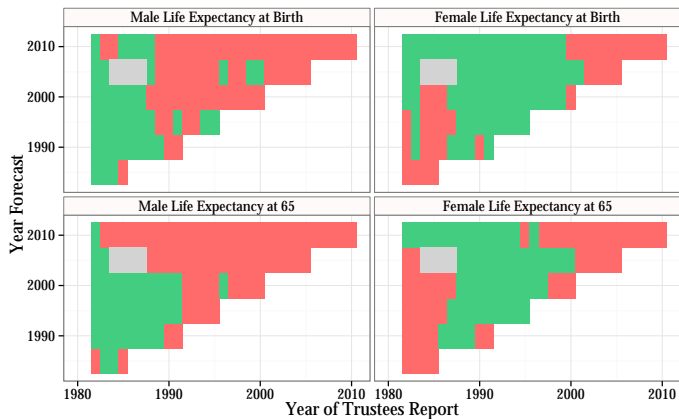
# Life Expectancy “Uncertainty Interval” Coverage

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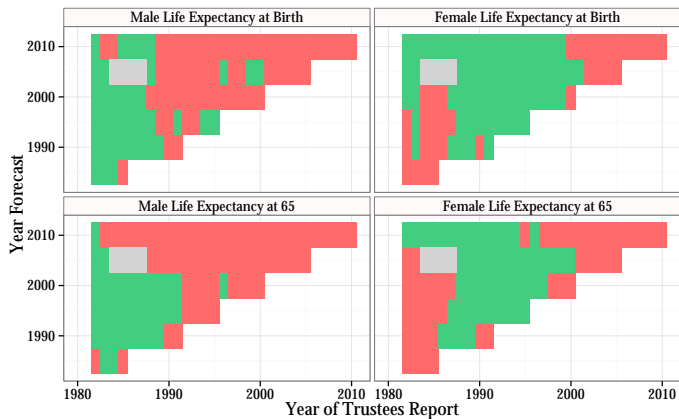
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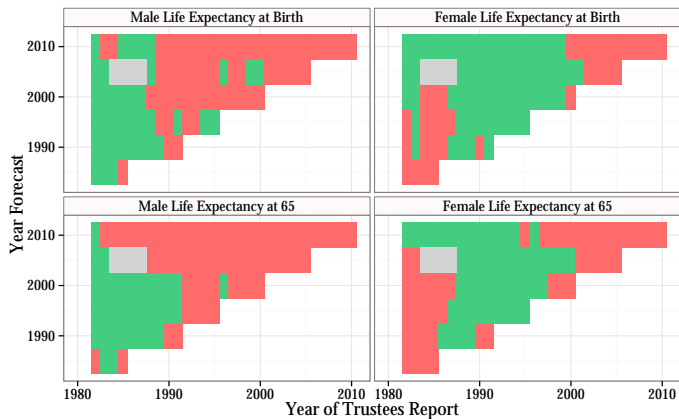
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Patterns:

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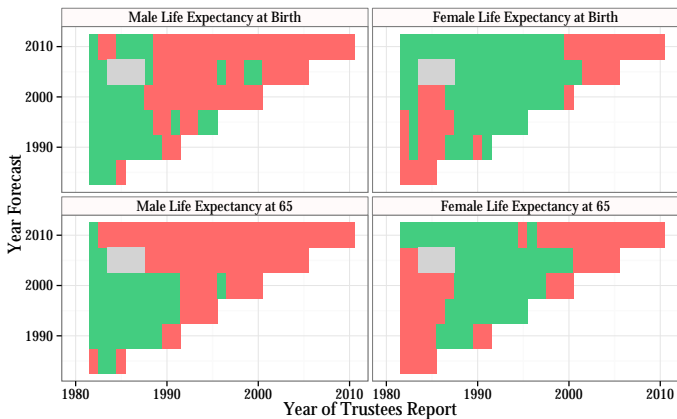


Patterns:

- Vertical: Later Trustees Reports are overconfident

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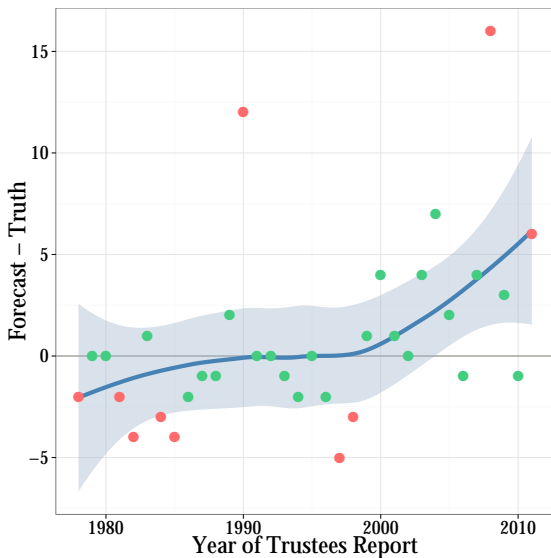


Patterns:

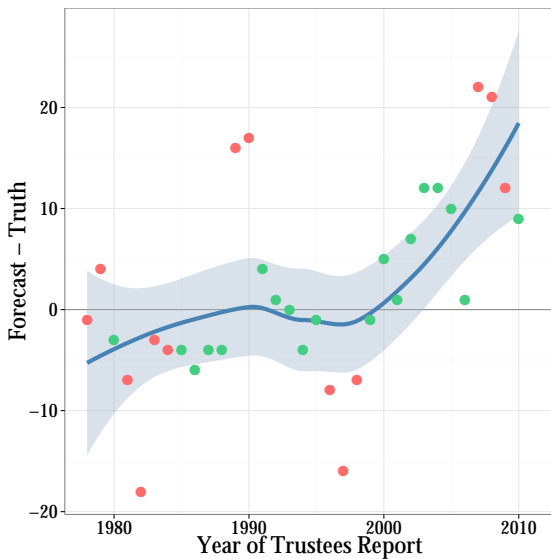
- Vertical: Later Trustees Reports are overconfident
- Not horizontal: Shorter term forecasts should be better, but aren't

# Trust Fund Ratio Forecasting Errors

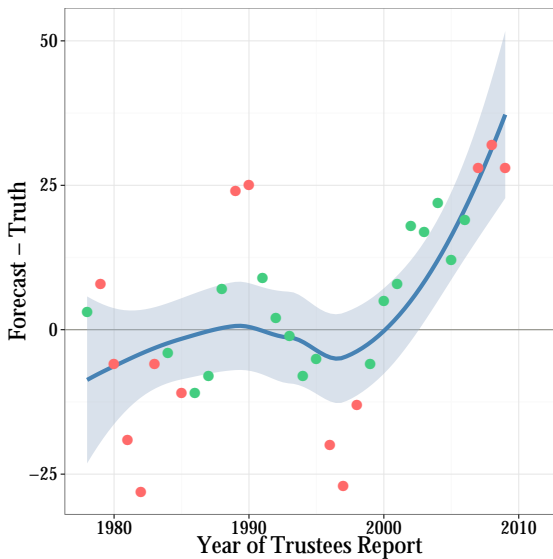
# Trust Fund Ratio Forecasting Errors: 1 Year Ahead



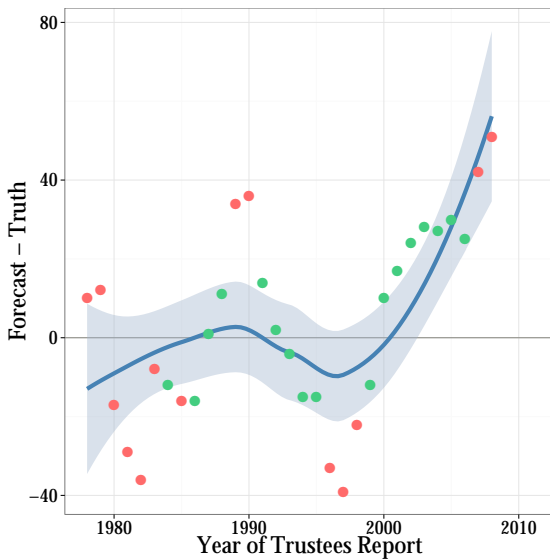
## Trust Fund Ratio Forecasting Errors: 2 Years Ahead



# Trust Fund Ratio Forecasting Errors: 3 Years Ahead

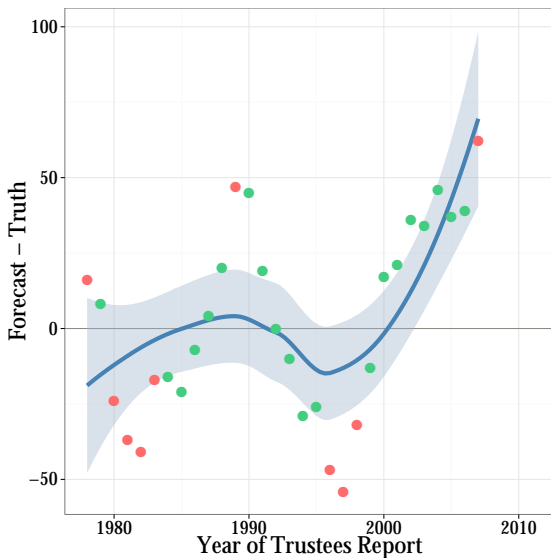


# Trust Fund Ratio Forecasting Errors: 4 Years Ahead

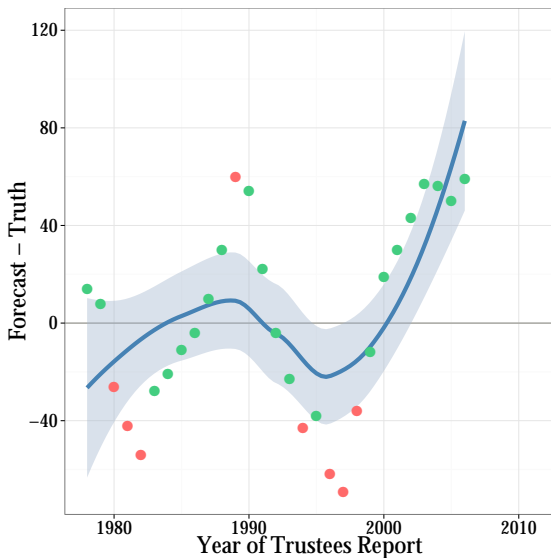




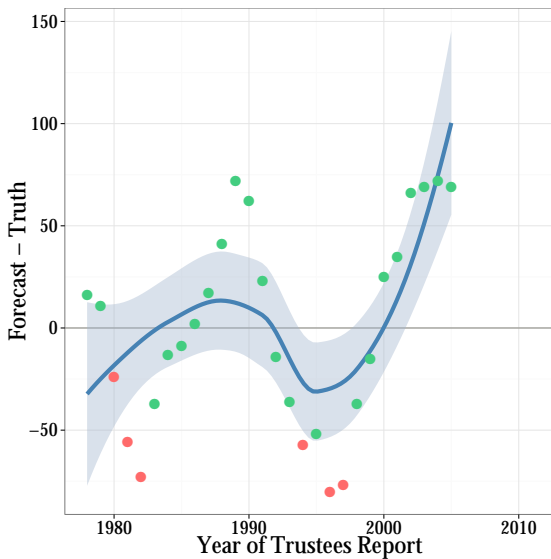
# Trust Fund Ratio Forecasting Errors: 5 Years Ahead



# Trust Fund Ratio Forecasting Errors: 6 Years Ahead



# Trust Fund Ratio Forecasting Errors: 7 Years Ahead



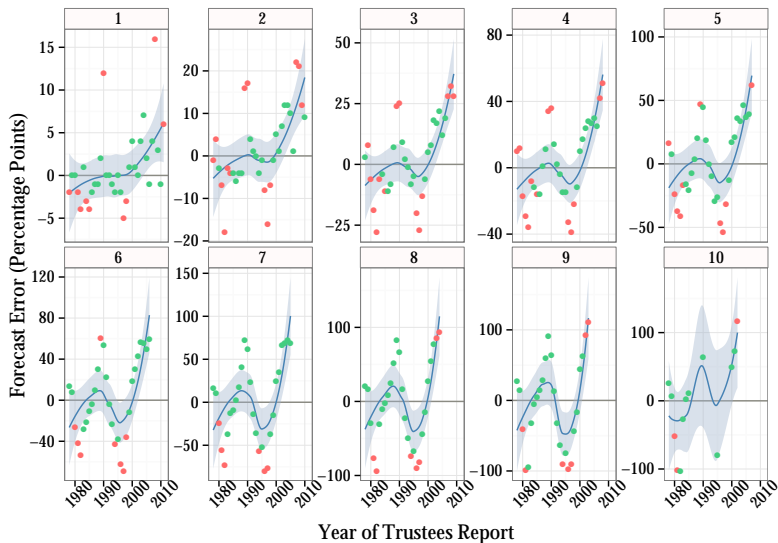
## Trust Fund Ratio Forecasting Errors: 8 Years Ahead



## Trust Fund Ratio Forecasting Errors: 9 Years Ahead



# Trust Fund Ratio Forecasting Errors: Summary



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- It can’t be learned: “Teaching psychology is mostly a waste of time” (Kahneman 2011)



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  - Many quotes; e.g. Goss: "The hard part is trying to balance the need to change on the basis of new ideas and understanding with the desire for consistency and stability over time"

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  - The Trustees and Technical Panel agree on many issues too, but the **lack of engagement or mutual understanding** is obvious

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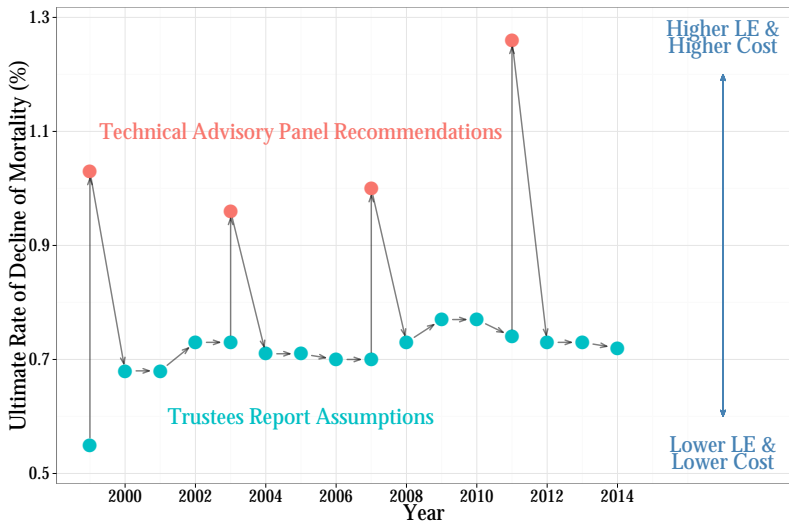
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E.g., Ultimate Rates of (All-Cause) Mortality Decline Assumptions



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Especially since 2000, Americans started living unexpectedly longer lives (due to statins, early cancer detection, etc.)

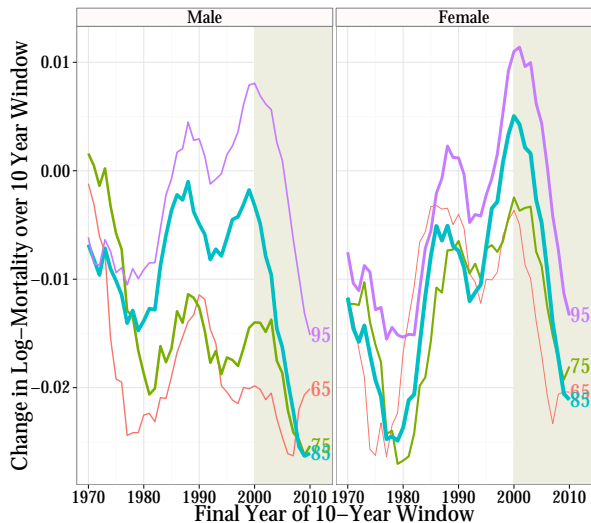
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For more information:

[GaryKing.org](http://GaryKing.org)