

What are we weighting for?

A mechanistic model for probability weighting

Ole Peters Alexander Adamou Yonatan Berman Mark Kirstein

D-TEA 2020, 16 June 2020



UNIVERSITÄT LEIPZIG

Mathematisches Institut





Mark Kirstein

Main results

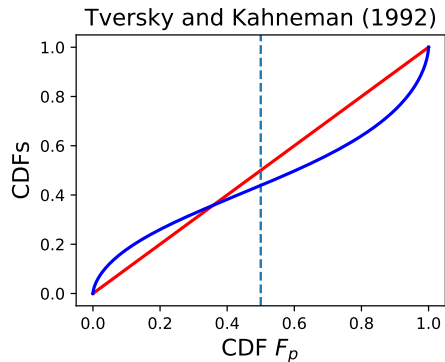
Main Results

Probability
Weighting

Ergodicity
Question

Estimation

Conclusion





Mark Kirstein

Main results

Main Results

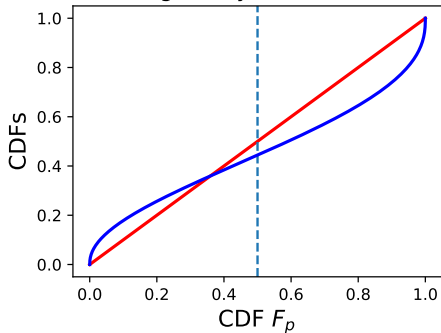
Probability
Weighting

Ergodicity
Question

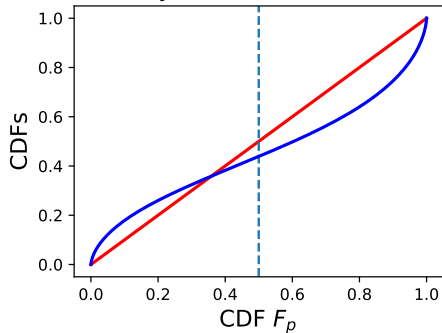
Estimation

Conclusion

Ergodicity Economics



Tversky and Kahneman (1992)





Mark Kirstein

Main Results

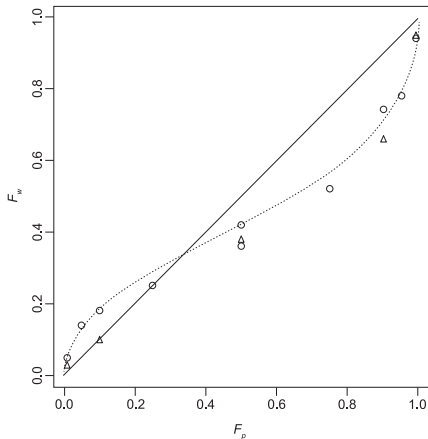
Probability
Weighting

Ergodicity
Question

Estimation

Conclusion

Definition of Probability Weighting (PW)



(Tversky and Kahneman 1992, p. 310, Fig. 1, relabelled axes)

- low probabilities treated as higher; high probabilities treated as lower
- stable empirical pattern: inverse-S shape
- Cumulative Prospect Theory (CPT)

Classical interpretation of PW:

- maladaptive irrational cognitive bias

In search of a mechanism

- ↪ How does this pattern emerge?
- ↪ Can we derive a functional form (rather than fit a function)?

Task: model payout, x , of a gamble as a random variable.

Disinterested Observer (DO)



DO assigns

PDF $p(x)$

CDF $F_p(x)$

Decision Maker (DM)



DM assigns different

PDF $w(x)$

CDF $F_w(x)$



Mark Kirstein

Main Results

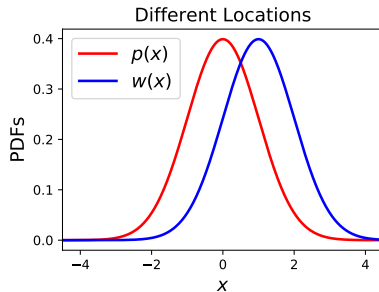
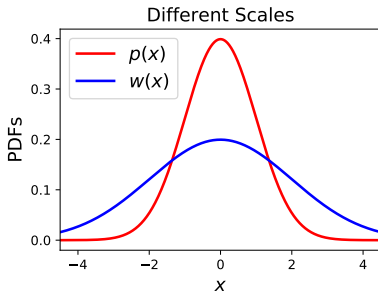
Probability
Weighting

Ergodicity
Question

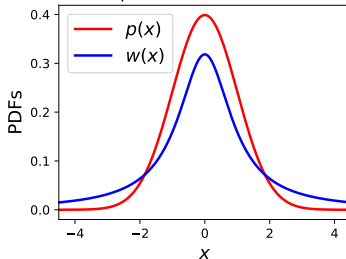
Estimation

Conclusion

Scales, Locations, Shapes



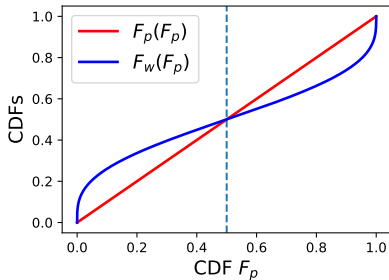
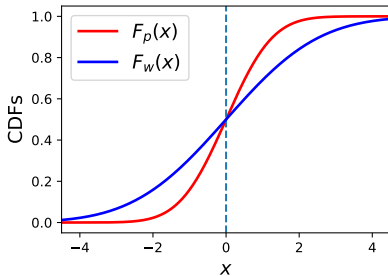
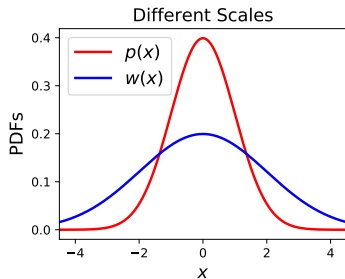
Different Shapes: Gaussian and t -distribution





Mark Kirstein

Thought experiment: DM assumes greater scale





For the case of two Gaussians with different scale we derive a functional form

$$w(p) = p^{\frac{1}{\alpha^2}} \underbrace{\frac{(2\pi\sigma^2)^{\frac{1-\alpha^2}{2\alpha^2}}}{\alpha}}_{\text{normalisation factor}}, \quad (1)$$

where

- DO's scale is σ
- DM's scale is $\alpha\sigma$
- $\alpha < 1 \rightarrow$ S shape
- $\alpha > 1 \rightarrow$ inverse-S shape



Mark Kirstein

Main Results

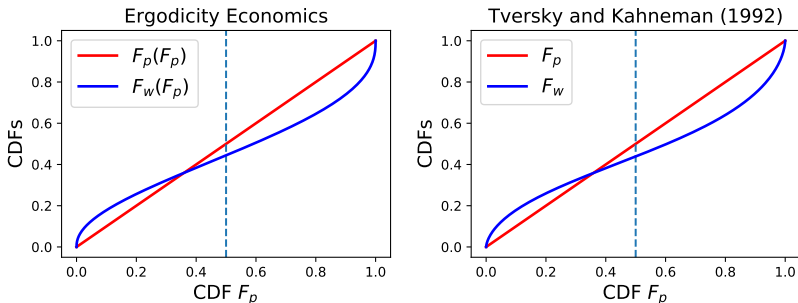
Probability
Weighting

Ergodicity
Question

Estimation

Conclusion

Interim conclusion



- DM's greater scale gives inverse-S shape (unimodal distributions)
- difference in locations gives asymmetry
- reproduces observations of probability weighting

Job done. Thank you for your attention ;)



Mark Kirstein

The Ergodicity Question

Main Results

Probability
Weighting

Ergodicity
Question

Estimation

Conclusion

Typical DO concern

What happens on average to
the **ensemble** of subjects?

\neq

Typical DM concern

What happens to me
on average over time?



Mark Kirstein

Why DM's greater scale?

- DM has no control over experiment
- experiment may be unclear to DM
- DM may not trust DO
- ...

Main Results

Probability
Weighting

Ergodicity
Question

Estimation

Conclusion



Mark Kirstein

Experiencing probabilities

Main Results

Probability
Weighting

Ergodicity
Question

Estimation

Conclusion

- probabilities are not observable
- probabilities encountered as
 - known frequency in ensemble of experiments (DO)
 - frequencies estimated over time (DM)

↪ **estimates have uncertainties – cautious DM accounts for these**



Rare Event

- $p(x) = 0.001$
 - 100 observations
 - $\sim 99.5\%$ get 0 or 1 events
 - $\hat{p}(x) = 0$ or $\hat{p}(x) = 0.01$
- ↪ 1000% uncertainty in $\hat{p}(x)$

Common Event

- $p(x) = 0.5$
 - 100 observations
 - $\sim 99.5\%$ get between 35 and 65 events,
 - $0.35 < \hat{p}(x) < 0.65$
- ↪ $\pm 15\%$ uncertainty in $\hat{p}(x)$

↪ small $p(x)$, small count
↪ small count, big uncertainty



Mark Kirstein

DMs don't like surprises

To avoid surprises, DMs add estimation uncertainty $\varepsilon [p(x)]$ to every estimated probability, then normalize, s.t.

$$w(x) = \frac{p(x) + \varepsilon [p(x)]}{\int (p(s) + \varepsilon [p(s)]) ds}$$

Main Results

Probability
Weighting

Ergodicity
Question

Estimation

Conclusion

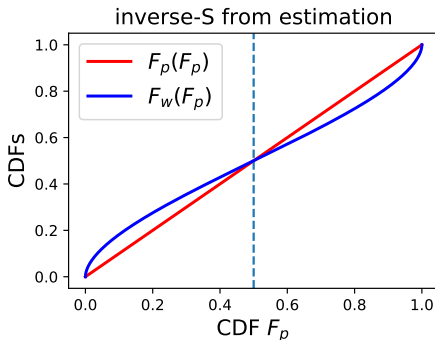
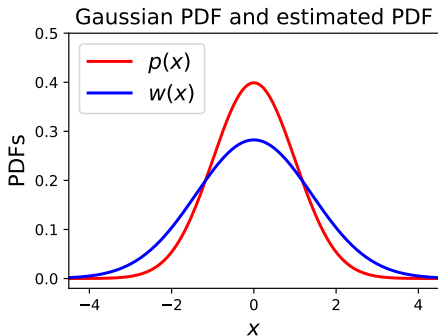


Mark Kirstein

DMs don't like surprises

To avoid surprises, DMs add estimation uncertainty $\varepsilon [p(x)]$ to every estimated probability, then normalize, s.t.

$$w(x) = \frac{p(x) + \varepsilon [p(x)]}{\int (p(s) + \varepsilon [p(s)]) ds}$$





Mark Kirstein

Main Results

Probability
Weighting

Ergodicity
Question

Estimation

Conclusion

Conclusion

Classical interpretation of PW:

- maladaptive irrational cognitive bias

Ergodicity Economics and probability weighting

- inverse-S shape: neutral indicator of different models of the world
- reported observations consistent with DM's extra uncertainty
- may arise from DM estimating probabilities over time

↪ Probability weighting is rational cautious behaviour under uncertainty over time

- Manuscript at <https://www.researchers.one/article/2020-04-14>
- Interactive code at <https://bit.ly/lml-pw-count-b>



Mark Kirstein

Main Results

Probability
Weighting

Ergodicity
Question

Estimation

Conclusion

Conclusion

Classical interpretation of PW:

- maladaptive irrational cognitive bias

Ergodicity Economics and probability weighting

- inverse-S shape: neutral indicator of different models of the world
- reported observations consistent with DM's extra uncertainty
- may arise from DM estimating probabilities over time

↪ Probability weighting is rational cautious behaviour under uncertainty over time

- Manuscript at <https://www.researchers.one/article/2020-04-14>
- Interactive code at <https://bit.ly/lml-pw-count-b>

Thank you for your attention!



Mark Kirstein

References I

References



Tversky, Amos and Daniel Kahneman (1992). "Advances in Prospect Theory: Cumulative Representation of Uncertainty". *Journal of Risk and Uncertainty* 5 (4), pp. 297–323.
DOI:[10.1007/BF00122574](https://doi.org/10.1007/BF00122574) (cit. on p. 4).