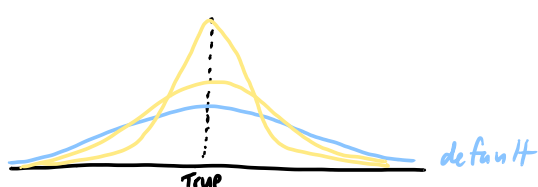


- Motivation:
- ① Determine which sequential sampling models manage to extend to biased prior scenarios. How well do these models capture the distribution of choices at various exogenous response times?
 - ② Provide striking evidence that the standard Bayesian model breaks in an exogenous response time paradigm.
 - ③ Document how priors are manipulated.

- Models:
- ① Models that look at KL divergence between prior ("default distribution") and posterior ("choice probability distribution").
 - Cheyette + Piantadosi (2020) numerosity estimation model
 - Bayesian models that sample from the posterior.
 - ② Bayesian models that optimally respond to posterior.
 - ③ DDM (simple and w/ collapsing bounds).

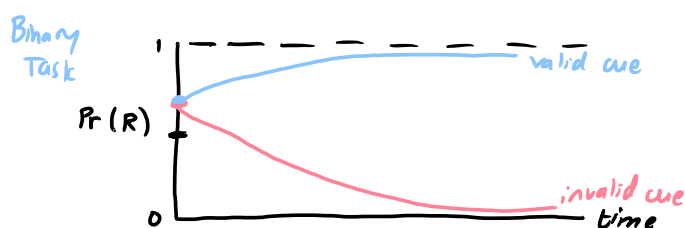
Cheyette:



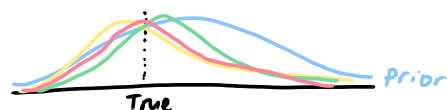
Choice probability distributions are (1) initially wide and (2) always centered on the true value.



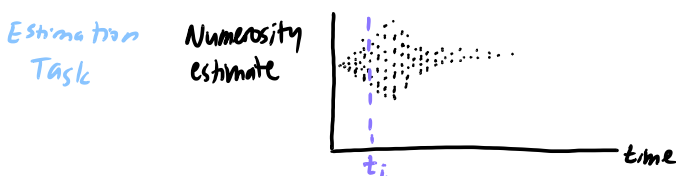
Randomness in choice \downarrow in response time.



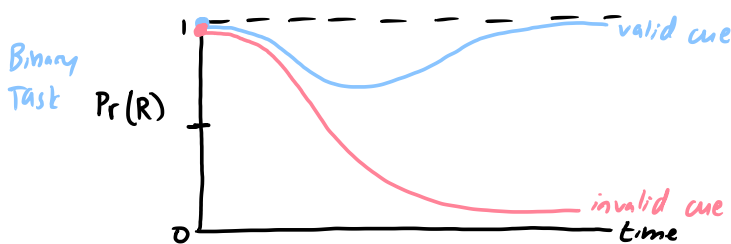
Bayesian:



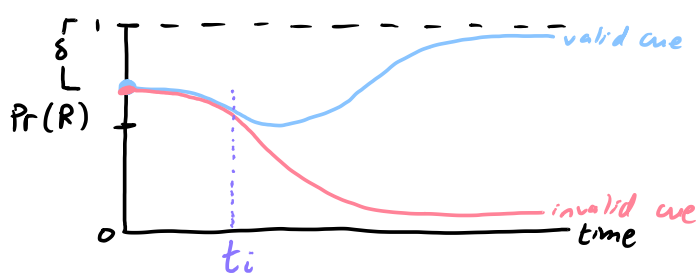
The Bayesian MAP estimate will land close to the true value, relative to the variance of samples. Part of the key is that with a biased prior, responder should be very predictable.



Randomness in choice starts low, rises, then dips again.

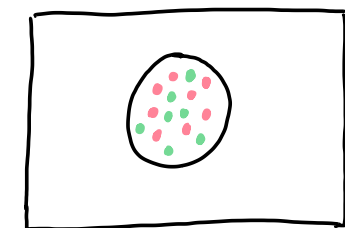


Expectation:
(I feel that DDM predictions will look like this.)



This δ -deviation from the standard Bayesian model is very interesting. Also, evidence that looks like this might be evidence against the KL-divergence models.

Task:



More red or green balls?

- Use 52-48 proportions.
- Exogenous RTs.
- Induce 60-40 priors.
- Small circle to limit attention effects.

- Notes:
- ① Need an adaptive algorithm to select a time t when subjects exhibit more randomness. Expect subjects to have different t_i .
 - ② Subjects cannot know the exogenous RT.
 - ③ Careful w/ how you influence priors.
 - ④ Look carefully at deliberation time. We don't have much control over how quickly subjects respond, but we don't want them to deliberate after the observation time has ended.