

Globally Inaccurate Stereotypes Can Result from Locally Rational Exploration:

Evidence from a Formal Model and Human Experiments



Xuechunzi Bai



Susan T. Fiske

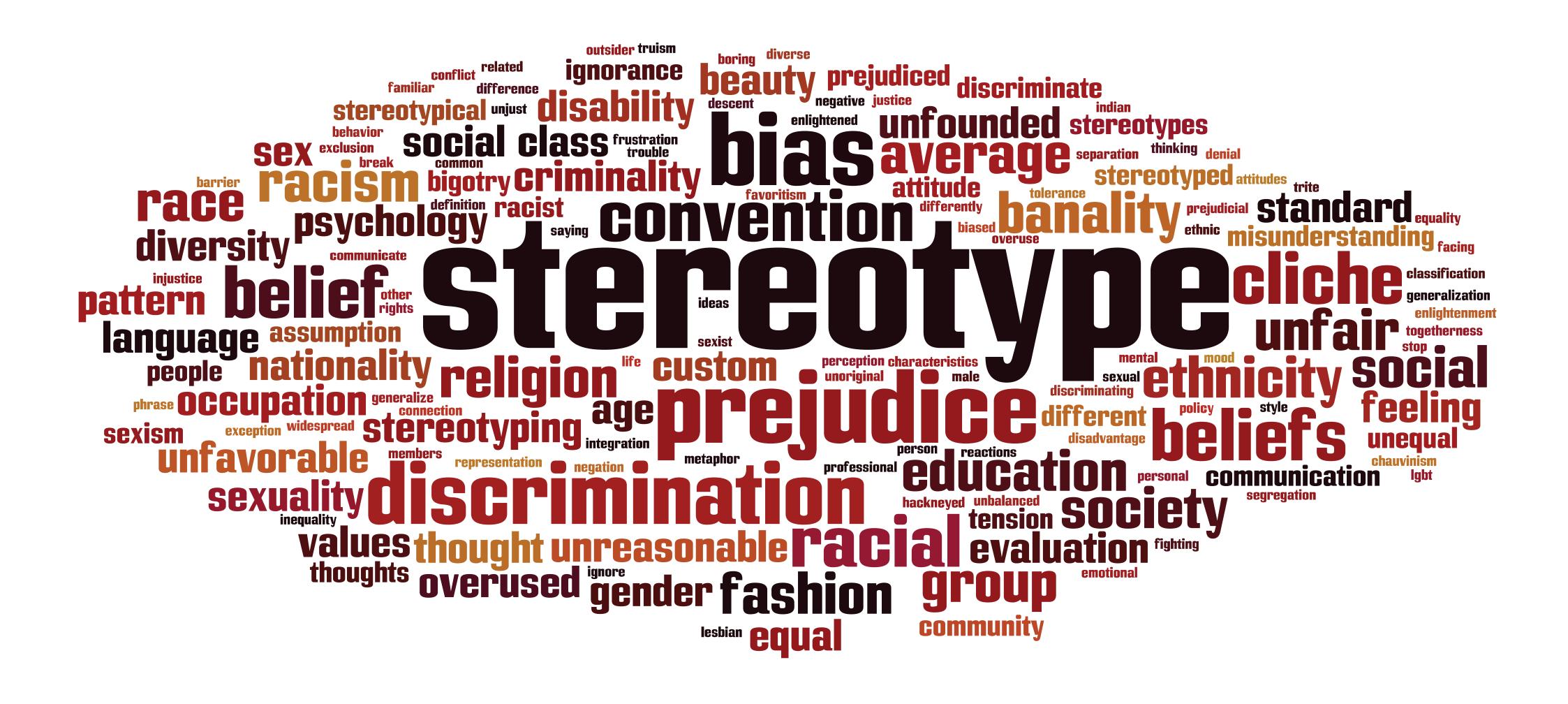


Thomas L. Griffiths

(invited revision PsychSci)

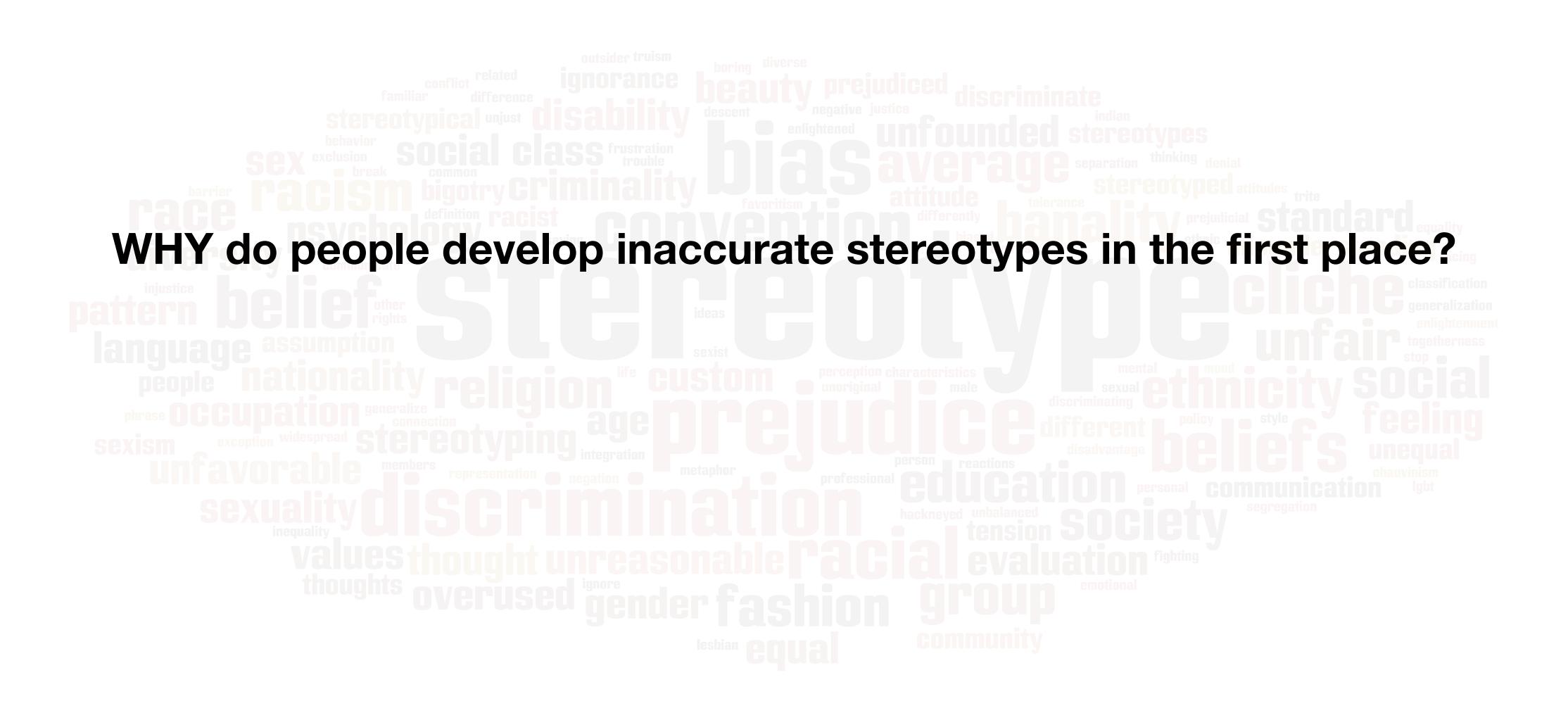
Motivation

Inaccurate stereotypes are prevalent and consequential



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Existing research

Motivational and Cognitive

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- Motivational (as a group member):
 - Identity
 - Dominance

Existing research

Motivational and Cognitive

- Motivational (as a group member):
 - Identity
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- Cognitive (as an information processor):
 - Categorization
 - Selective attention

Our work: A functional minimal-process paradigm

A simple exploratory sampling

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A simple exploratory sampling

- Minimal:
 - Without motivational bias
 - Without cognitive limits

Our work: A functional minimal-process paradigm

A simple exploratory sampling

- Minimal:
 - Without motivational bias
 - Without cognitive limits
- Functional:
 - Optimal solution in the environment given

Our hypothesis

Globally inaccurate stereotypes can result from locally rational exploration

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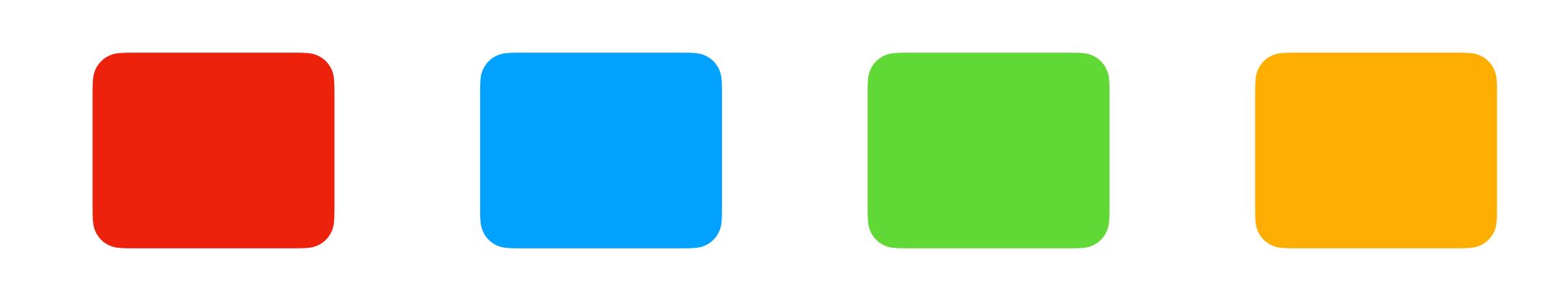
By adopting a functional/rational analysis,

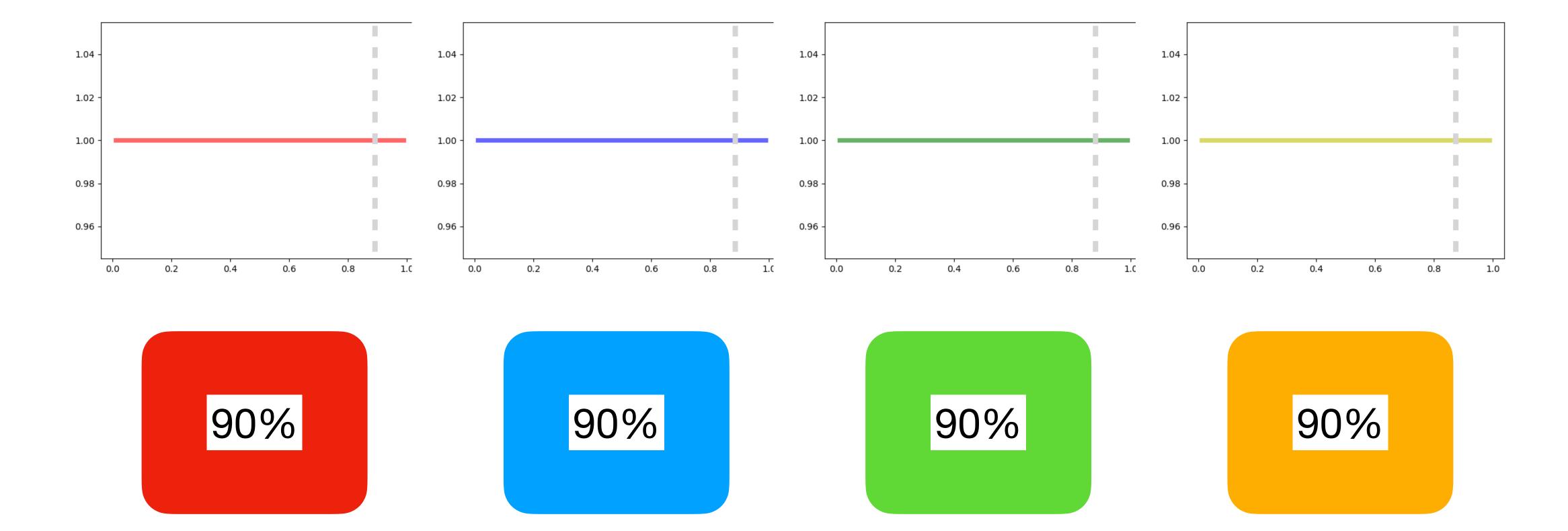
we are NOT saying social stereotypes are accurate or morally right.

Our hypothesis

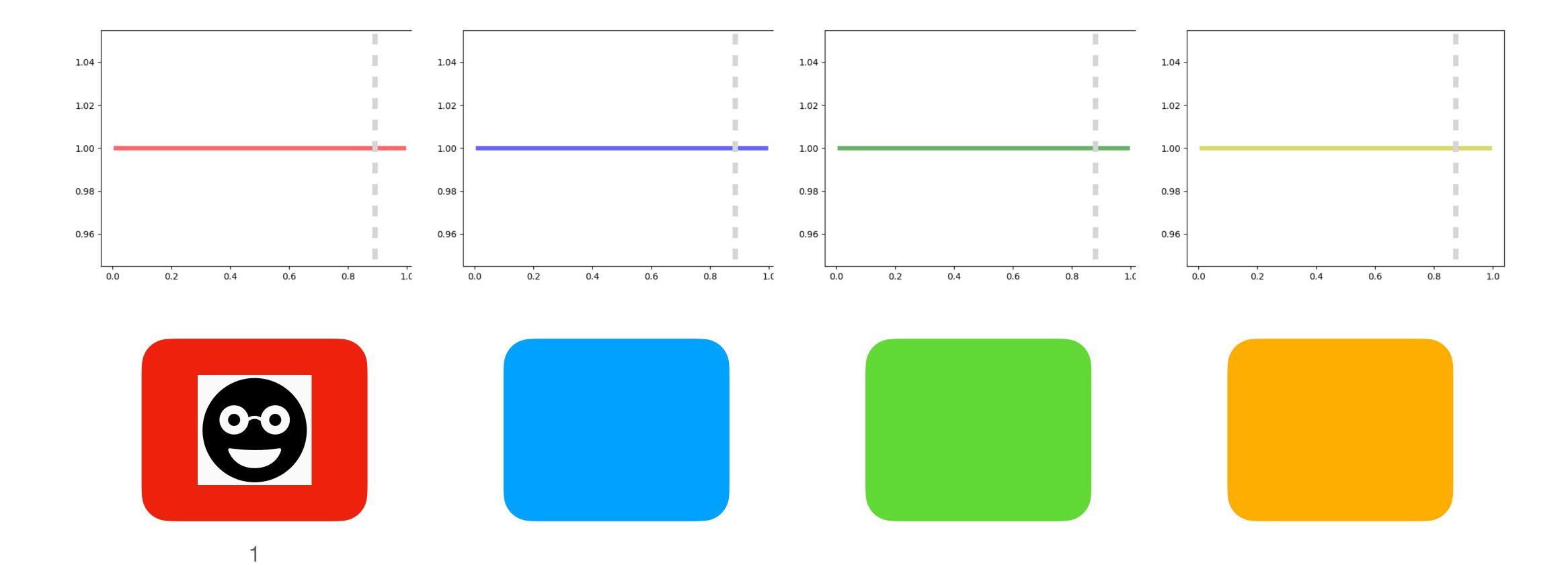
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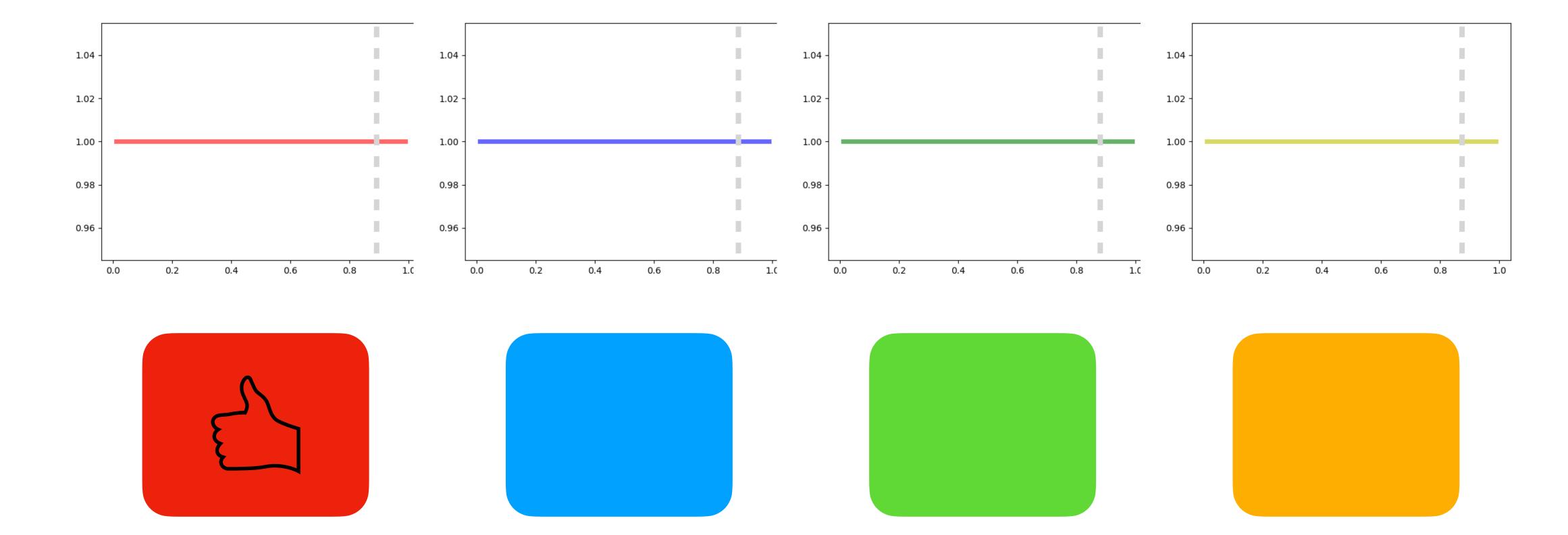
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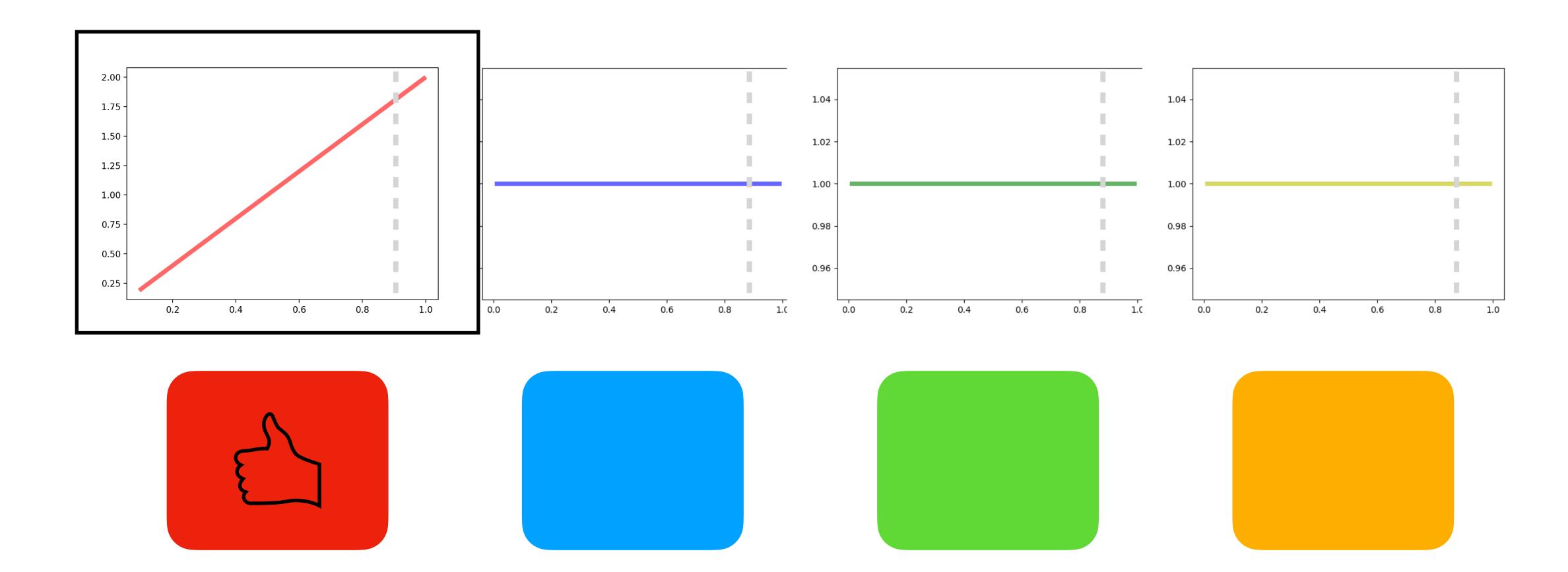


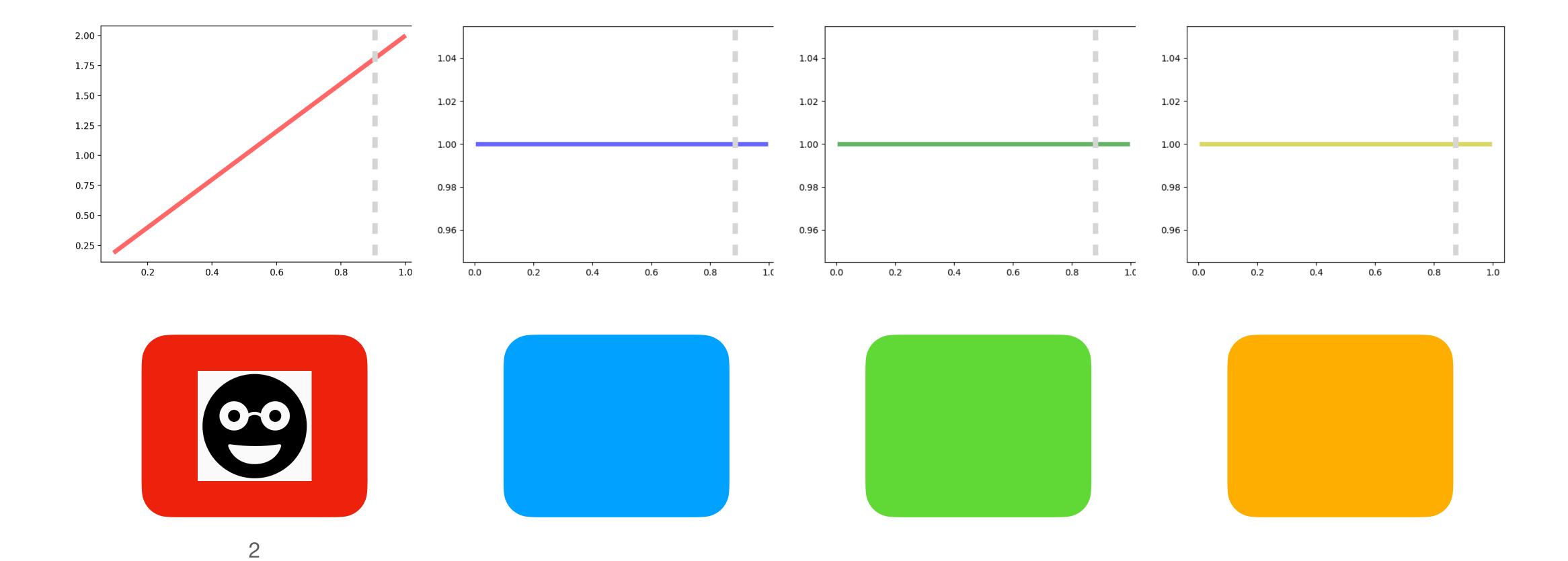


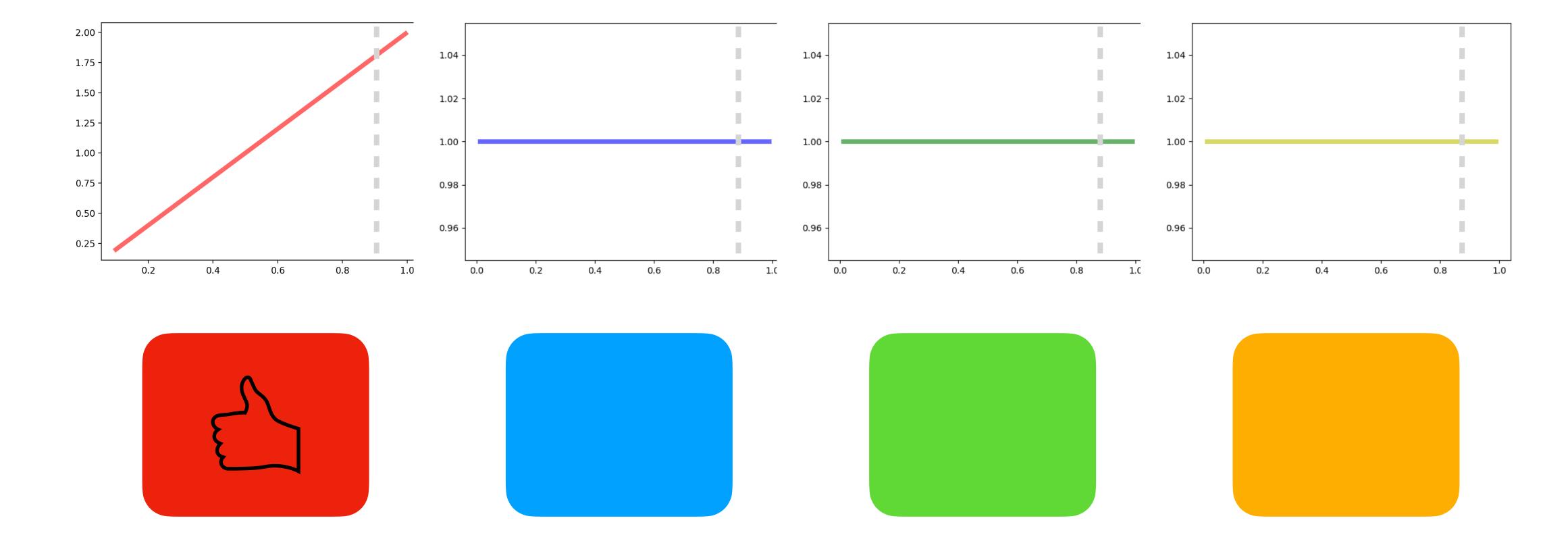


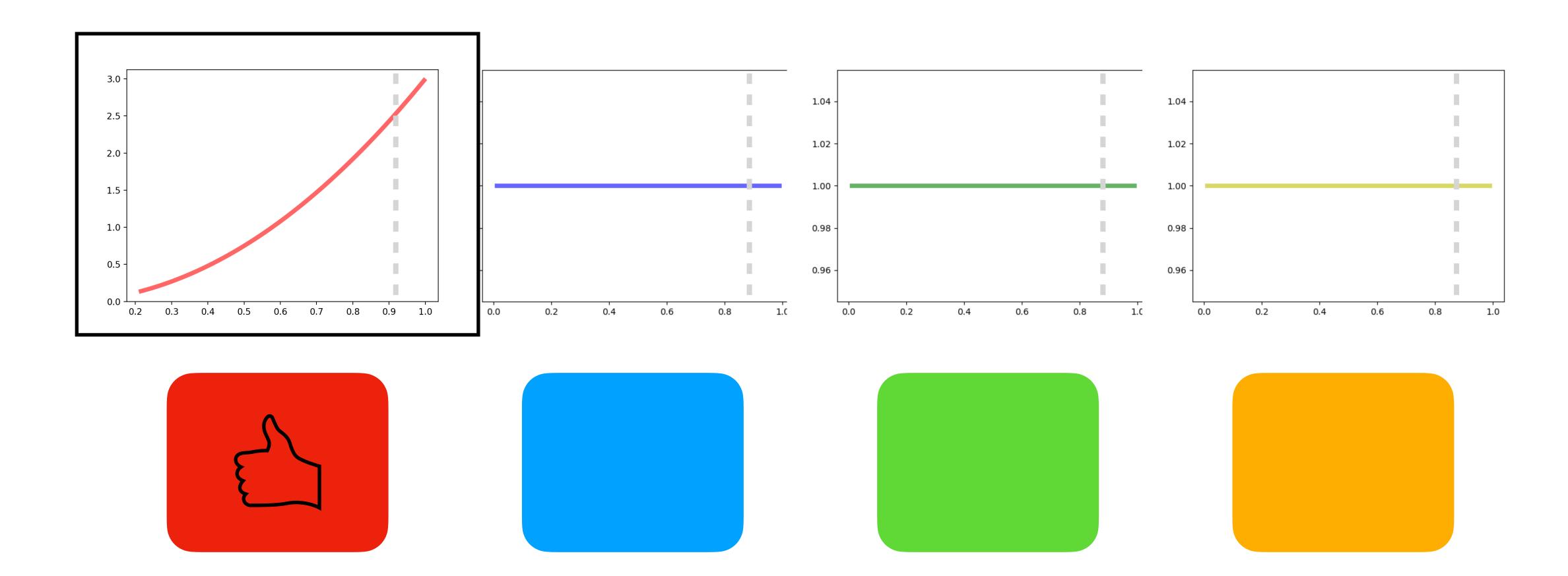


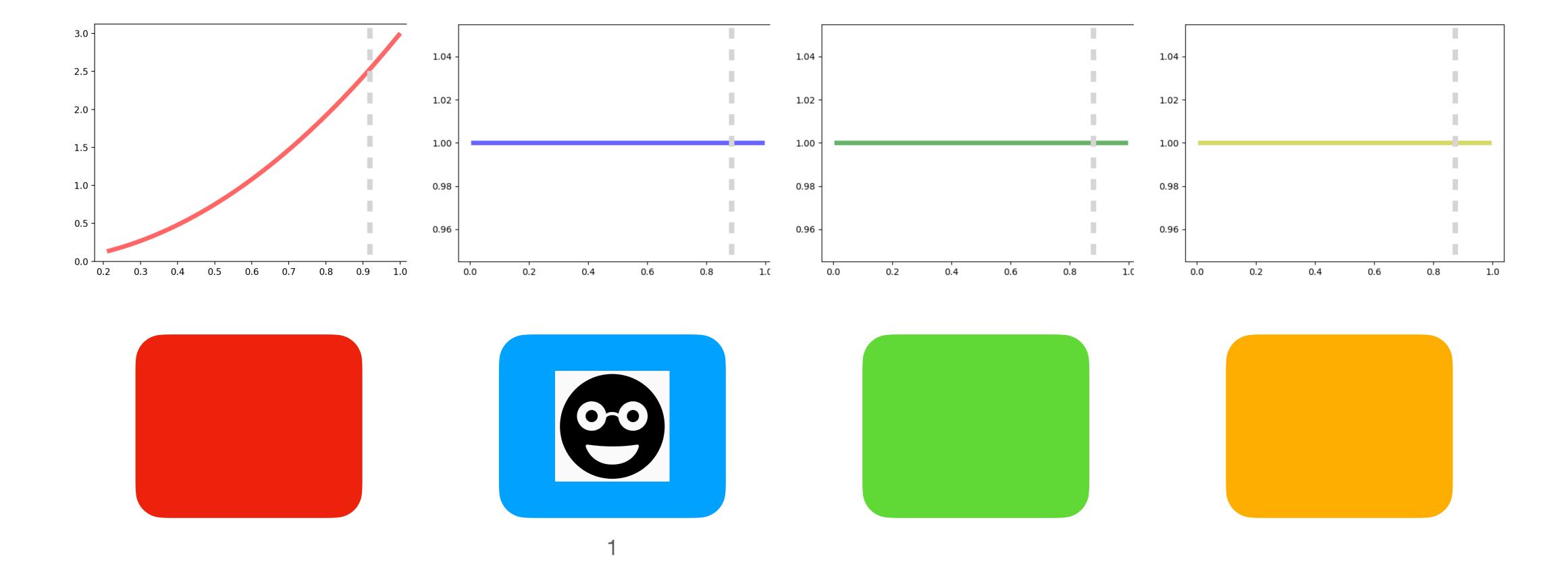


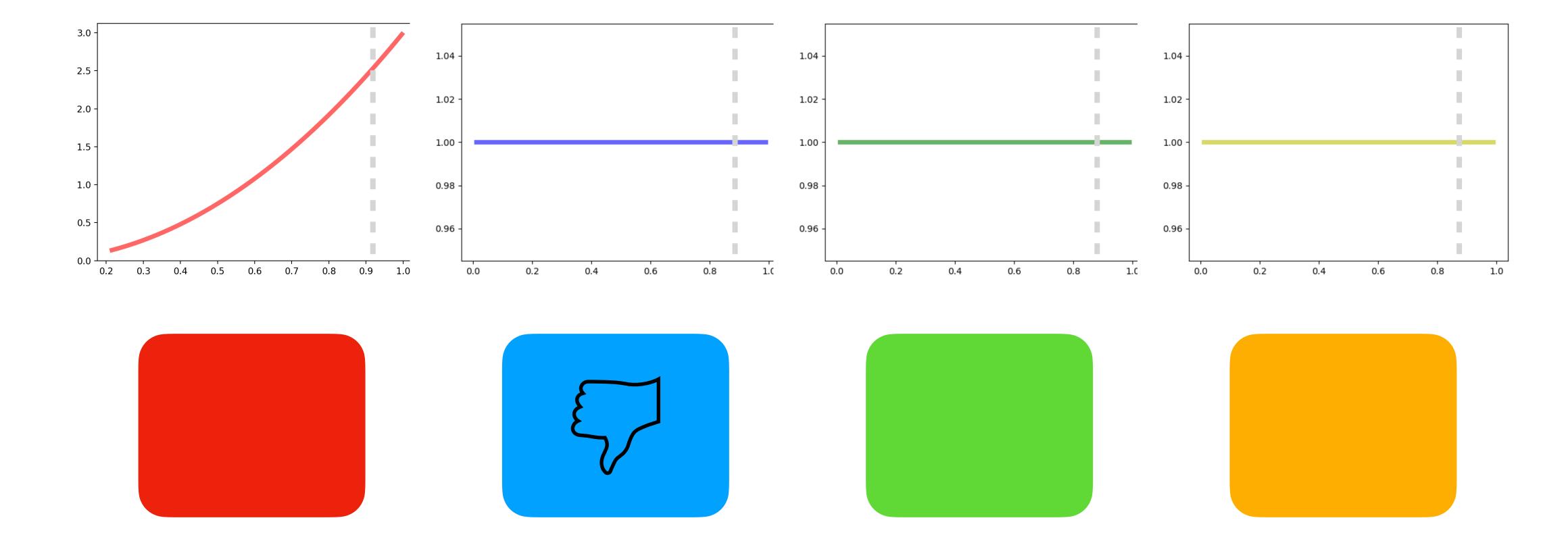


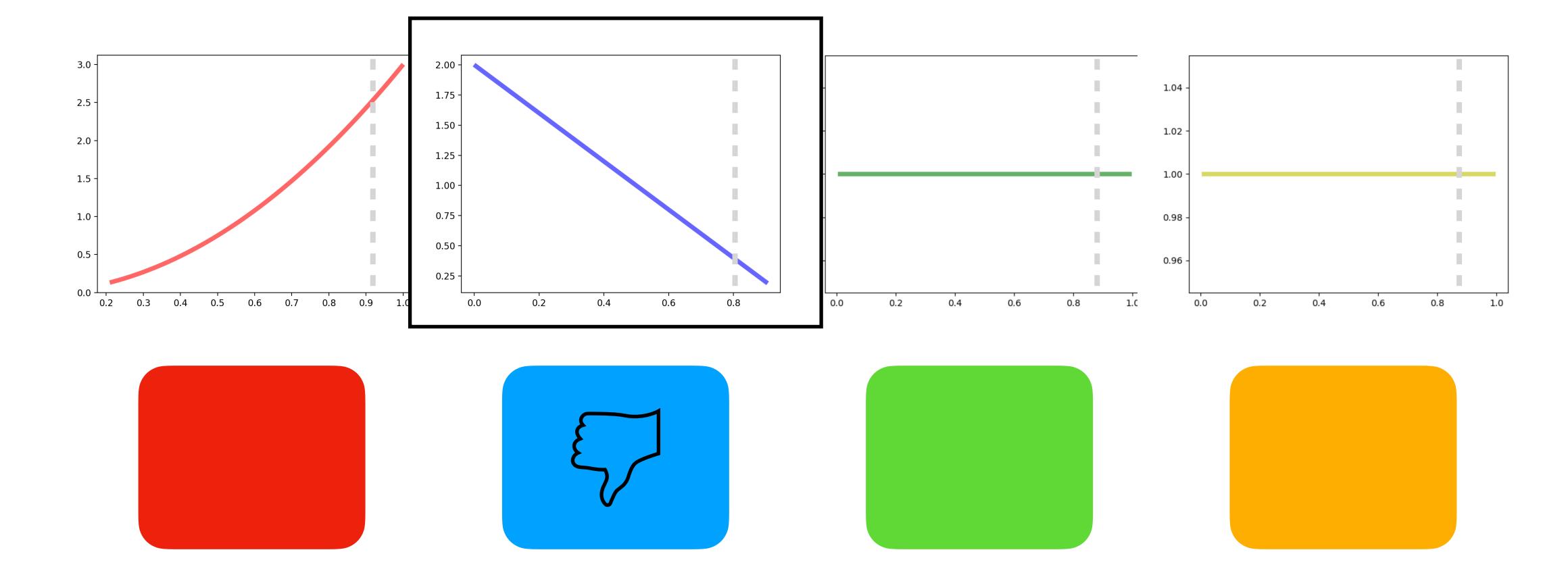


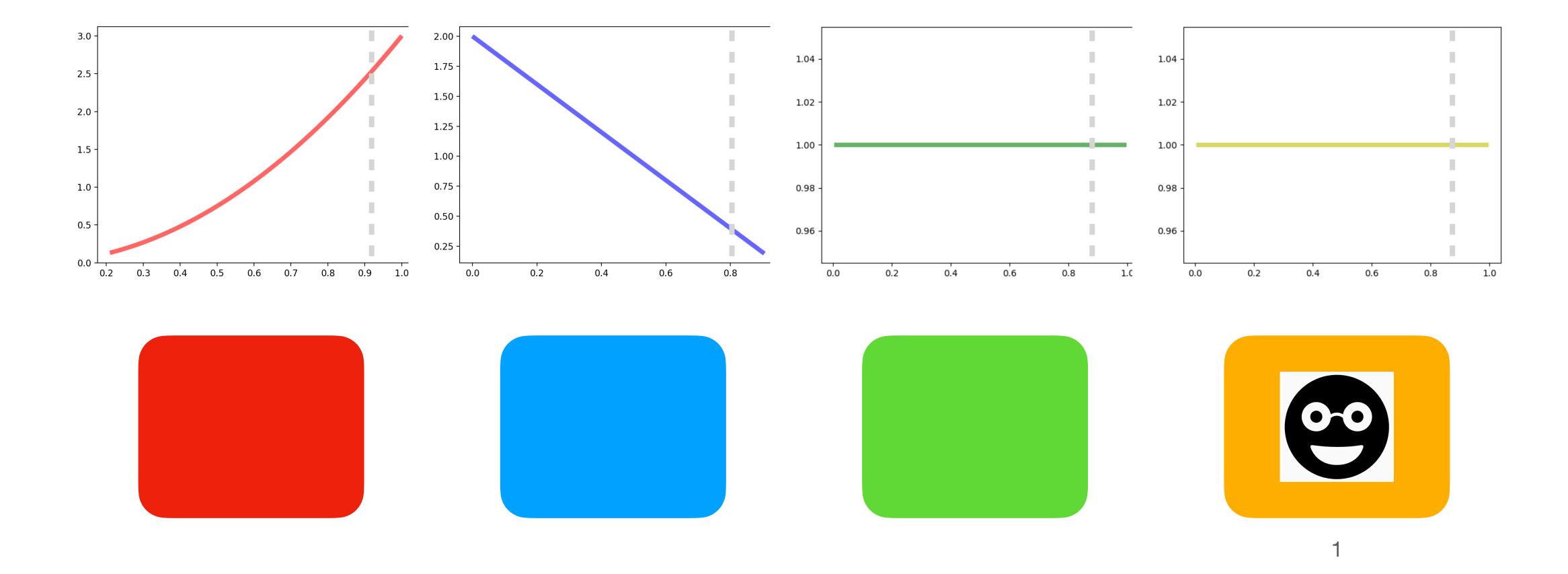


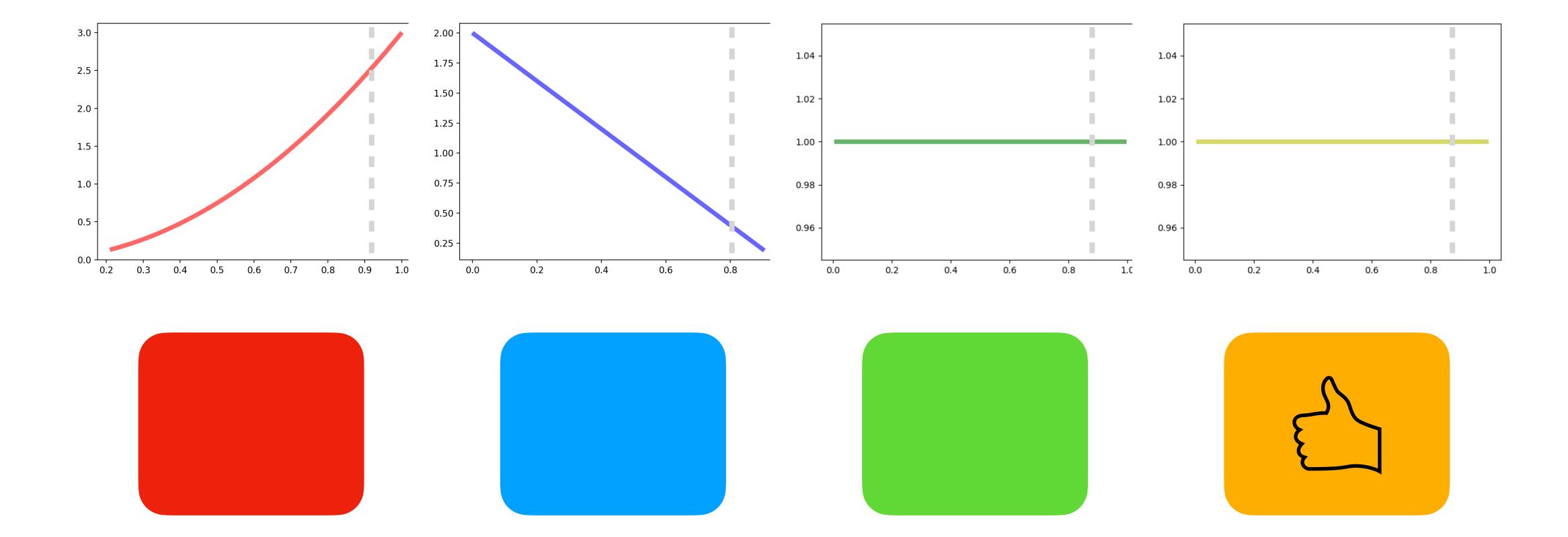


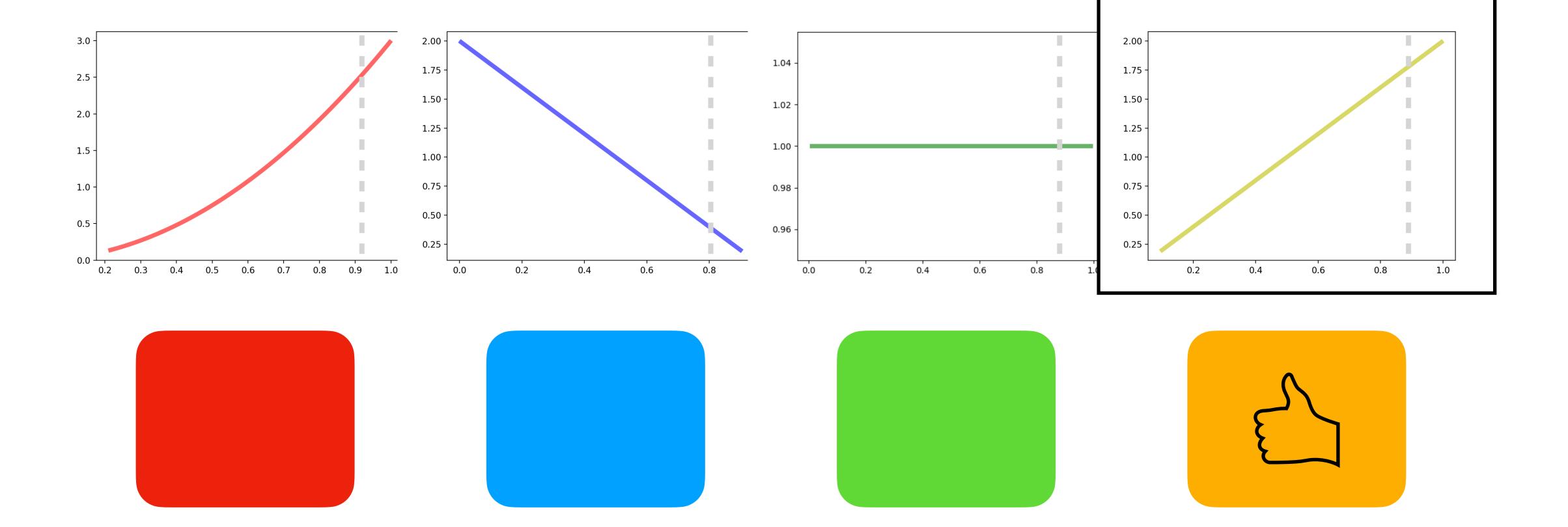


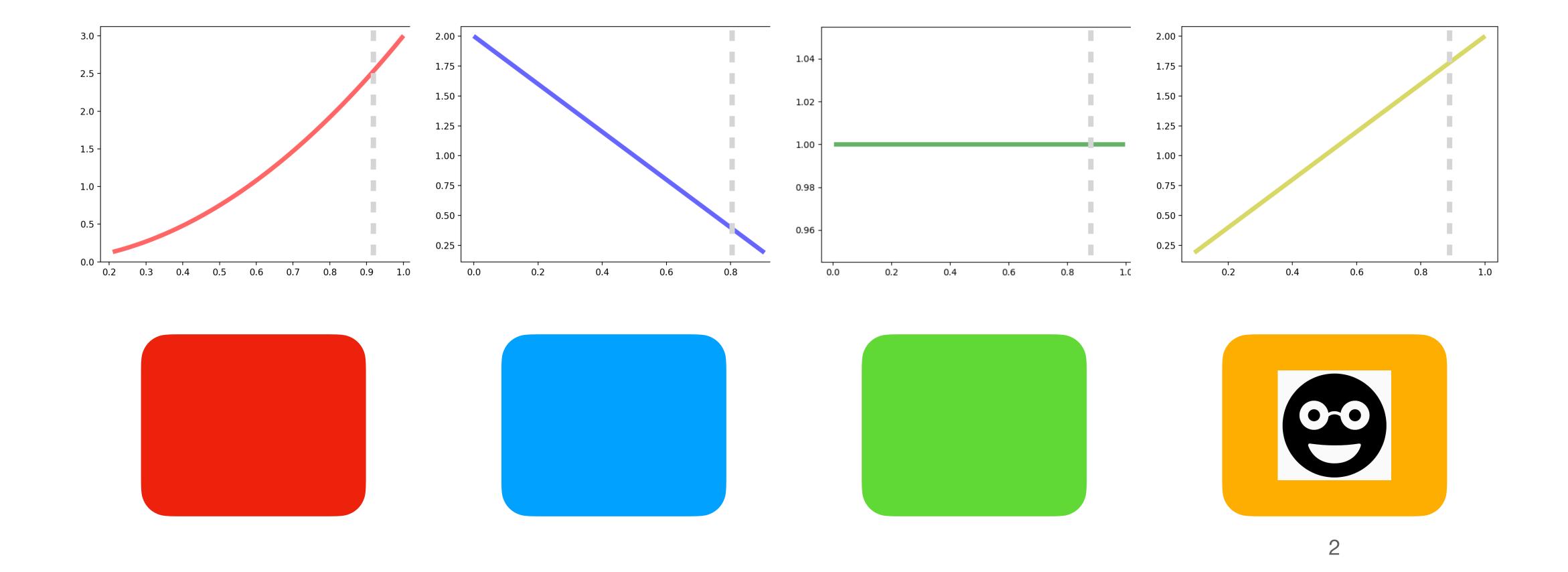


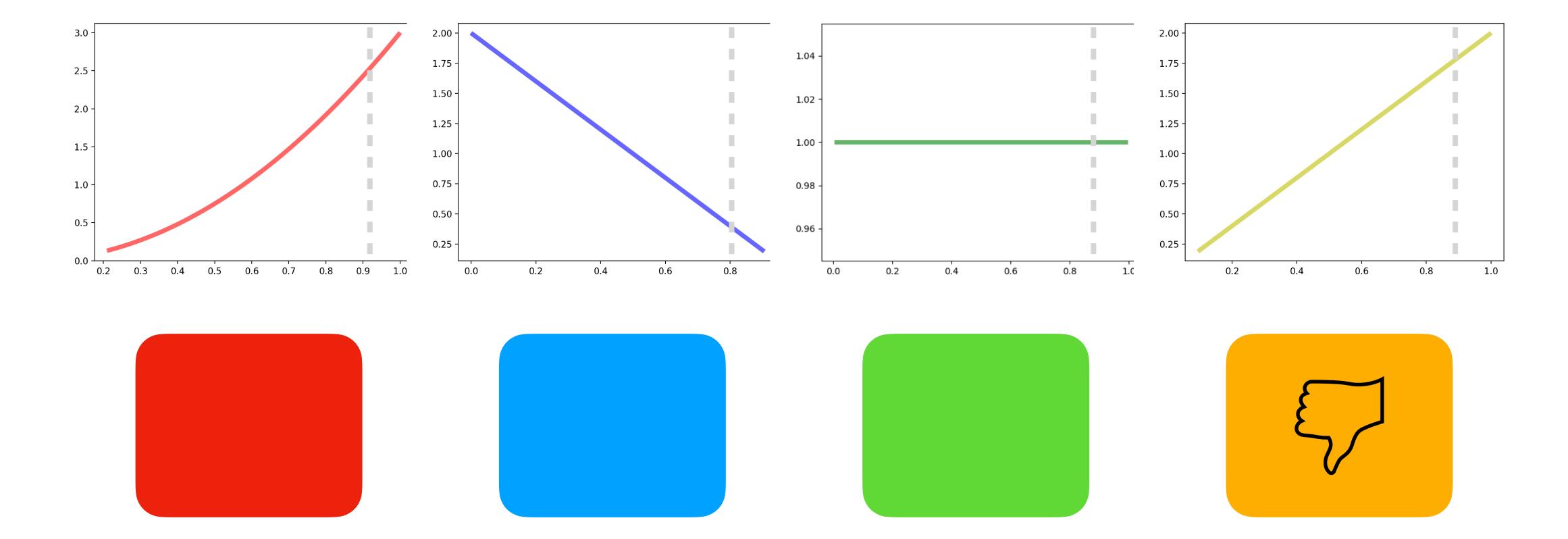


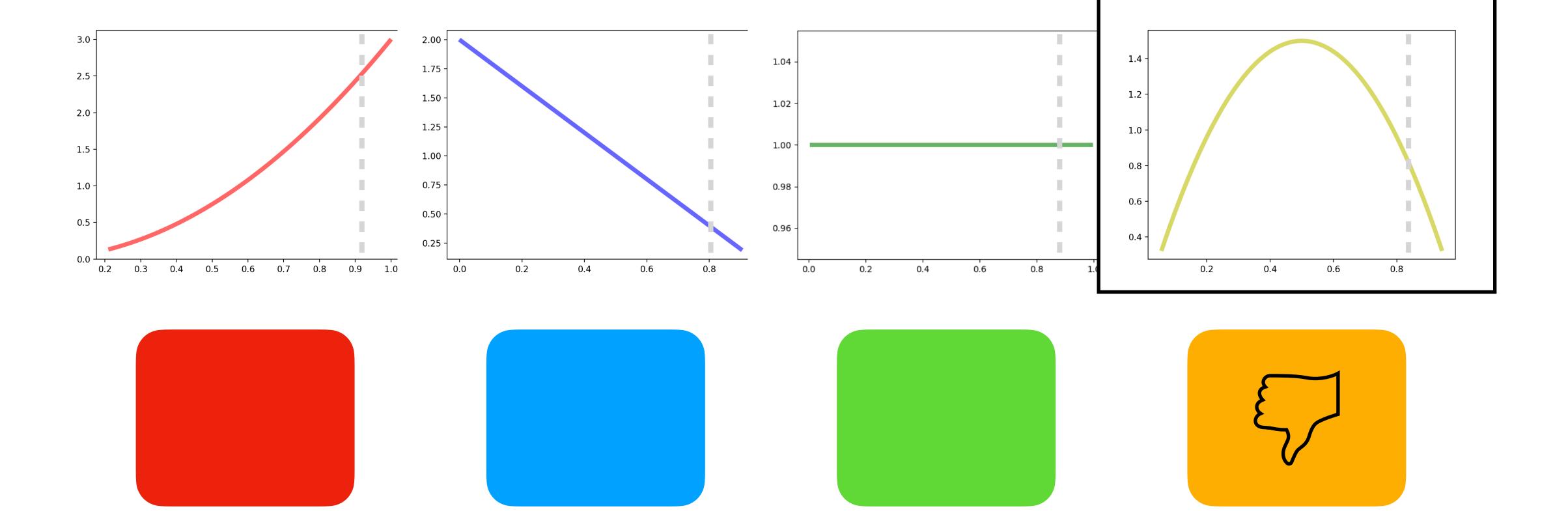


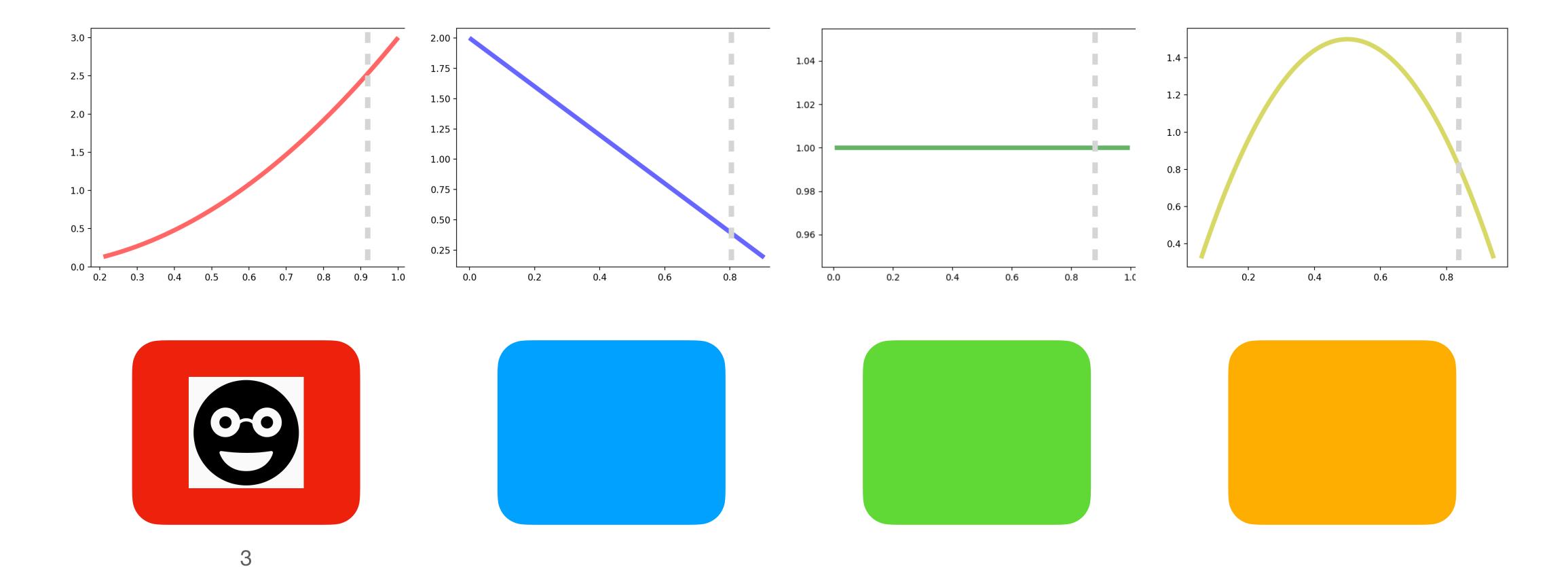


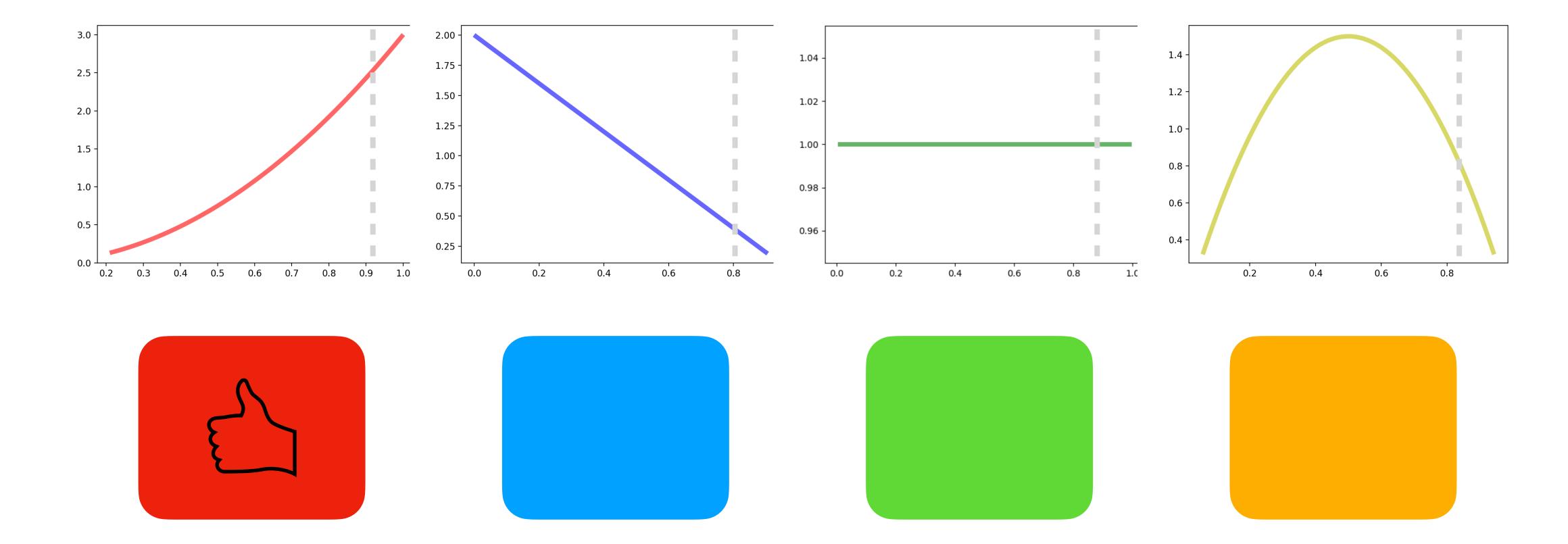


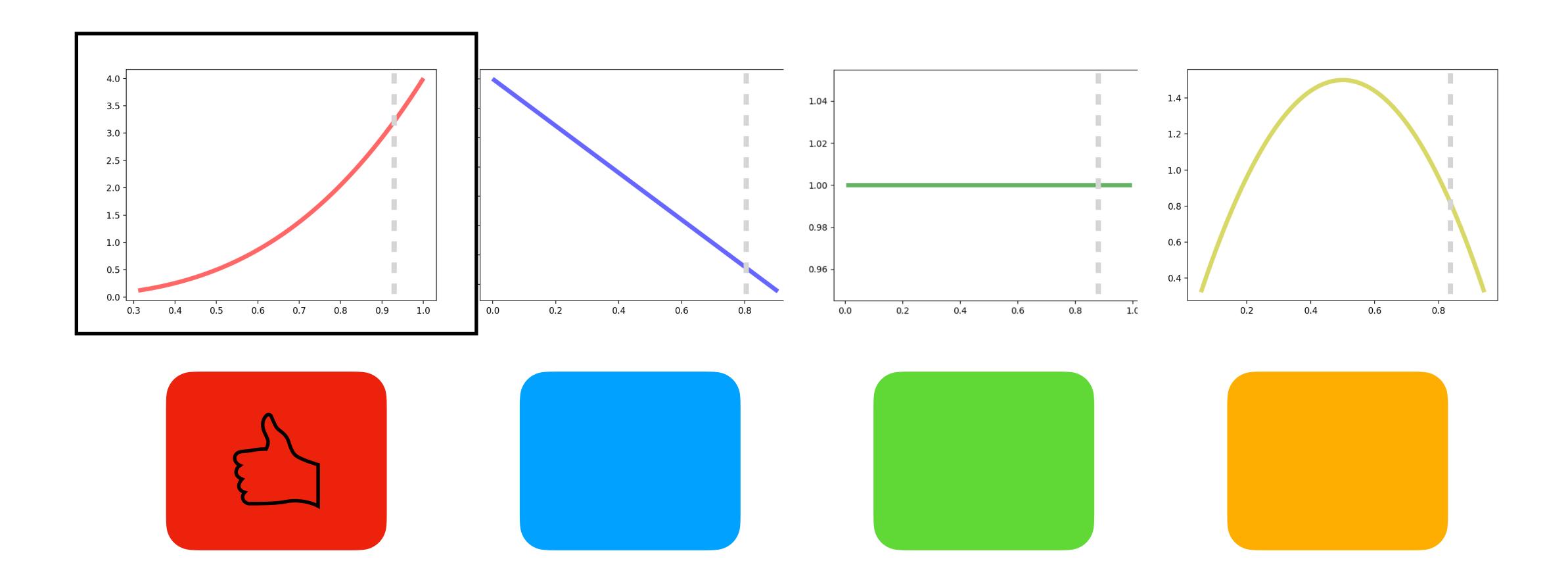


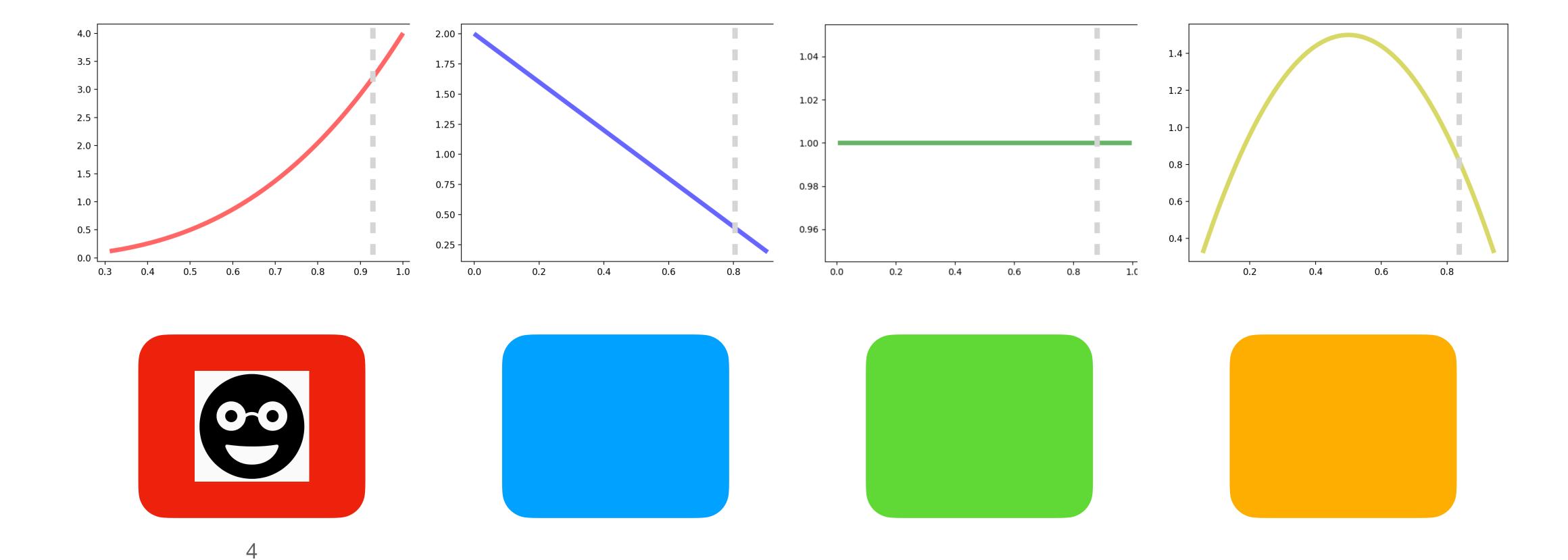


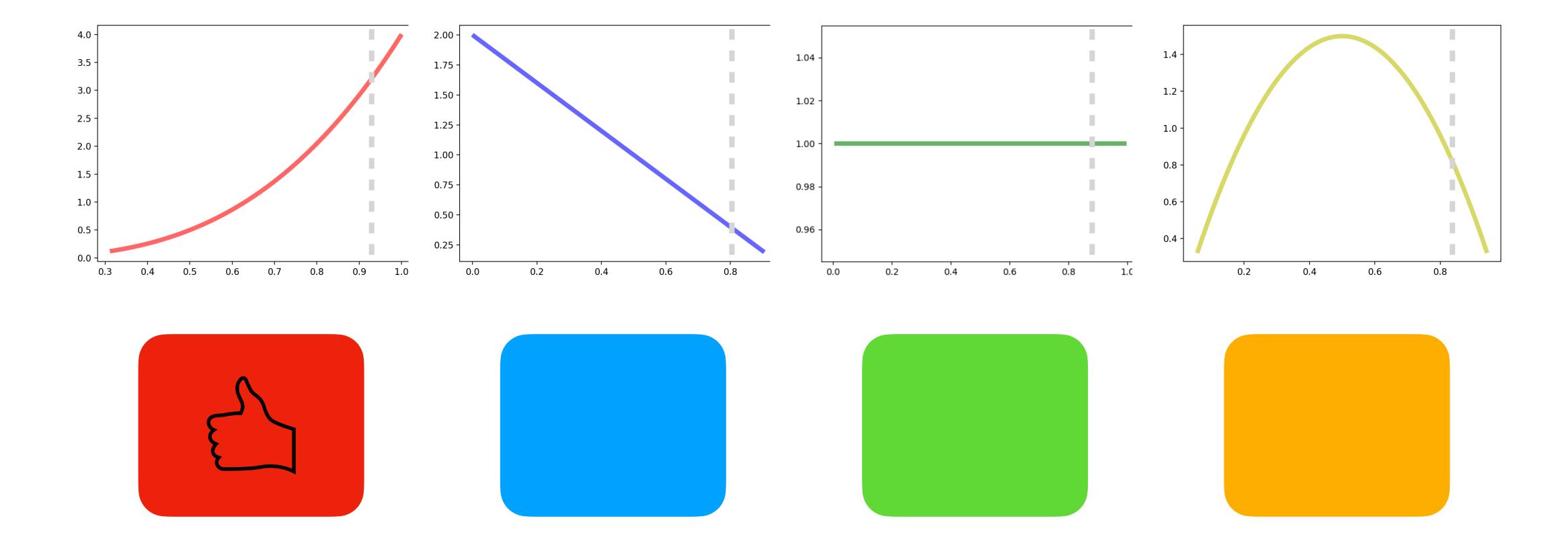


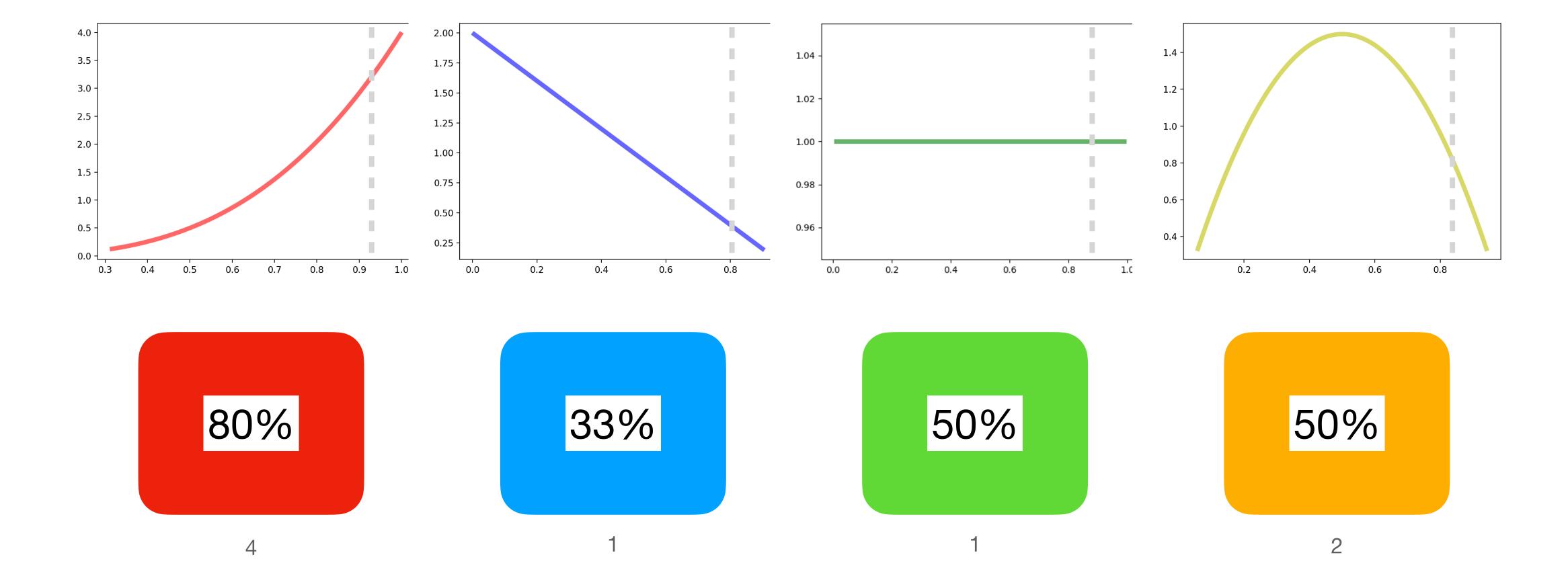












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- The process is rational/minimal: Belief updating
 - Not unable to absorb new information
- The outcome is inaccurate: Find an ostensibly best group
 - Although unintended, inaccurate impressions about the under-explored groups

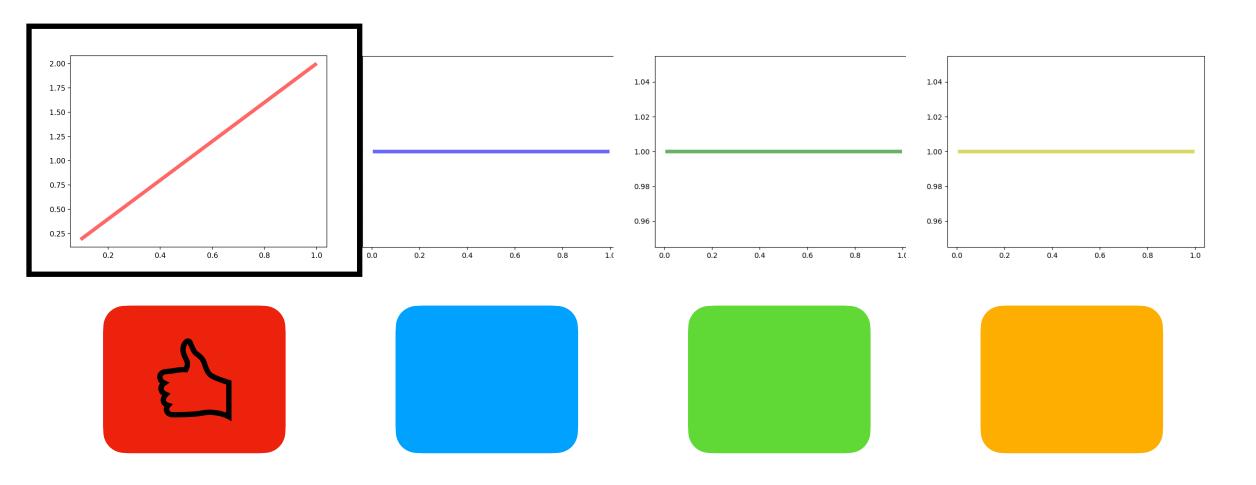
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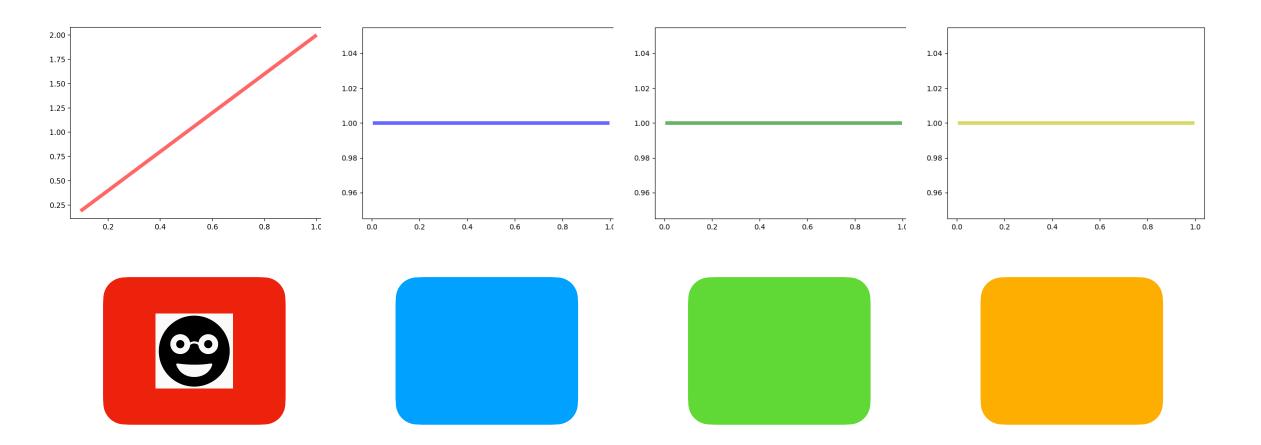
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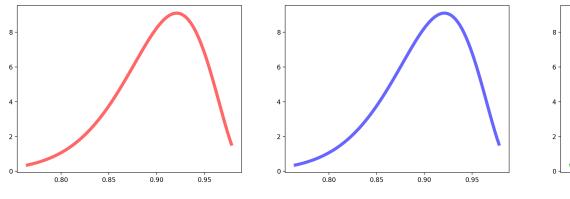
Multi-Armed Bandit Problem

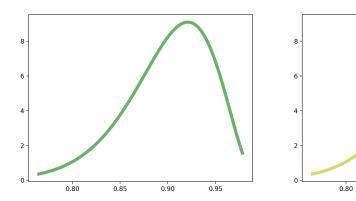
Explore v. Exploit dilemma in Reinforcement Learning (Sutton & Barto, 2018)



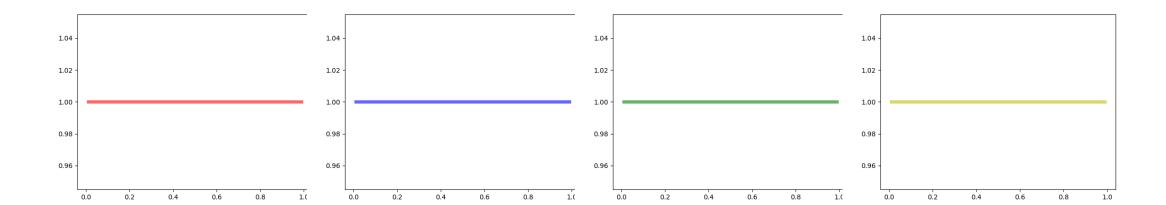


Thompson Sampling









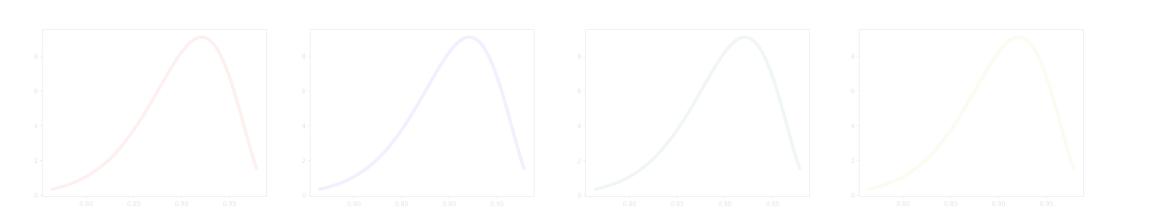
Initial structure: No prior bias



Environment structure: Identical high reward

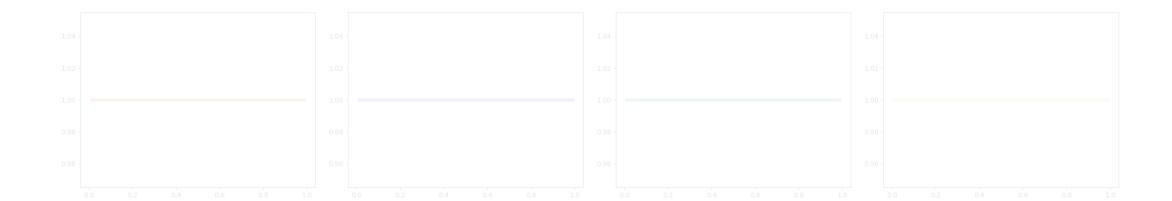
Initial structure: No prior bias

Sampling strategy: <u>Thompson</u> v. <u>Random</u>



Environment structure: Identical reward

Partner choices: How many times did the model interact with each group?



Initial structure: No prior bias

Reward estimations: What is the estimated reward for each group?











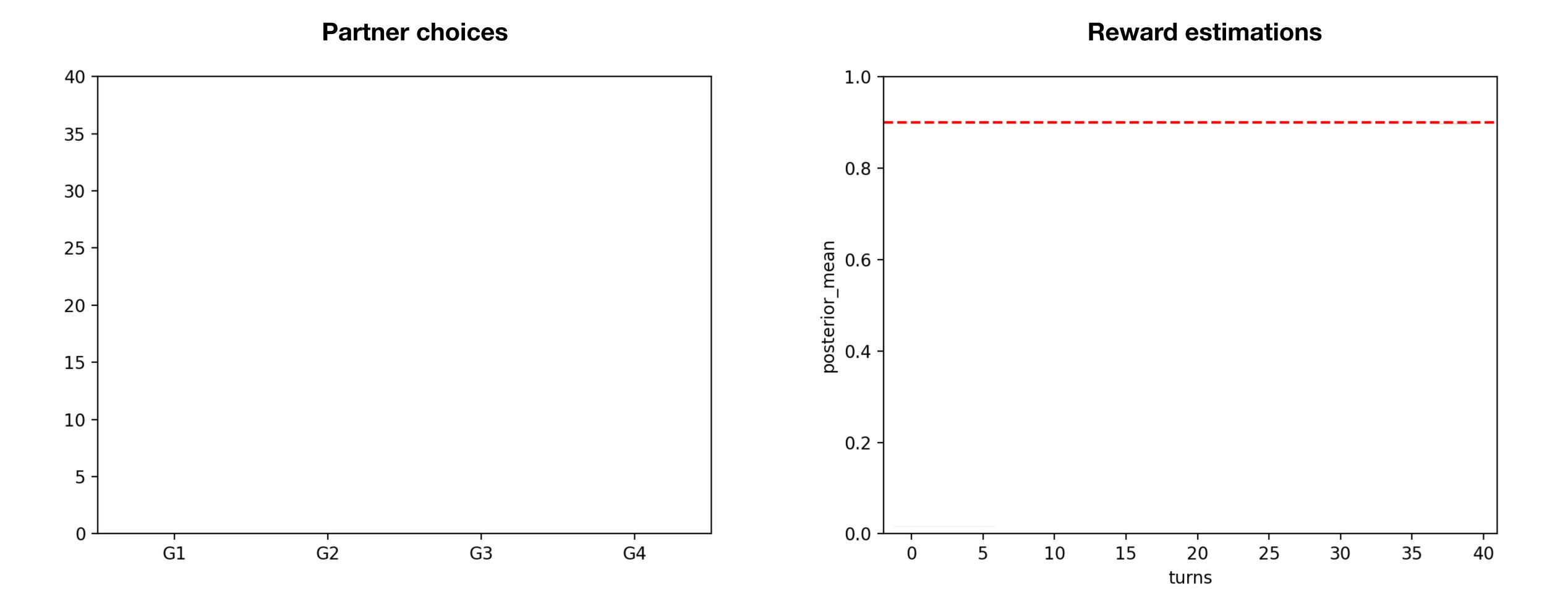




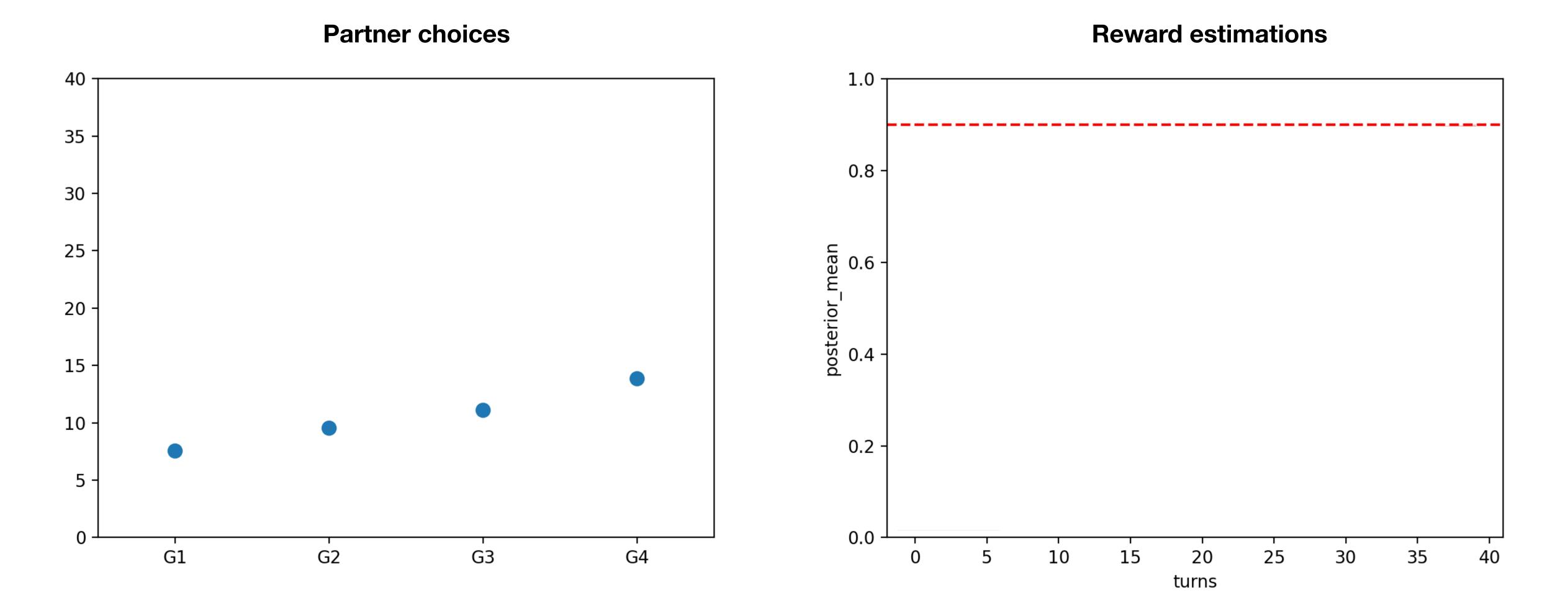


Sampling strategy: Thompson v. Random

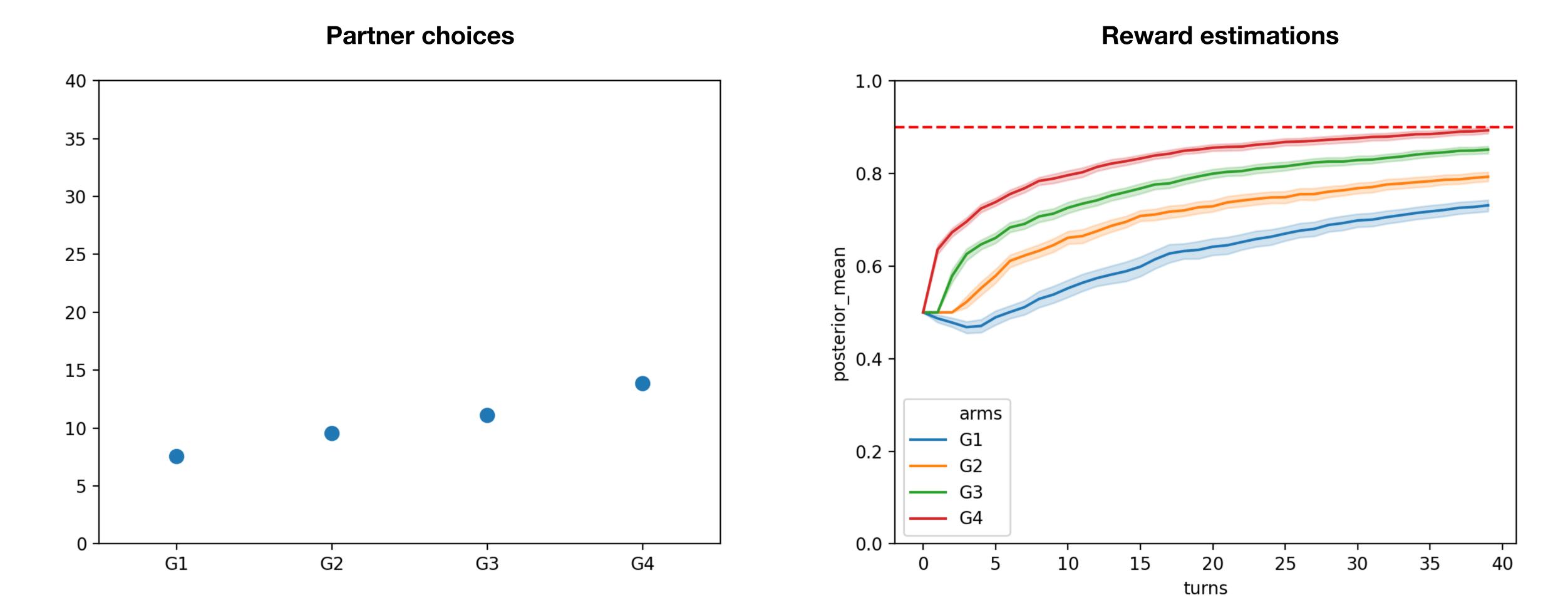
Identical reward (μ = 0.9), No prior, Random sampling



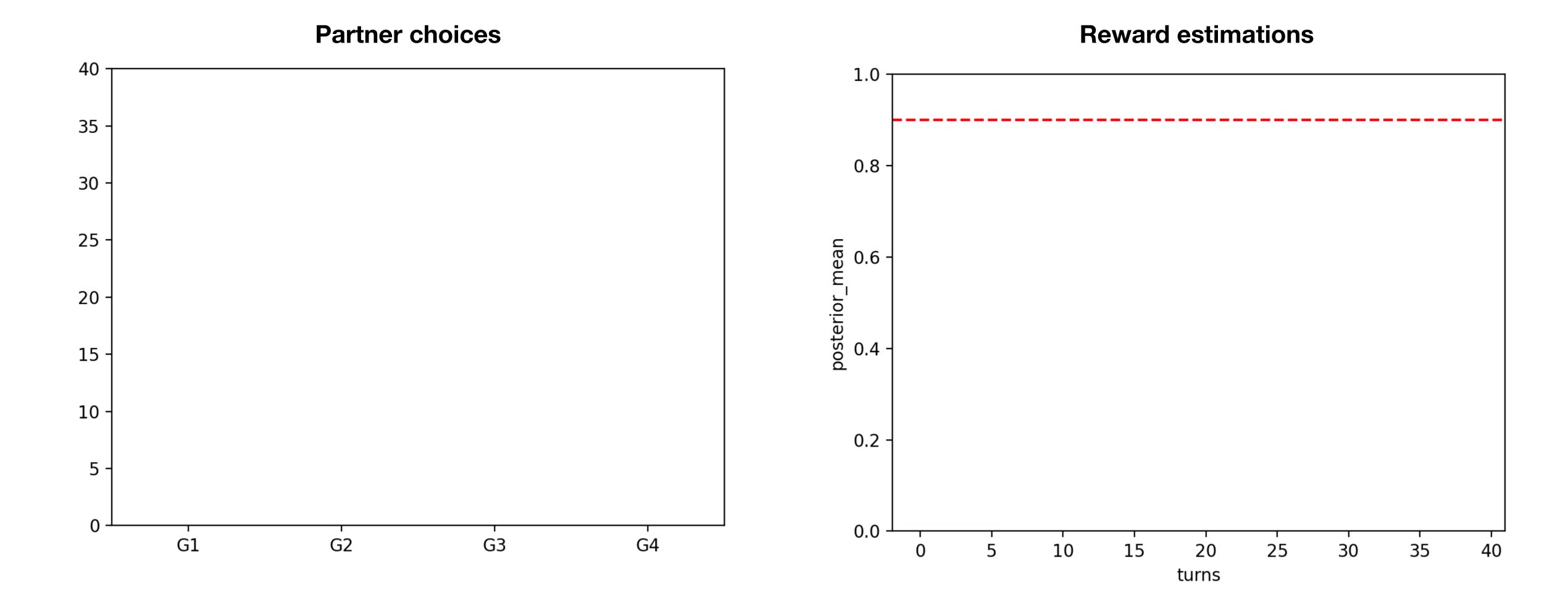
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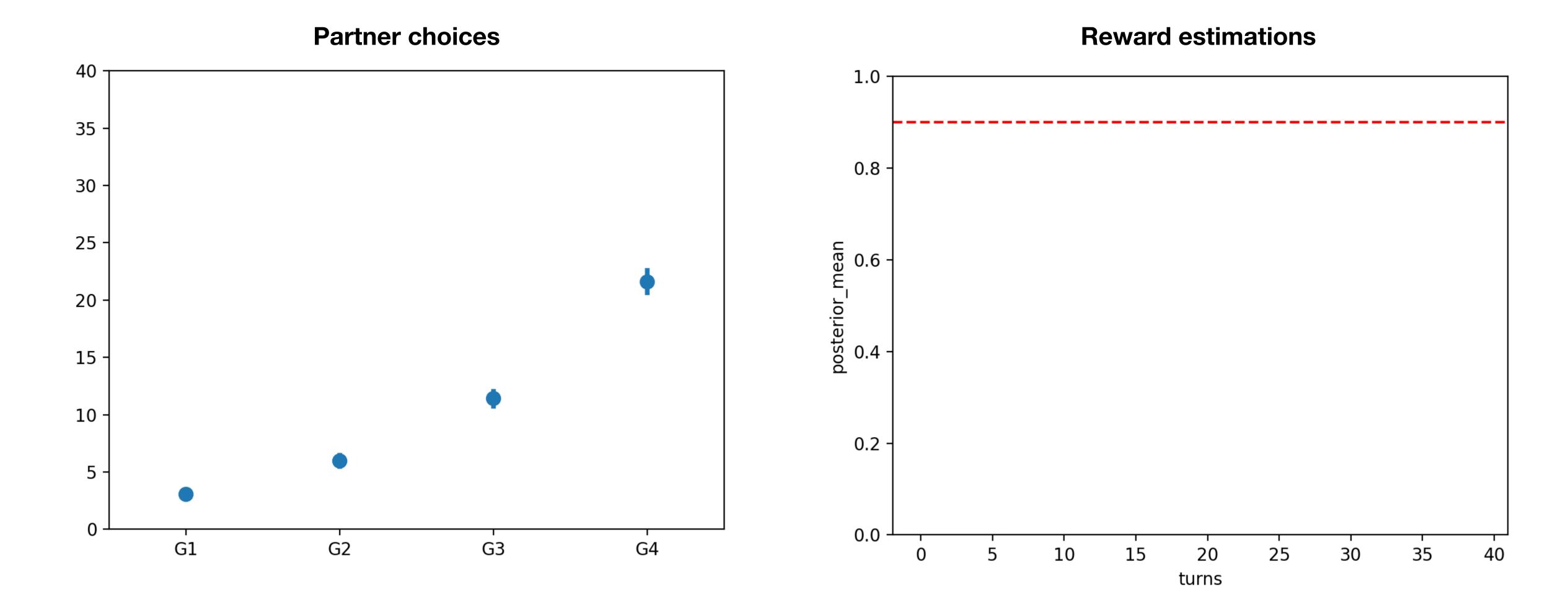
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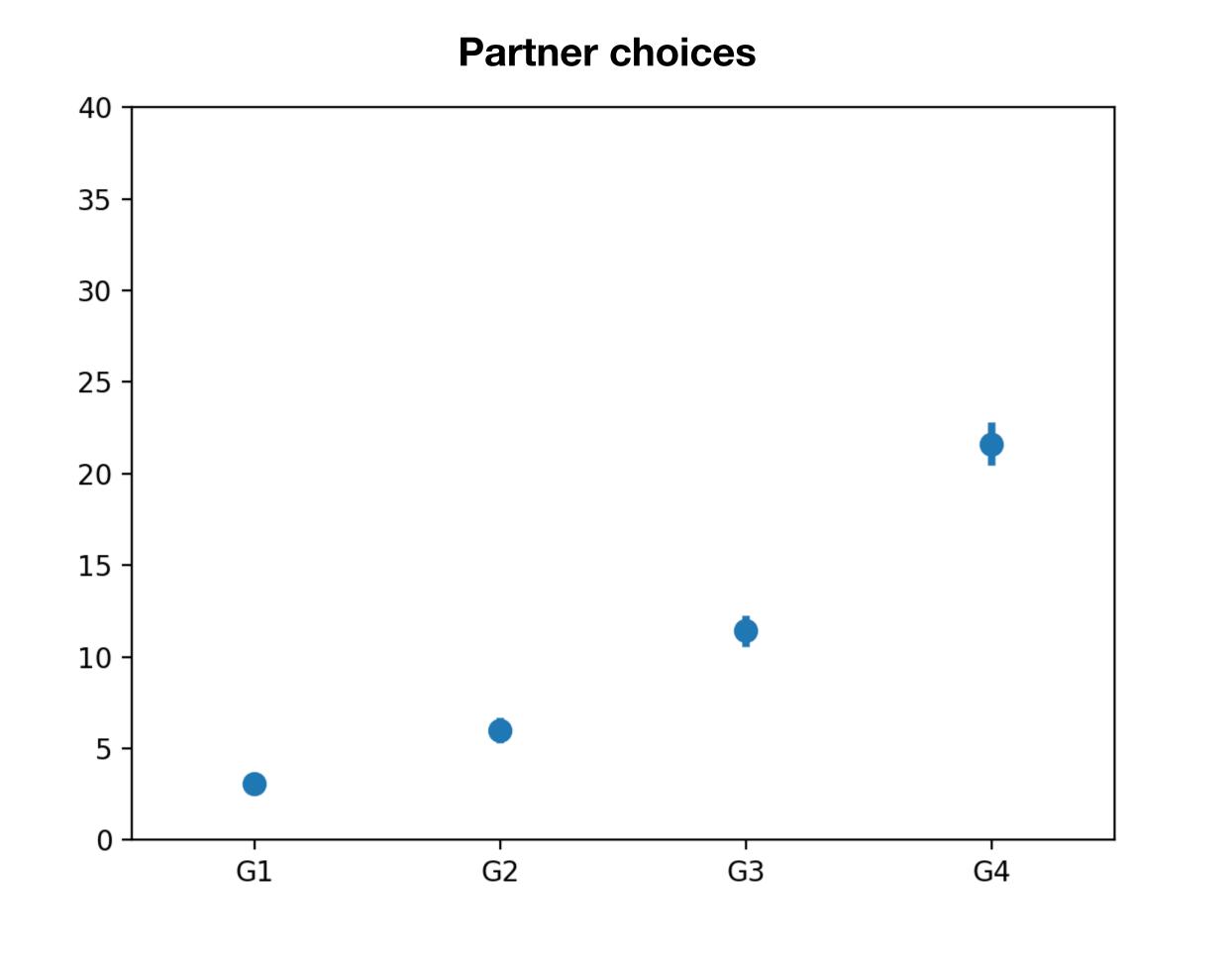
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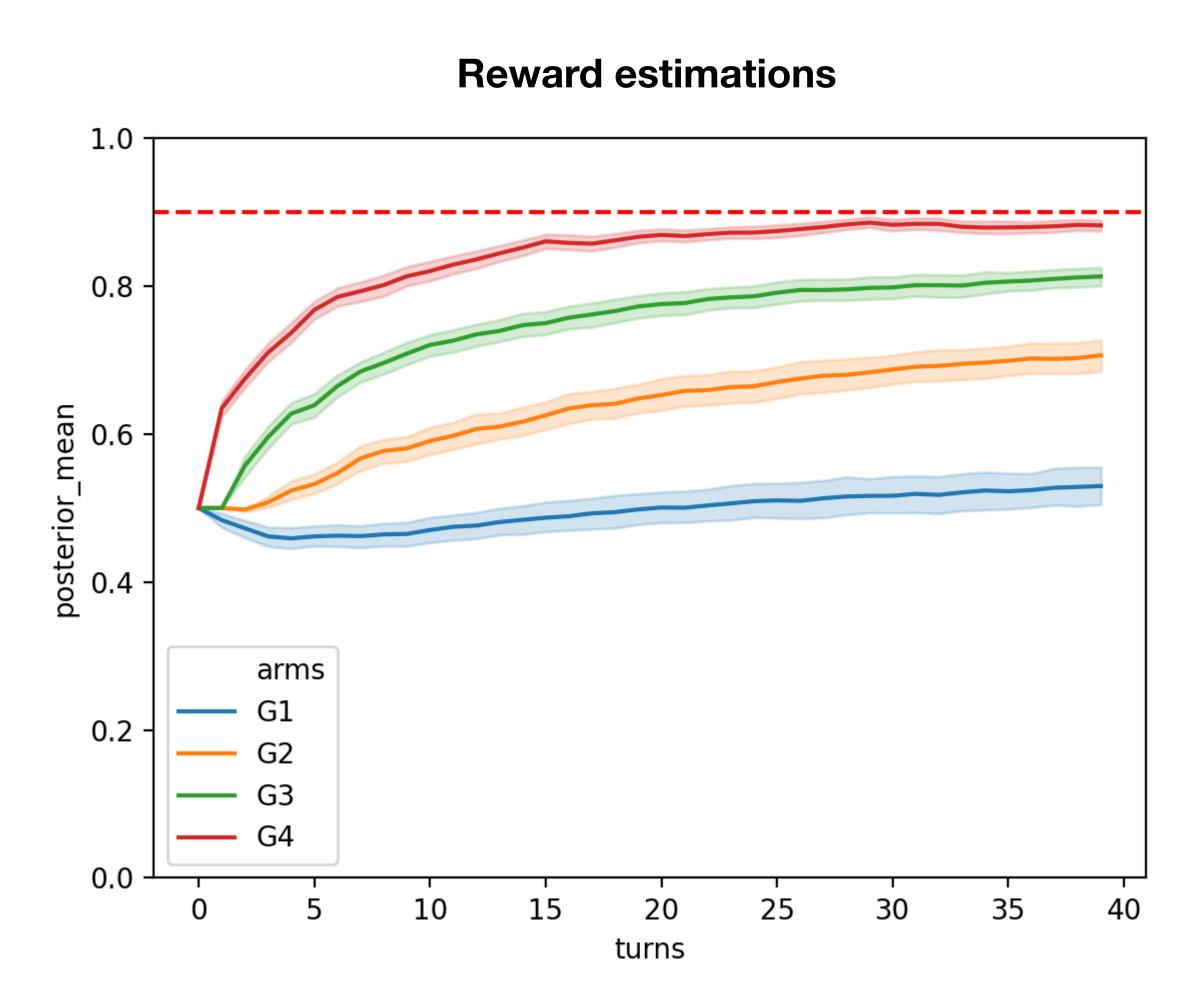


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Selective interactions and biased impressions emerged when

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The environment gives identical and high rewards;

Prior bias is unnecessary;

The agent uses adaptive sampling strategies

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How about human participants?

Globally inaccurate stereotypes can result from locally rational exploration

- An intuitive example
- A formal model and simulations
- Two human experiments
- Implications



A city with 100,000 residents

Ready to play?

Let's meet some new Tufas, Aimas, Rekus, and Wekis!

Tufa



X.



Weki





Independent variables:

- Underlying rewards:
 - Unbeknown to participants, identical ($\mu = 0.9$) v. different average rewards
- Prior stereotypes:
 - No description v. Rekus are warm and competent
- Sampling strategy:
 - You choose v. You meet [random]

Dependent variables:

Partner choices:

You choose a [group] vs. You meet a [group]

Reward estimations:

 For each group, how many times out of 100 do you think working with a person from that group would result you in earning 1 point?

400 online participants in Study 1 (N = 2000 in Study 2)

Partner choices

reward = identical & bias = no

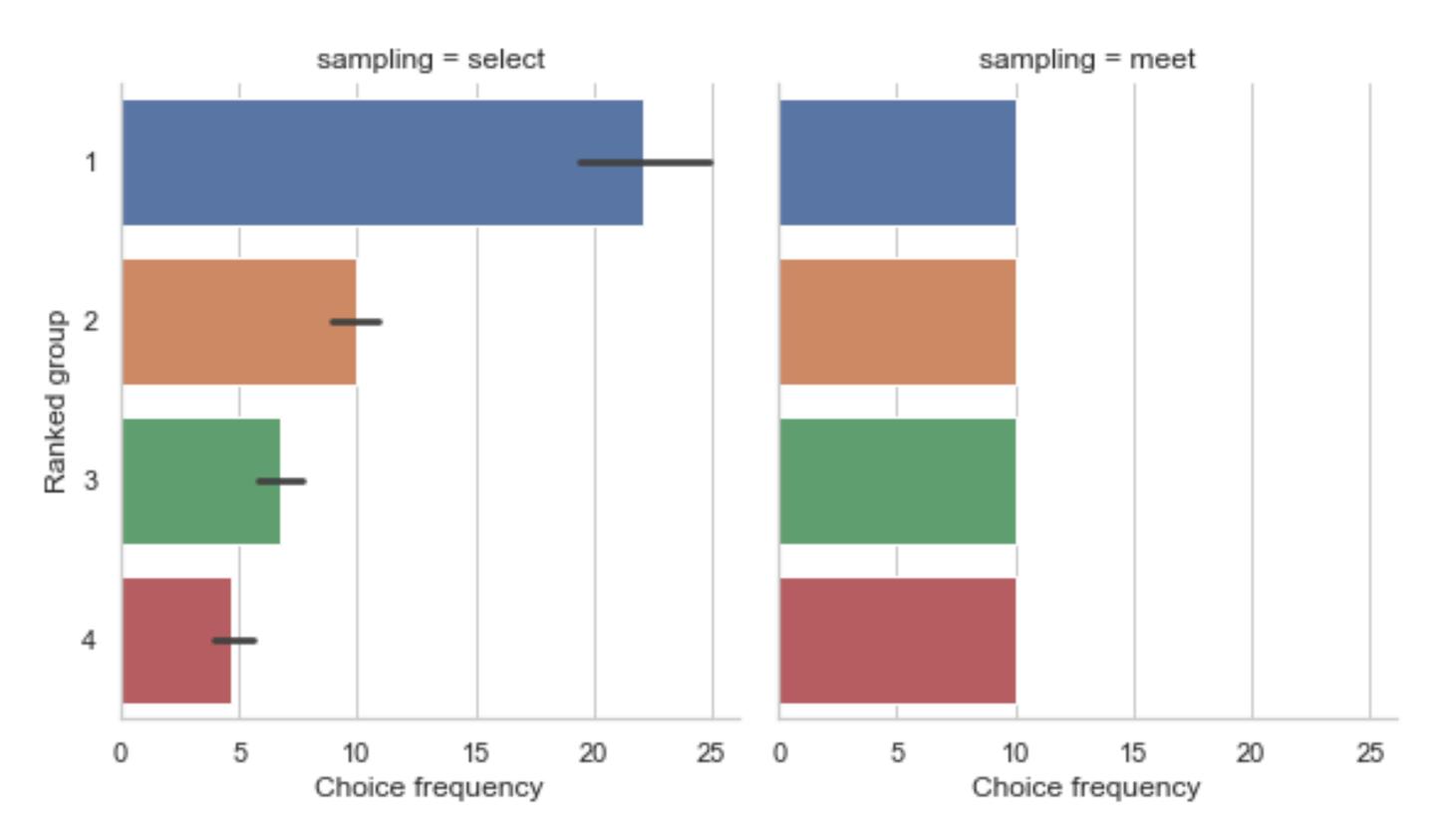
sampling = meet 20 25 5 Choice frequency

Reward estimations

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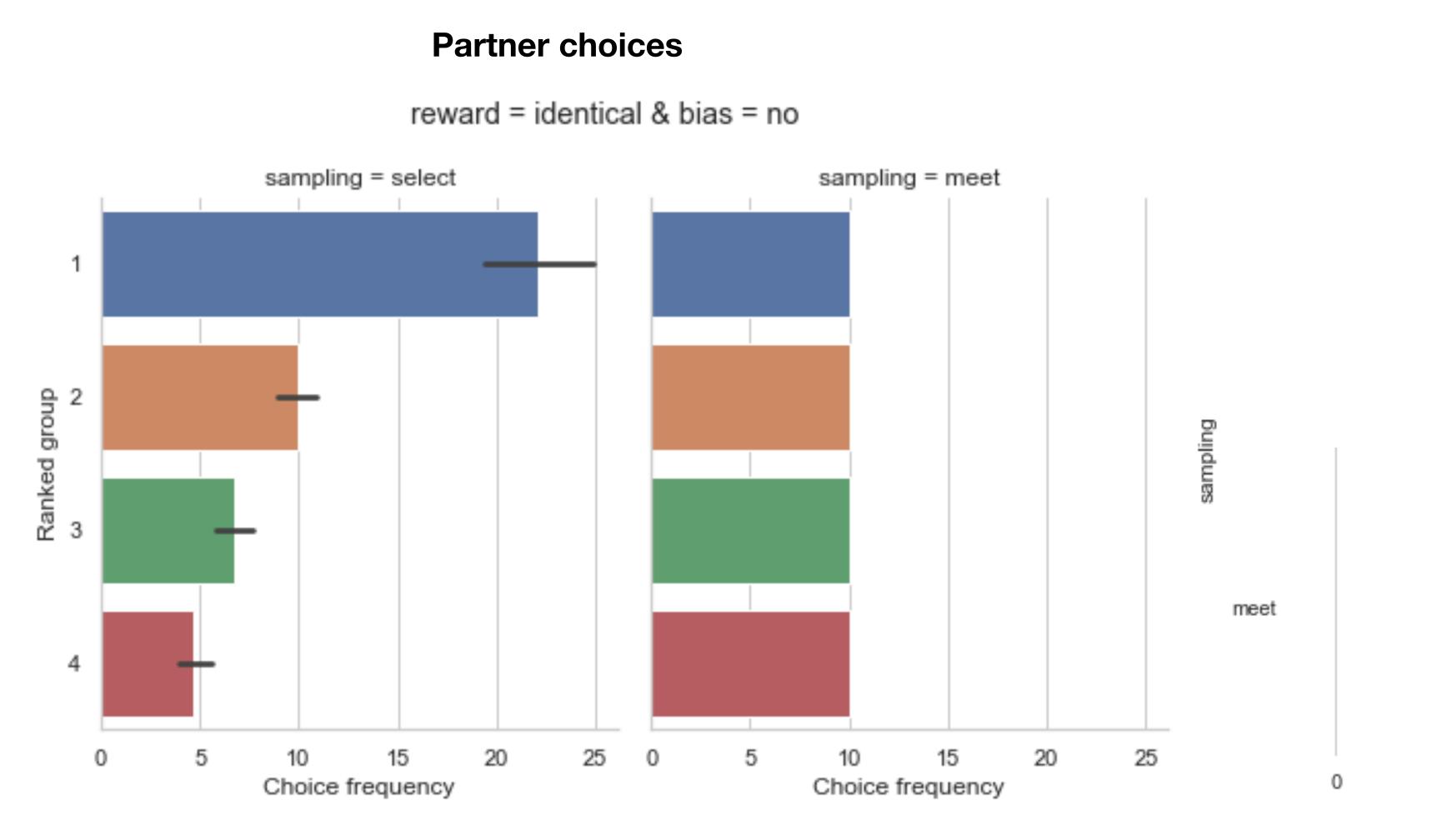
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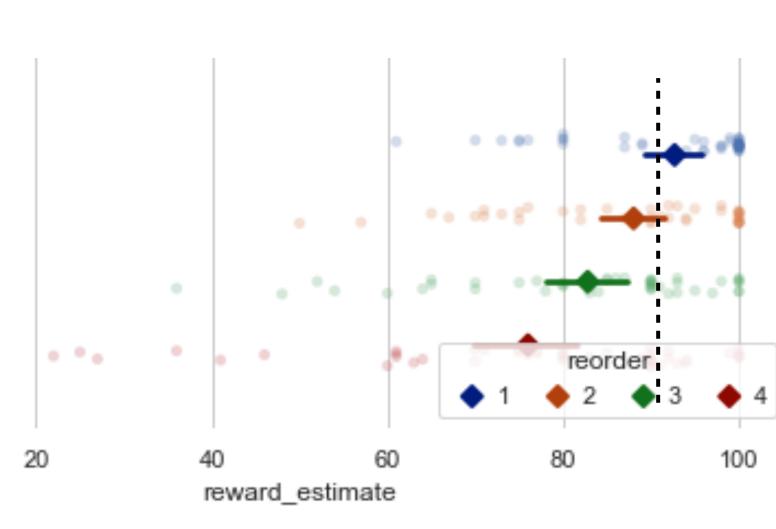
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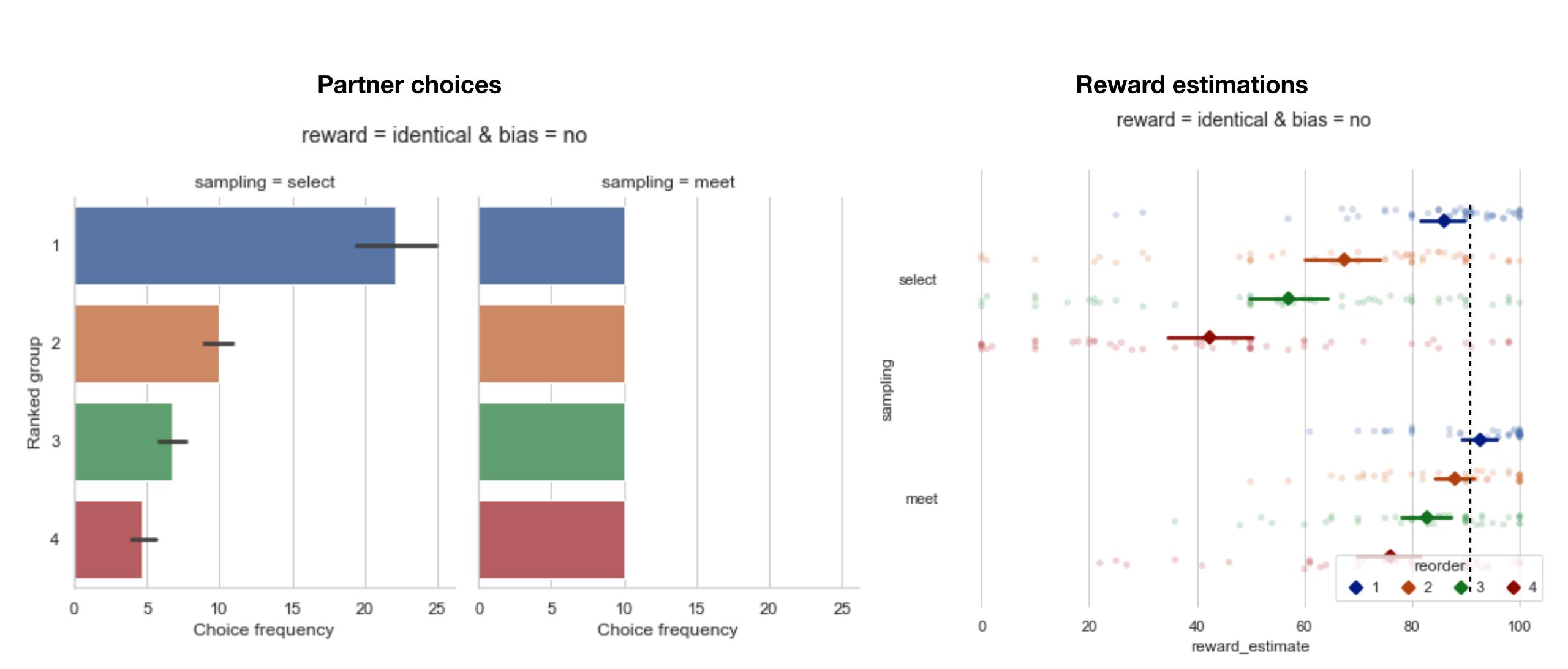


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How do humans behave in an environment with identical and high rewards & no prior bias

- Replicated model predictions.

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Selective interactions and biased impressions?

- Yes.

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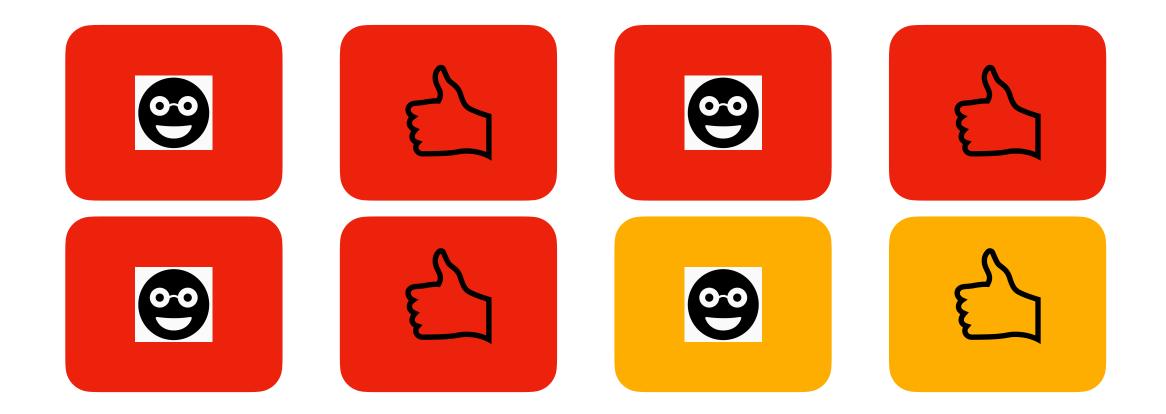
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Thompson sampling v. Random sampling

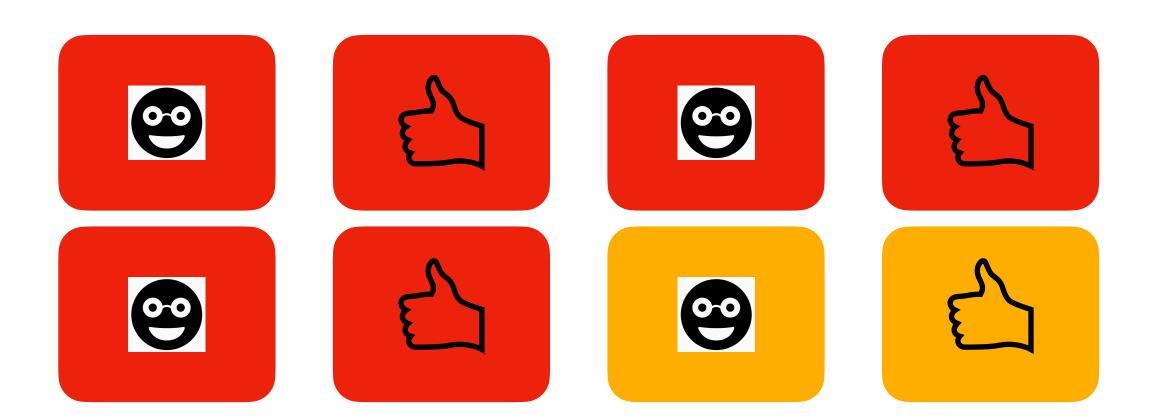
- Consistent with rational strategy thus biased.

Confounds and Mechanisms



You choose v. You meet [random draw]

Confounds and Mechanisms

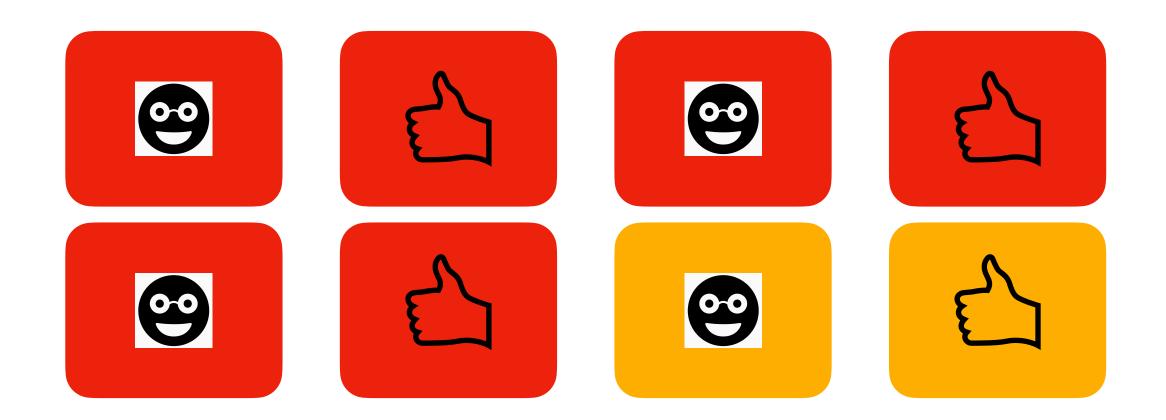


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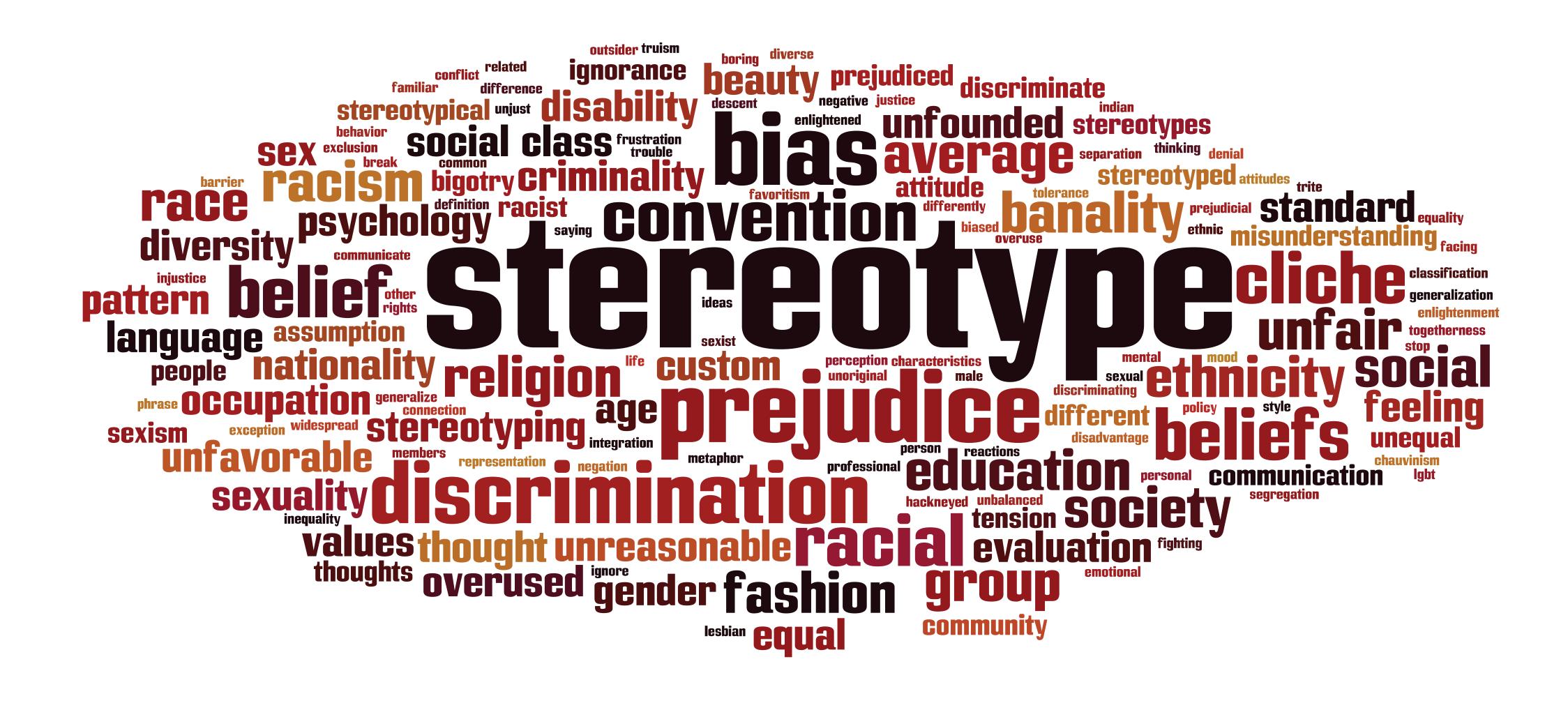
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Conclusion

WHY do people develop inaccurate stereotypes in the first place?



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- We offer one minimal condition that inaccurate stereotypes can develop.
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Conclusion

WHY do people develop inaccurate stereotypes in the first place?

- We offer one minimal condition that inaccurate stereotypes can develop.
 - Without requiring group motives or cognitive limits.
- A functional minimal-process explanation.
 - The mind adapts to a particular social environment.
 - Rational local choices still lead to inaccurate overall impressions.

Implication

Inaccurate stereotypes can result from rational exploration

- Social interventions.
 - Diversity: Habituate to de-segregated environment.
 - Contact: Encourage open-minded exploration.

Bai, Ramos, Fiske, 2020, As diversity increases, people paradoxically perceive social groups as more similar, PNAS

Implication

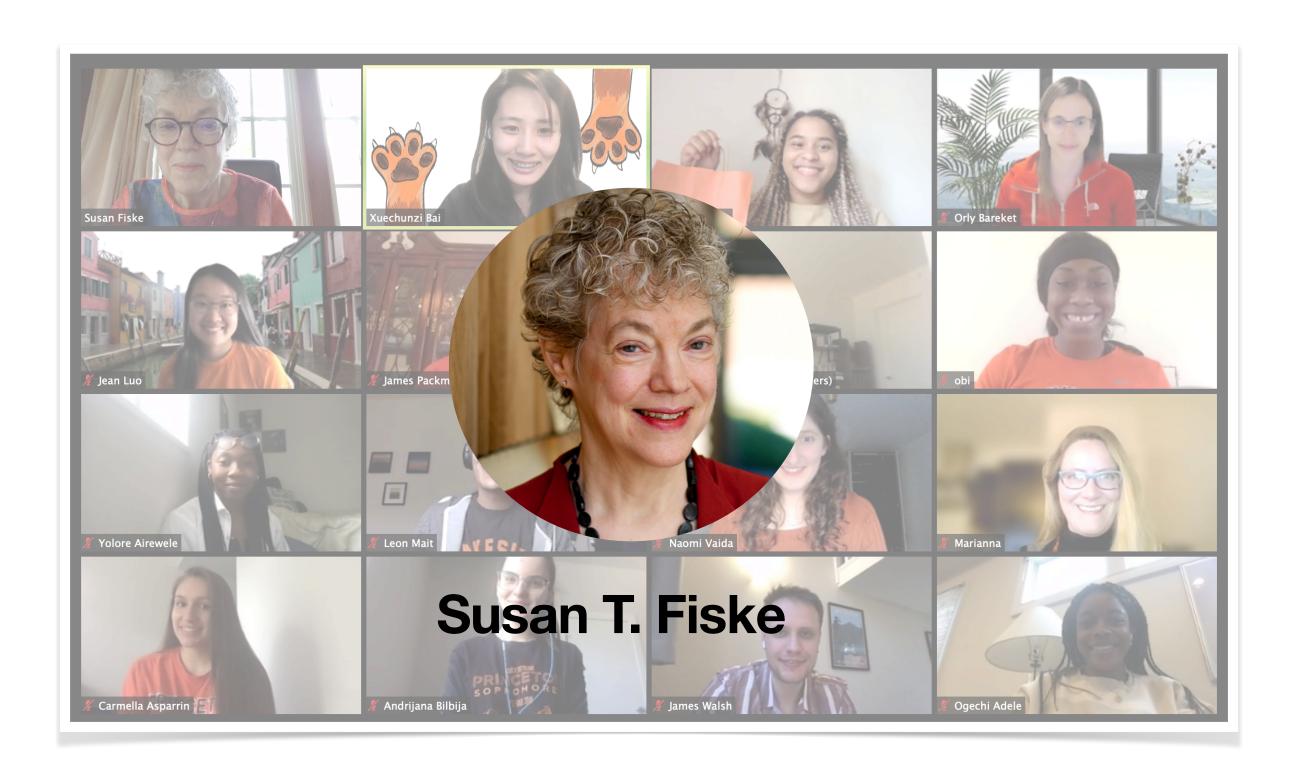
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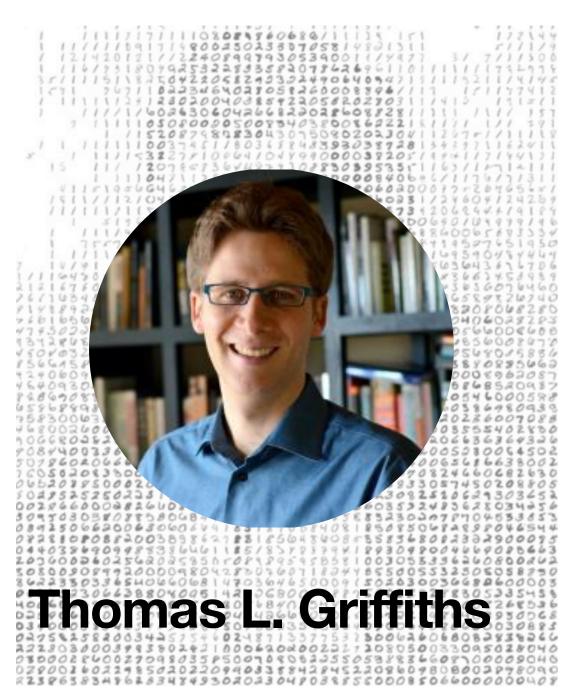
- Social interventions.
 - Diversity: Habituate to de-segregated environment.
 - Contact: Encourage open-minded exploration.
- Algorithmic interventions.
 - ML fairness: The origin of unequal base rates.
 - Filter bubble: Recommendation systems.





Thanks to my advisors for collaboration and labs for feedback





Thank you!

Our hypothesis

Inaccurate stereotypes can result from rational explorations

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Multi-Armed Bandit Problem

Explore v. Exploit dilemma in Reinforcement Learning (Sutton & Barto, 2018)



$$\theta_k$$

$$r_{t(k)} \sim Bern(\theta_k)$$

$$\sum_{t=1}^{T} r_{t(k)}$$

$$R = E\left[\sum_{t=1}^{T} Q_{t(k^*)} - \sum_{t=1}^{T} r_{t(k)}\right]$$

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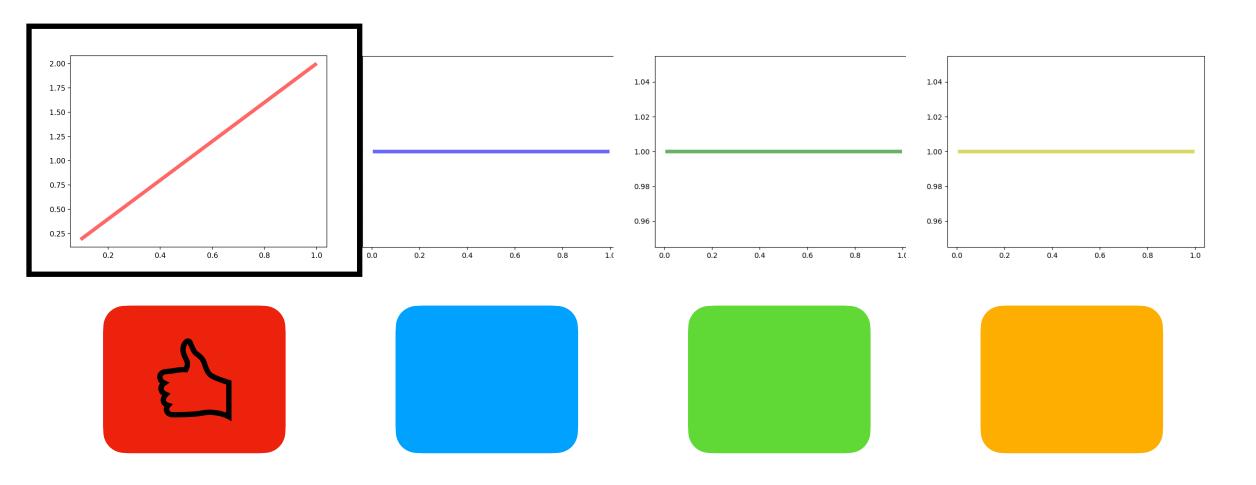
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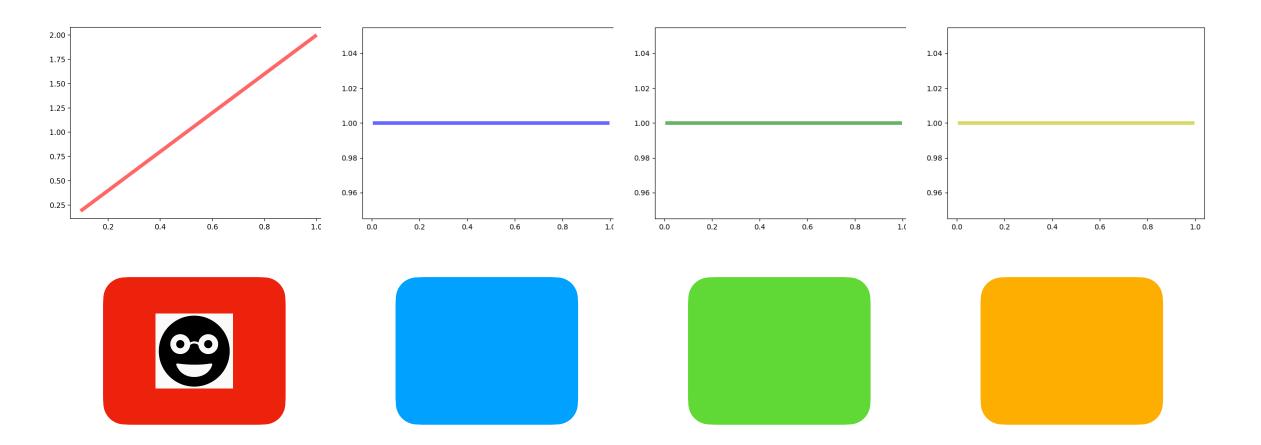
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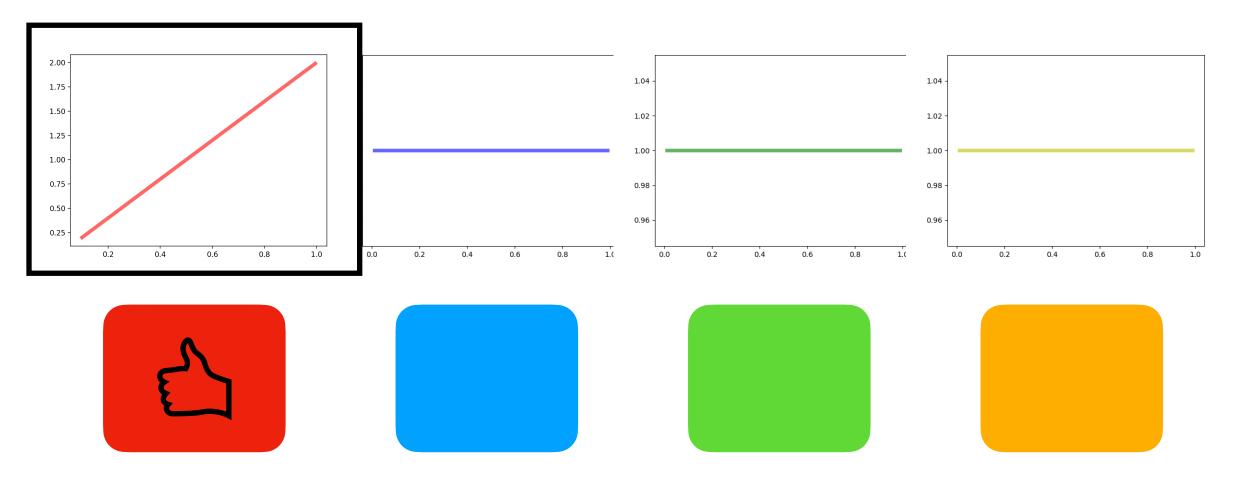
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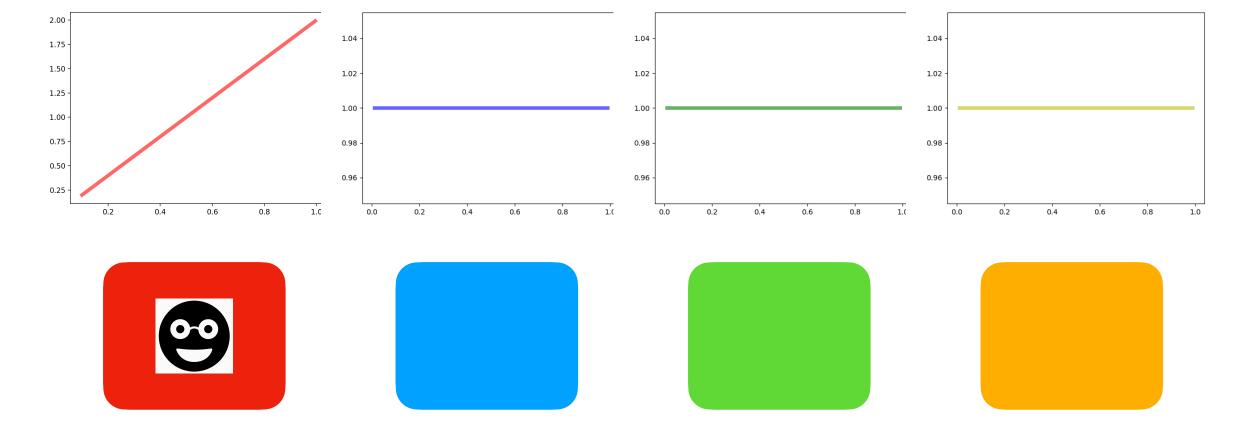
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Thompson Sampling



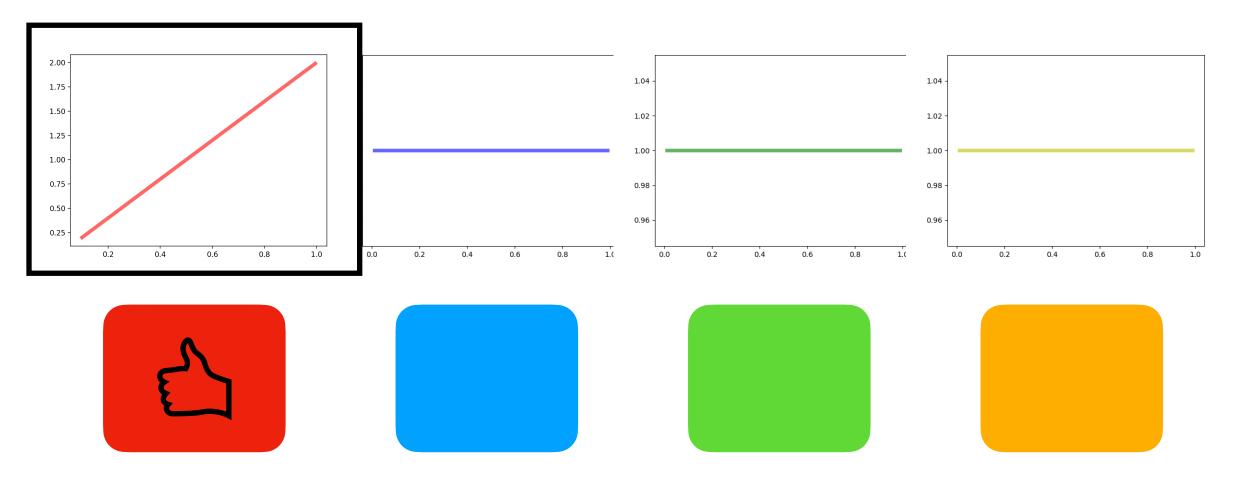


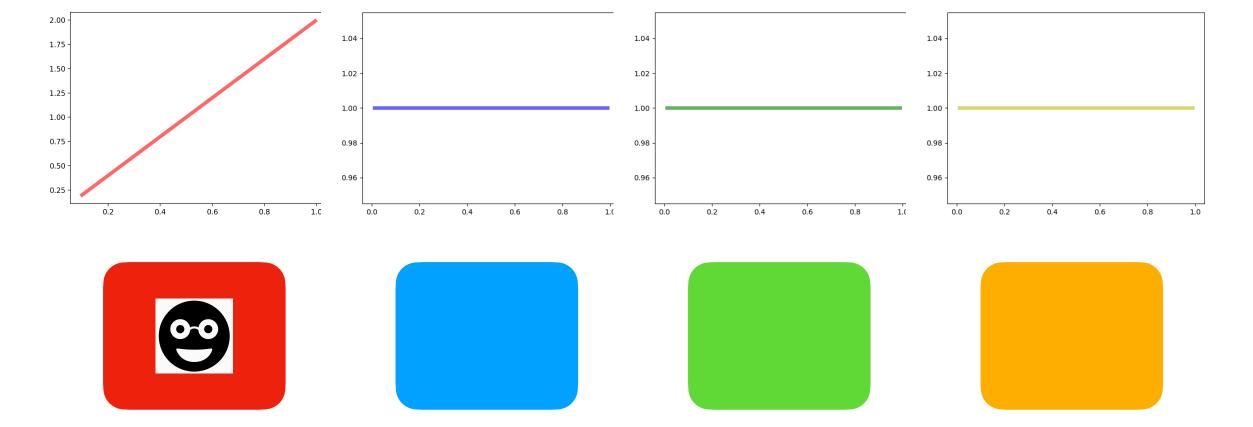
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$$E = \frac{\alpha}{\alpha + \beta}$$

$$S_{k(t)}, F_{k(t)}$$

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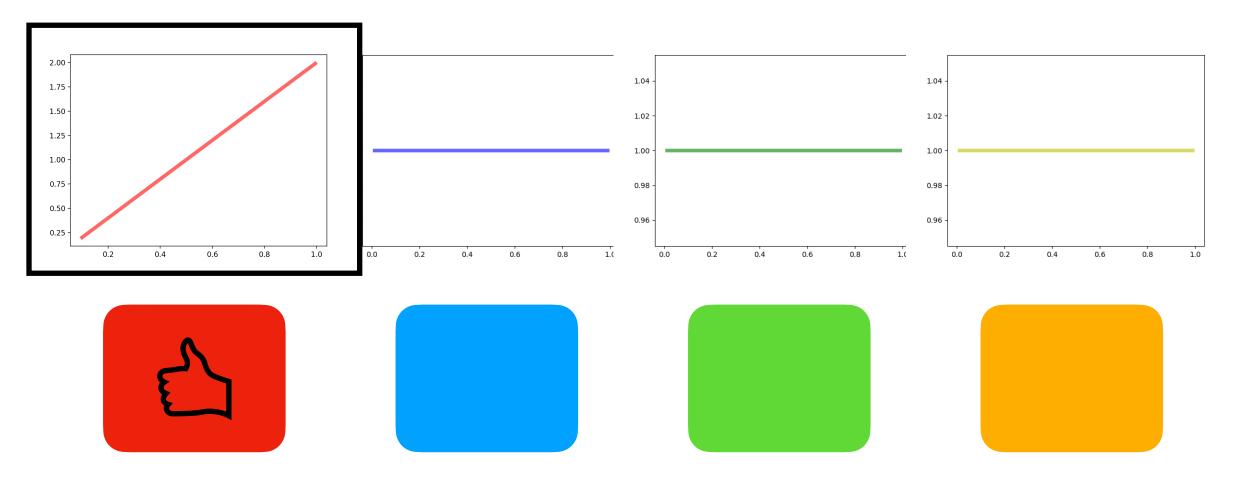


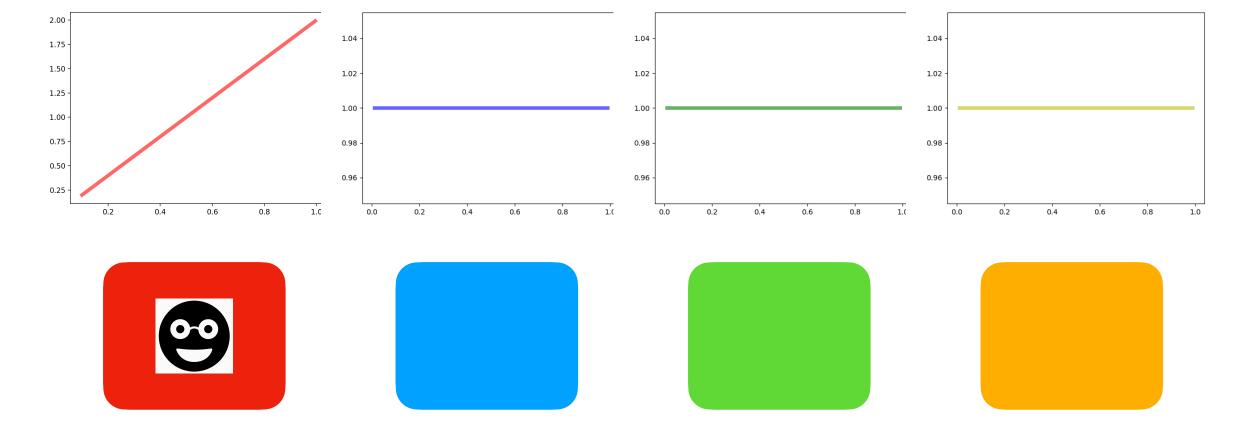
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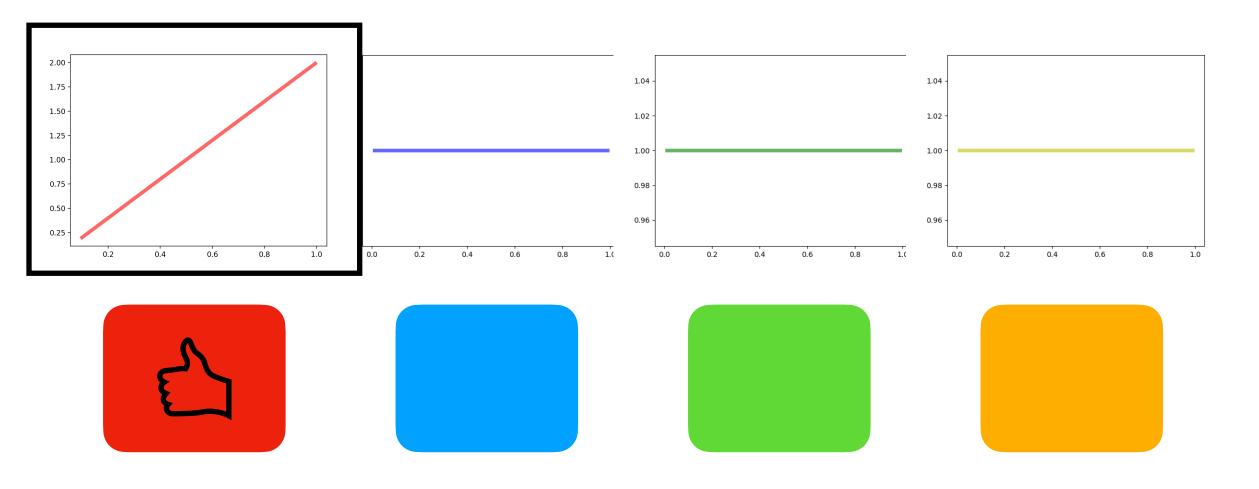


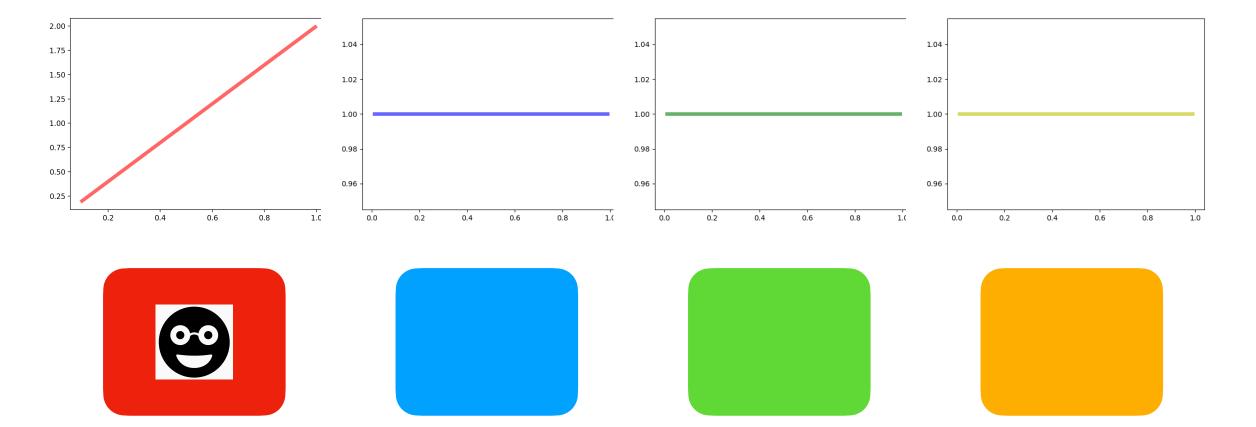
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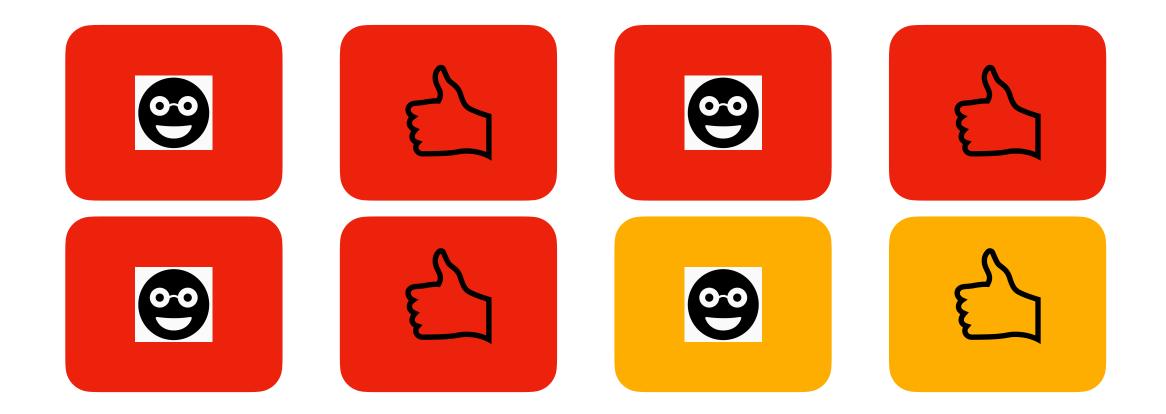
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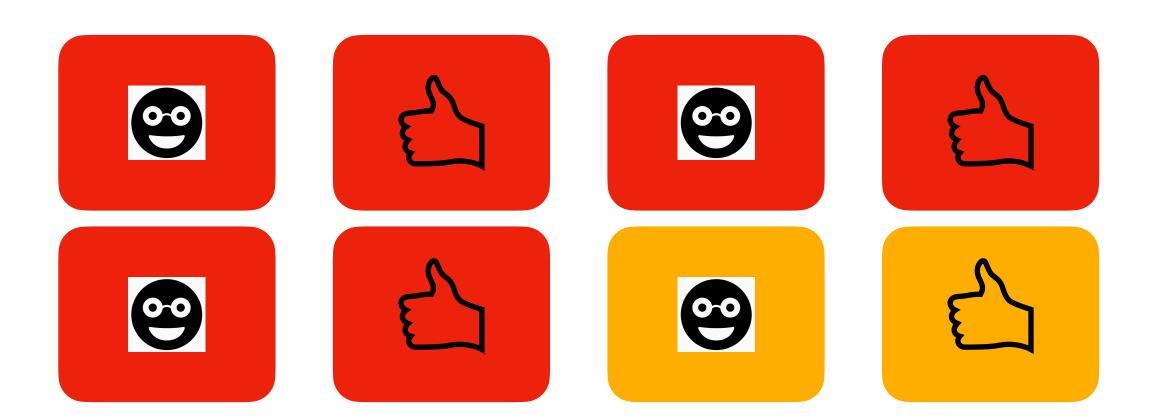


Confounds and Mechanisms



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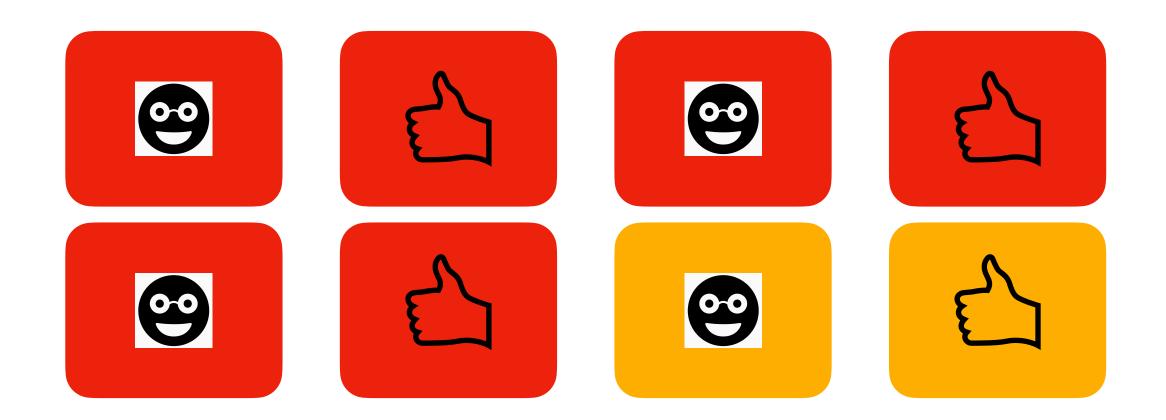


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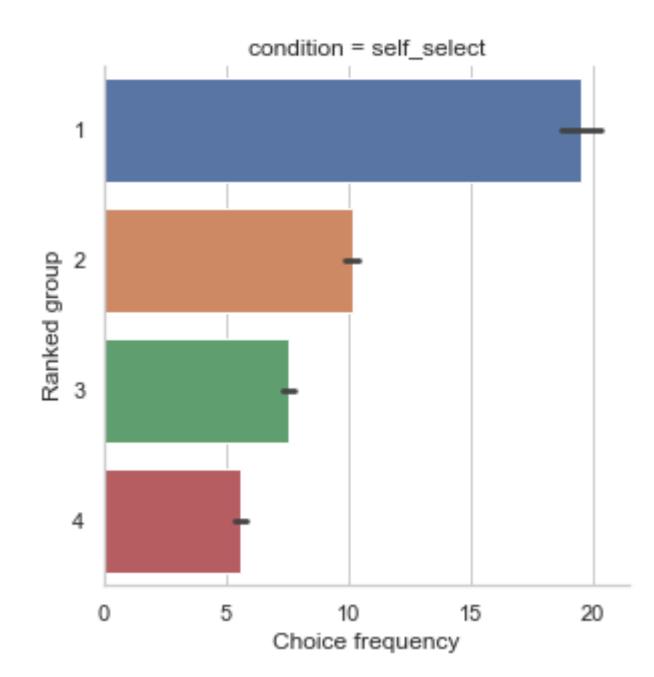
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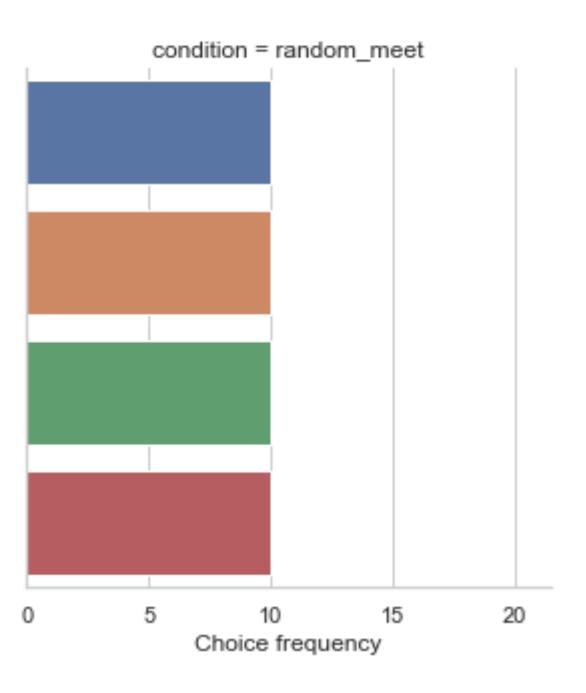
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Partner choices: replication

2000 online participants (500 in each 1 by 4)

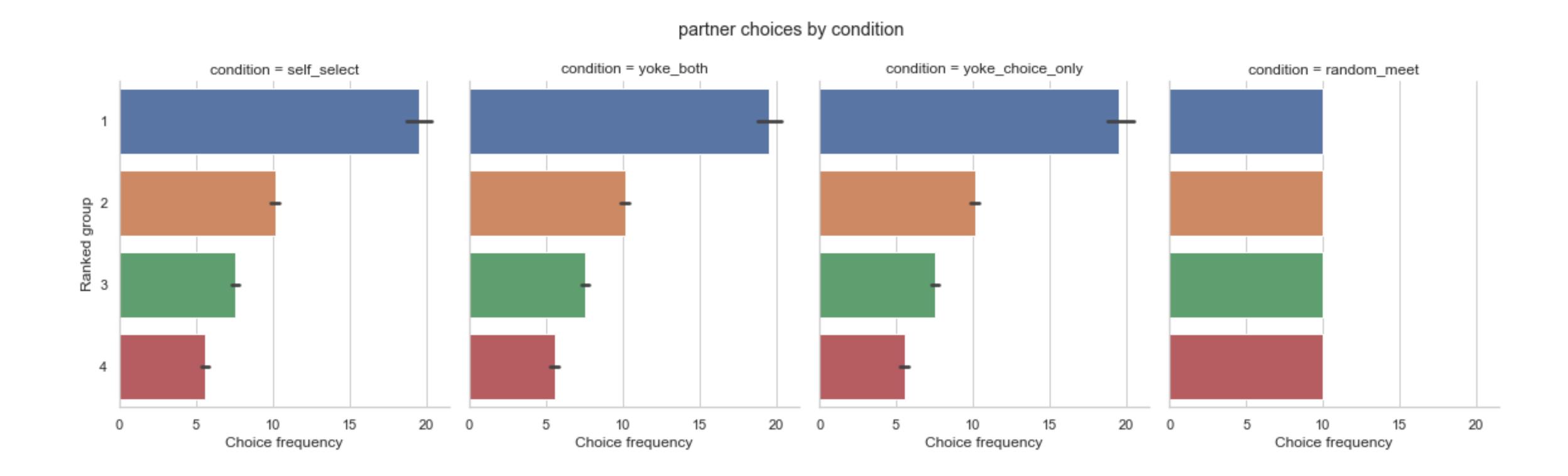
partner choices by condition





Partner choices: yoke

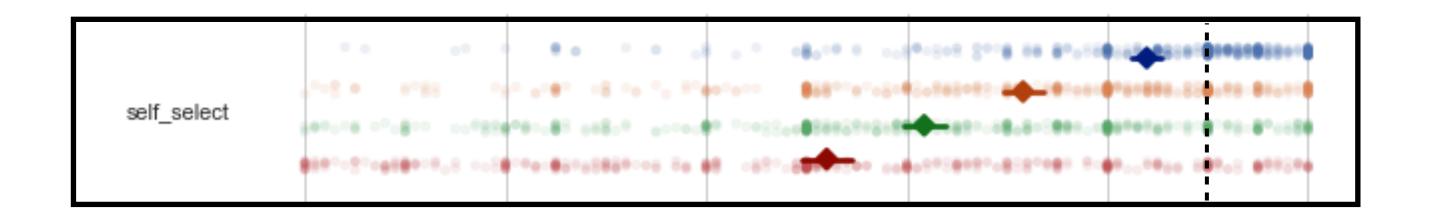
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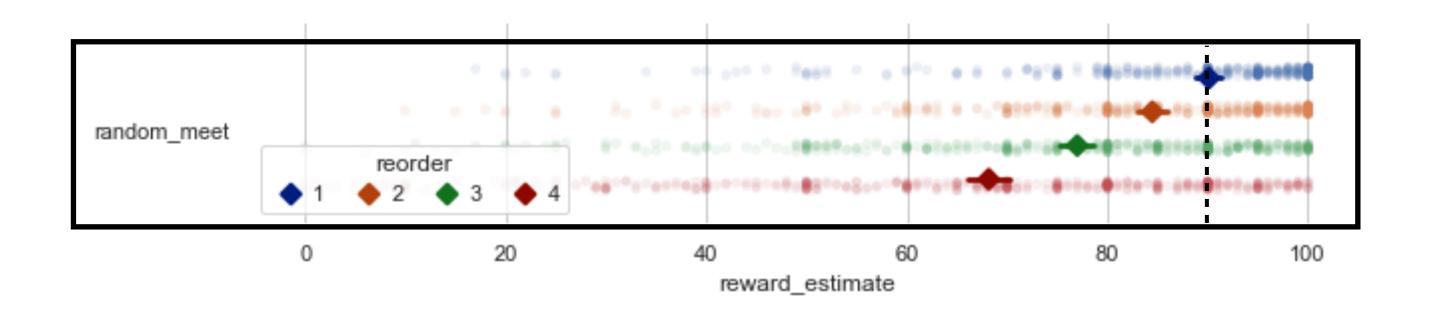
Reward estimations: replication

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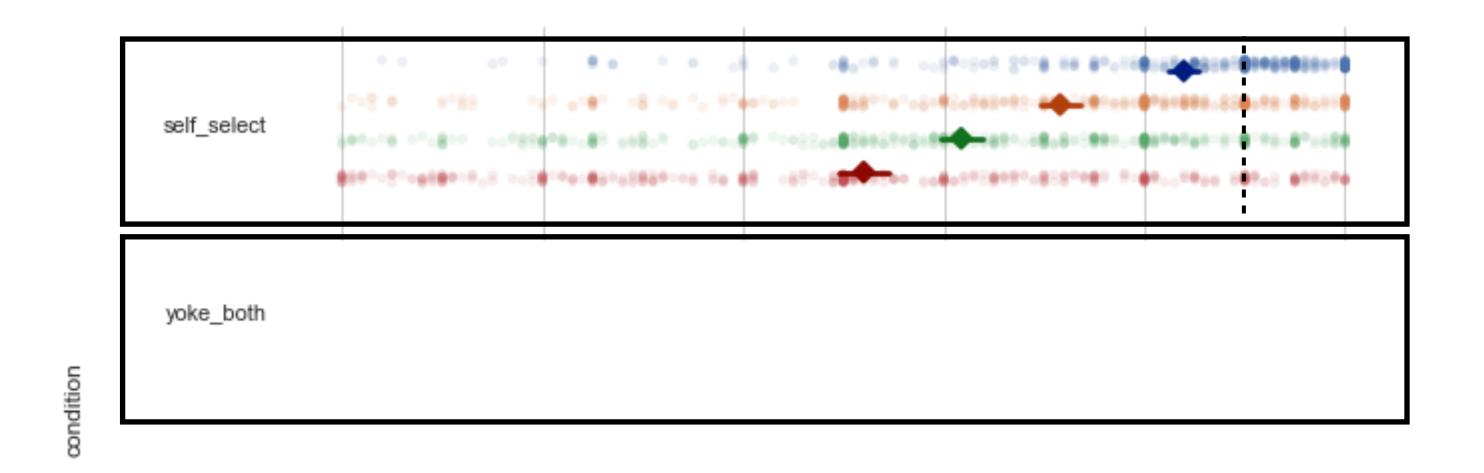
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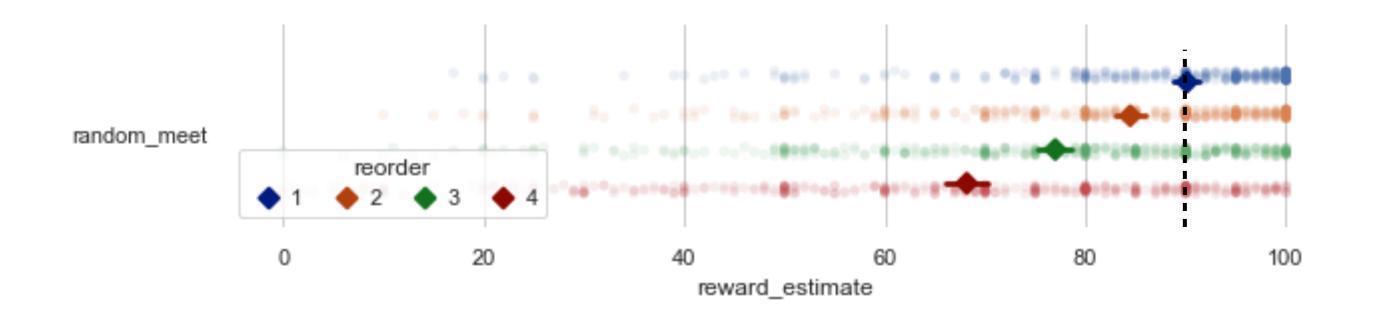


Reward estimations: selective sample or sense of control

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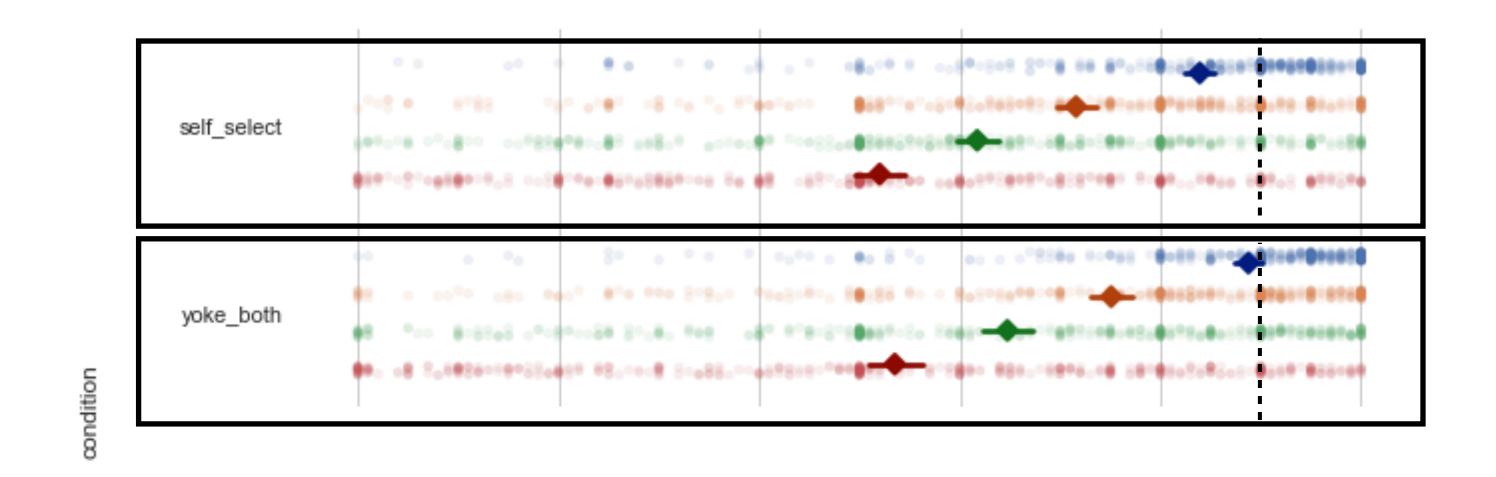


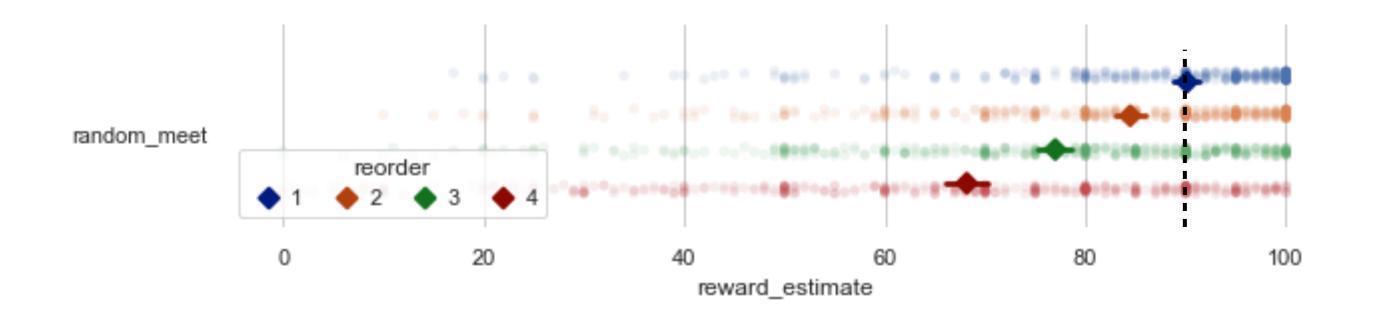


Reward estimations: selective sample or sense of control

2000 online participants (500 in each 1 by 4)

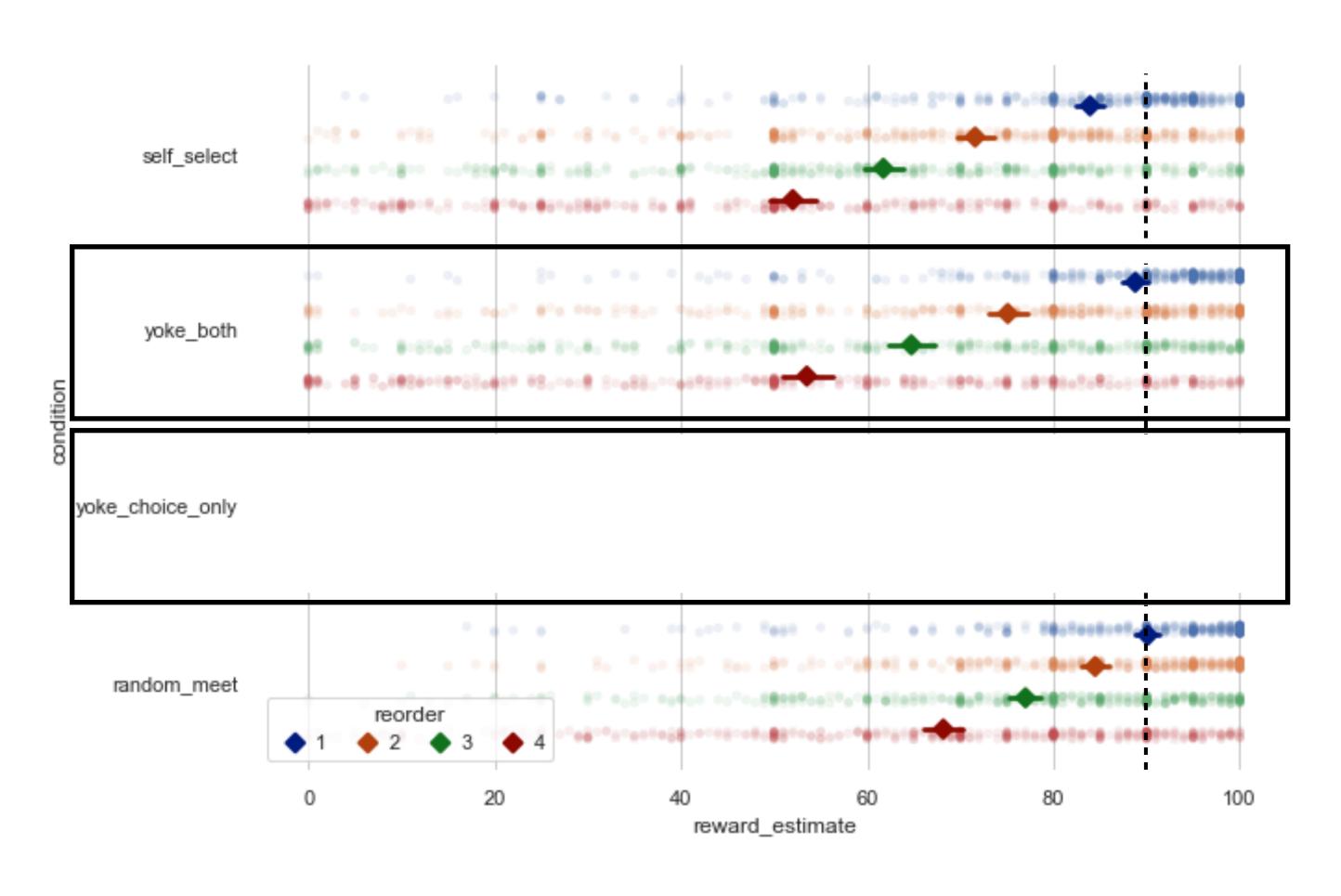
reward estimates by condition





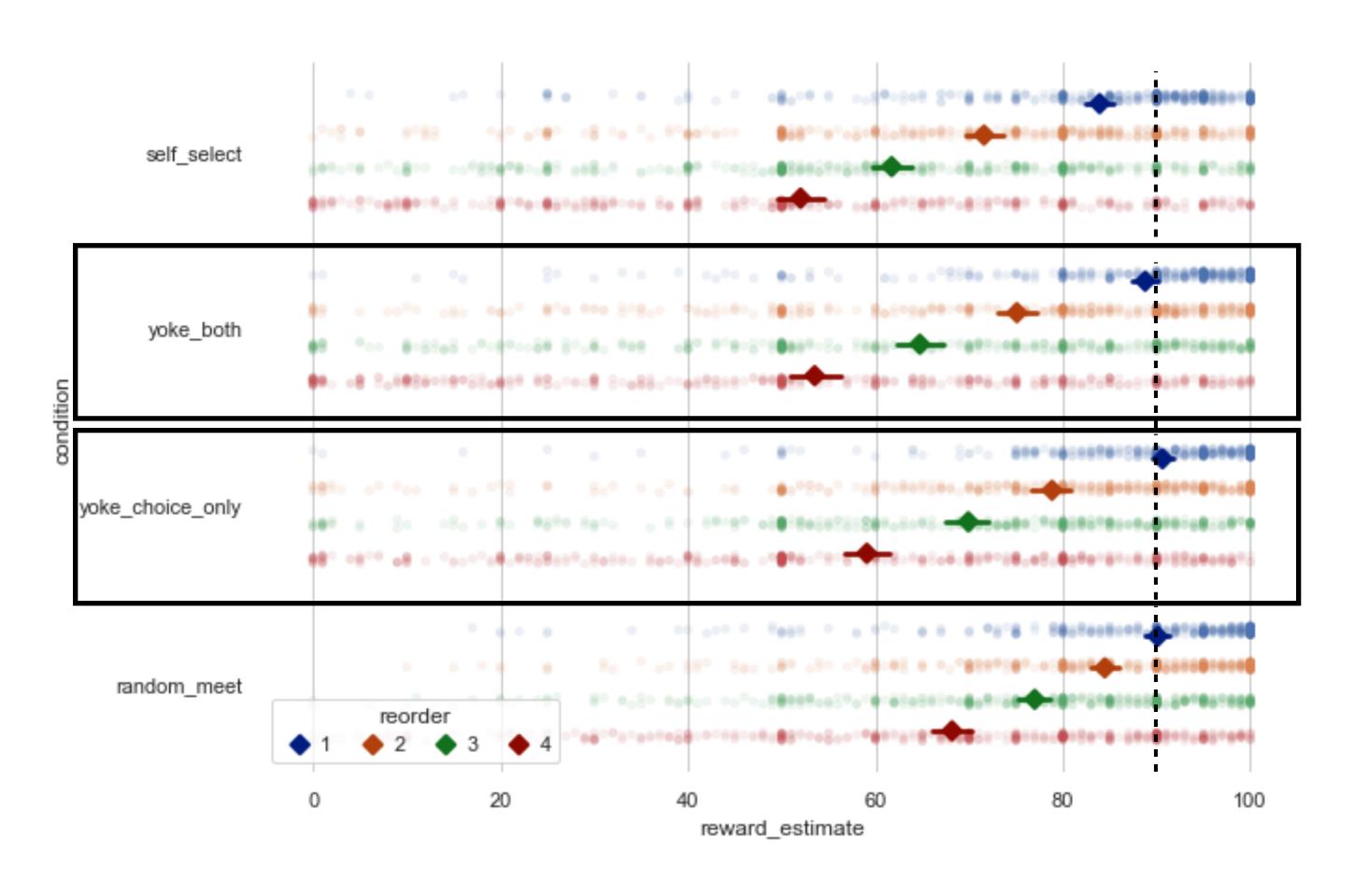
Reward estimations: order of rewards or mere presence of choices 2000 online participants (500 in each 1 by 4)

reward estimates by condition



Reward estimations: order of rewards or mere presence of choices 2000 online participants (500 in each 1 by 4)

reward estimates by condition



Interim summary

Replicating prior experiment?

- Yes

Interim summary

Replicating prior experiment?

- Yes

Selective samples or sense of control?

- Selective samples

Interim summary

Replicating prior experiment?

- Yes

Selective samples or sense of control?

- Selective samples

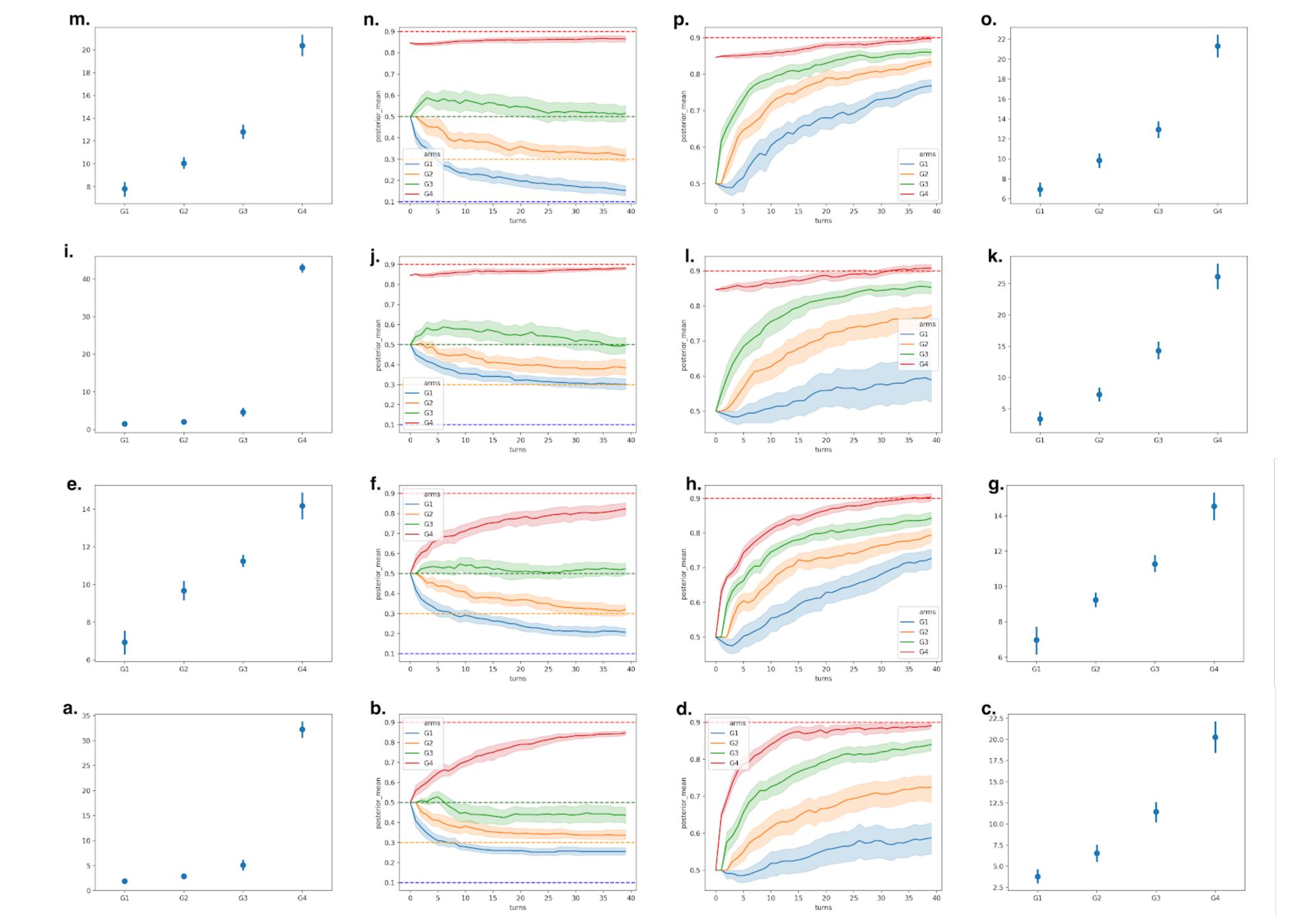
Order of rewards or mere presence of choices?

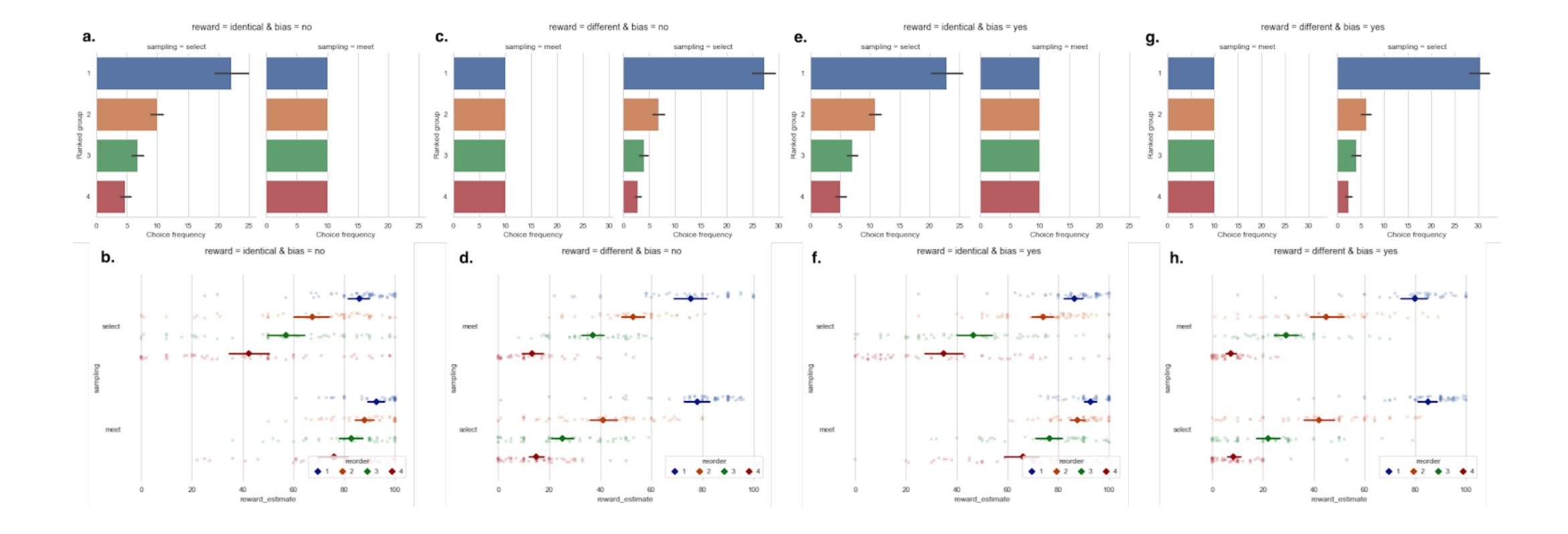
- Order of rewards

Appendix

Related works

- Denrell, Experience sampling
- Gureckis, Learning trap
- Fiedler, Reward rich





Analytic solutions

Dynamic programming

