

### The Standard Game

	guess	result
experiment 1:	R B G Y	● ○
experiment 2:	P O G Y	●
experiment 3:	G Y G Y	—
experiment 4:	R B P P	● ○
experiment 5:	R O B O	● ● ○ ○
experiment 6:	R O O B	● ● ● ●

Figure 1: A sample game (adapted from [1])

Explore (and foster) *intuitions* about:

- experimental design (what is a good ‘experiment’?)
- hypothesis testing
- interpretation of results
- effective use of controls

### The Code Jar Variation

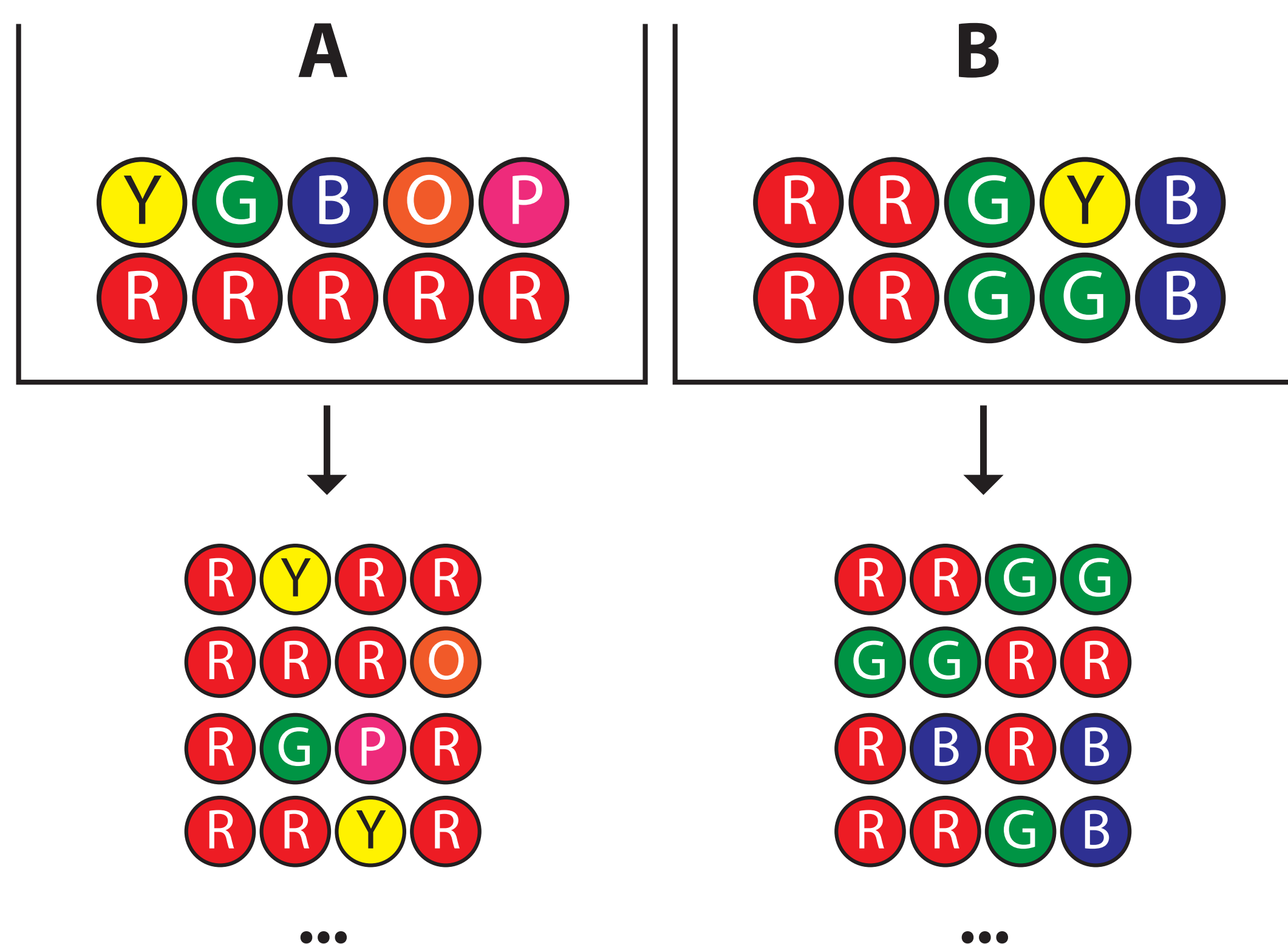


Figure 2: Depicting two different code jars (A and B) and several hidden codes sampled with replacement from the respective jar. Which jar would be easier to play?

### Entropy-based Strategies

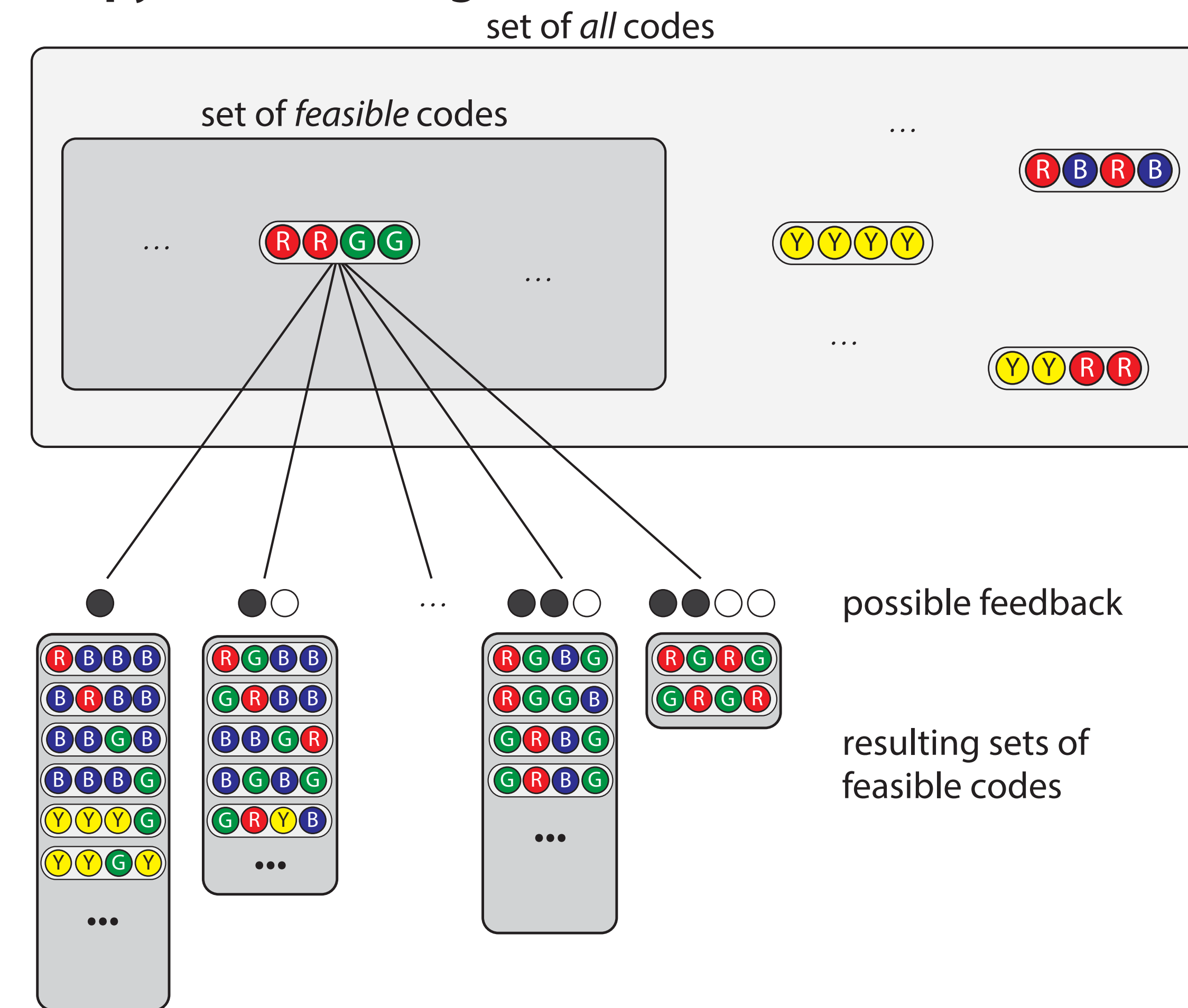


Figure 3: Generic, entropy-based strategy to determine the value (expected information gain) of an experiment.

Pick code with highest **expected information gain**:

$$\text{prior uncertainty} - (\text{expected}) \text{posterior uncertainty}$$

Sharma-Mittal provides a unified mathematical framework to quantify uncertainty via notion of **entropy** (= expected surprise):

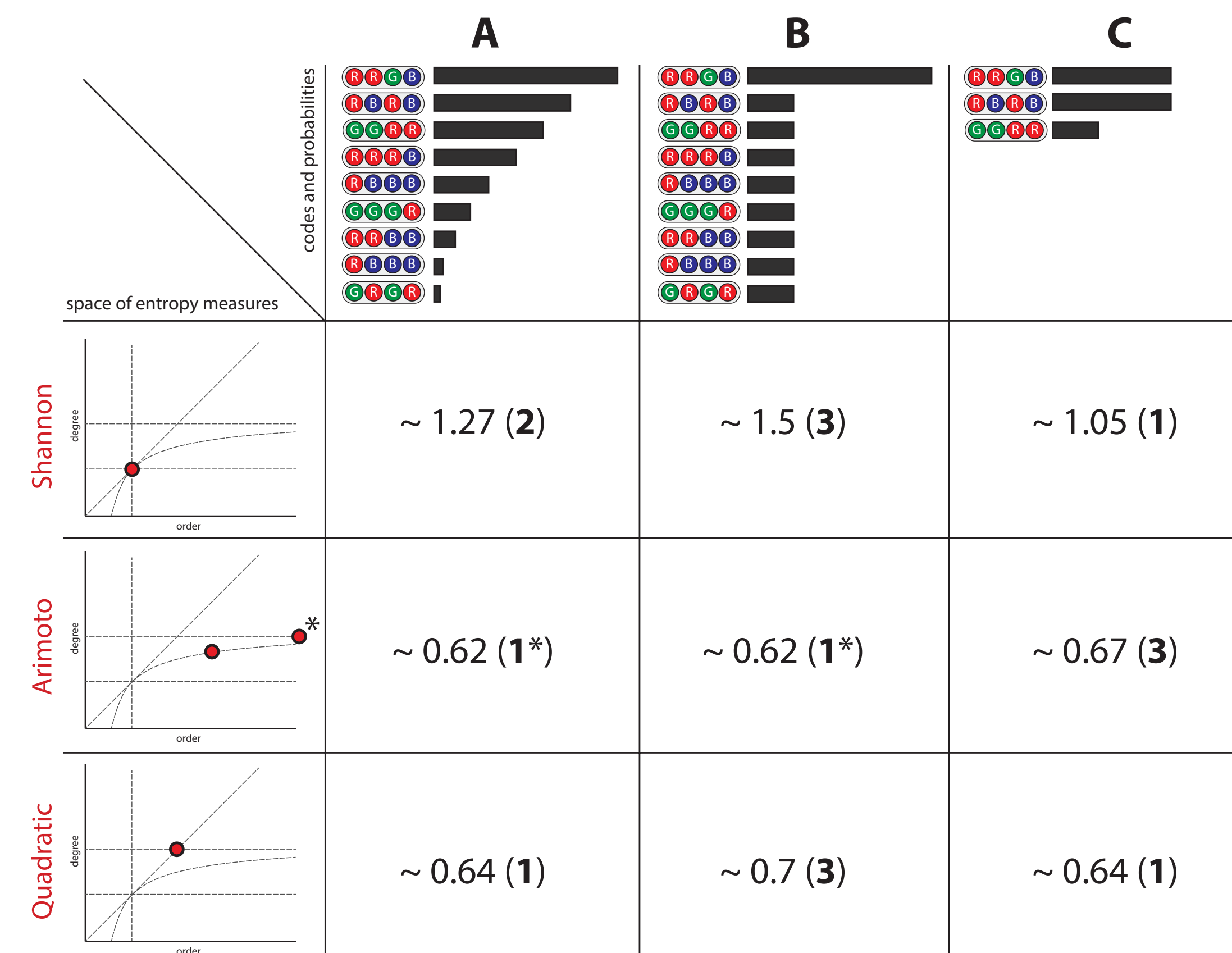


Figure 4: Entropies (in nats) of different probability distributions according to three different entropy measures: Shannon entropy, Arimoto, and Quadratic.

### Planned computer simulations and experiments

(P I) How do strategies based on information gain measures from [3] differ in terms of psychologically plausible **meta-features** (e.g., position, color count, number of different colors, ...)?

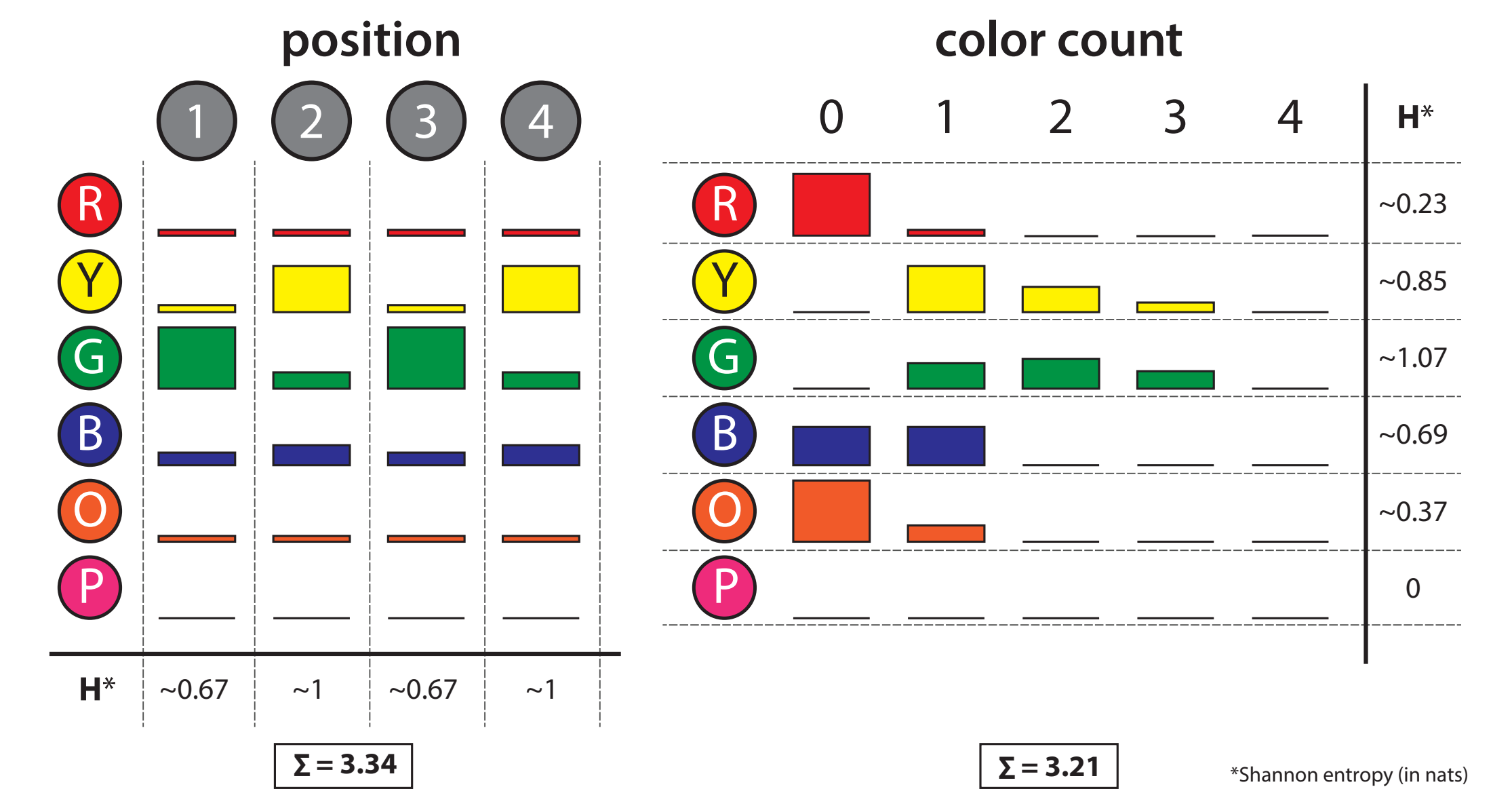


Figure 5: A selection of meta-features after the first guess.

(P II) Can we use **adaptive experimental design** techniques [4] to find out which information gain measures best describe human intuitions?

You're playing with the following code jar



You receive the following feedback for your first guess:

guess 1: R R G G ● ● ○

which of these guesses would you like to make next?



Figure 6: Planned human subject experiment (2-AFC task).

### References

- [1] Strom, A. R., & Barolo, S. (2011). Using the Game of Mastermind to Teach, Practice, and Discuss Scientific Reasoning Skills. *PLoS Biol*, 9(1), e1000578.
- [2] Cotta, C., Guervós, J. J. M., García, A. M. M., & Runarsson, T. P. (2010, September). Entropy-driven evolutionary approaches to the mastermind problem. In *International Conference on Parallel Problem Solving from Nature* (pp. 421-431). Springer Berlin Heidelberg.
- [3] Crupi, V., Nelson, J. D., Meder, B., Cevolani, G., & Tentori, K. (2016). Generalized information theory meets human cognition: Introducing a unified framework to model uncertainty and information search. (*in preparation*).
- [4] Cavagnaro, D. R., Myung, J. I., Pitt, M. A., & Kujala, J. V. (2010). Adaptive design optimization: A mutual information-based approach to model discrimination in cognitive science. *Neural computation*, 22(4), 887-905.