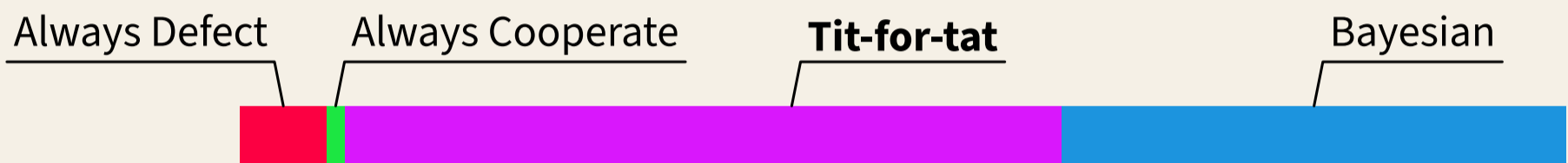
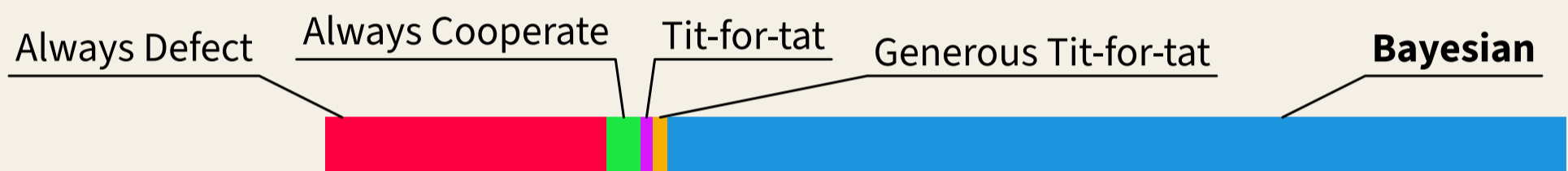


We can **infer** how **generous** someone is toward us. This may have **evolved** because we don't know **exactly** how much it **costs** them to **help** or **hinder** us. (i.e., there is **NOISE!**)

Evolutionary distribution **without noise:**



Evolutionary distribution **with noise:**



### The evolution of reciprocity based on welfare tradeoff ratios in games with asymmetric information

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#### Background

- People are motivated to help or harm another person to varying degrees
- **Welfare tradeoff ratios (WTRs)** quantify the direction and magnitude of such motivations
- **Definition of WTR:**  $u_A = w_A + \lambda_{AB} w_B$   
 $w_A / w_B$ : Person A/B's (objective) welfare  
 $\lambda_{AB}$ : A's **welfare tradeoff ratio** toward B  
 $u_A$ : A's utility, which she tries to maximize
- People can infer another person's **WTR** toward themselves; **How did such abilities evolve?**

#### Methods

- **Tournament** of different agents playing **repeated alternating games** with each other
- Each game is a **one-player binary** allocation decision; e.g., Option 1 gives \$2 to A and \$8 to B; Option 2 gives \$5 to A and \$2 to B
- The **payoffs change** from round to round
- B's decision tells A whether  $\lambda_{BA}$  is **above or below a threshold**, determined by the payoffs
- **Evolutionary simulation** based on the resulting **pairwise mean payoff matrix**
- Specifically, we look at the **long-run distribution** of a Moran process
- In Experiment 1, players perceive the payoffs **without noise**
- In Experiment 2, **noise** is added to the two players' perceptions **independently**, and neither players knows exactly what the other player sees

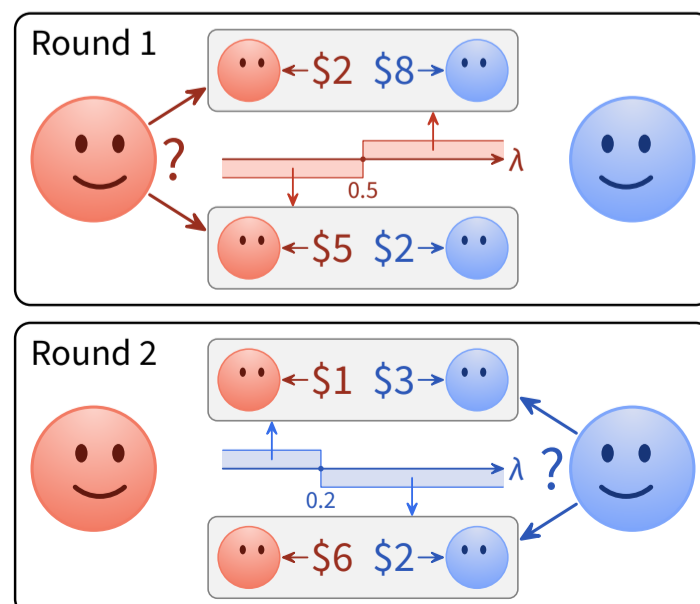
**Agents** (assuming A is self and B is opponent)

- **Always Defect/Cooperate:**  $\lambda_{AB}$  fixed to 0/1
- **Tit-for-tat** (a heuristic strategy): Assumes  $\lambda_{BA} = 0$  or 1 in each round; Starts with  $\lambda_{AB} := 1$ ; When B's decision in the last round distinguishes between  $\lambda_{BA} = 0$  and 1, sets  $\lambda_{AB} := \lambda_{BA}$
- **Generous Tit-for-tat:** Like Tit-for-tat, but unconditionally cooperates with some probability
- **Bayesian:** Does Bayesian inference on  $\lambda_{BA}$  with Hidden Markov Model; Sets  $\lambda_{AB}$  to the median of posterior of  $\lambda_{BA}$ , with slight bias toward 1

**Results** (see figure 🖱️)

- **Without noise in payoff perception:** Tit-for-tat performs well; Bayesian can do no better
- **With noise in payoff perception:** Tit-for-tat suffers from misperception and can't cooperate well with itself; Generous Tit-for-tat cooperates better with itself but is less resistant to invasion by Always Defect; Bayesian is robust and successful

#### Repeated one-player games with variable payoffs:



**When there is noise in payoff perception:**

