# Exercise 7.1: Expected Payoff

Setup:

- online learning, k = 2 actions
- probabilities algorithm selects each action in round *i* are:

$$\pi^i = (\pi^i_1, \pi^i_2) = (2/3, 1/3)$$

• payoffs of each action in round *i* are:

$$\mathbf{v}^i = (v_1^i, v_2^i) = (3, 9)$$

**Question:** What is the expected payoff of the algorithm in round *i*? (Answer on Canvas)

### Last Time:

- online learning (cont)
- warmup: geometric random variables
- follow the perturbed leader (analysis)

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- online learning (cont)
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## Today:

- multi-armed bandit learning
- reduction to online learning

## Per-stage Regret Review

**Recall:** the per-round regret of exponential weights alg is  $2h\sqrt{\ln k/n}$ 

- dependence on maximum value h is O(h)
- dependence on number of rounds *n* is  $O(\sqrt{1/n})$
- dependence on number of actions k is  $O(\sqrt{\log k})$

# Exercise 7.2: MAB-EW

Setup:

- payoffs in [0, h]
- apply multi-armed-bandit reduction to exponential weights alg
- recall Theorem:  $\mathbf{E}[MAB] \ge (1 2\epsilon) \operatorname{OPT} \frac{h k}{\epsilon^2} \ln k$
- optimally tune the learning rate  $\epsilon$  for n rounds

**Question:** Analyze the per-round regret, what is dependence on maximum payoff h? Number of rounds n? Number of actions k? (Answer on Canvas)