

# Entropy

Let  $P_r: [n] \rightarrow (0, 1)$  be a probability function.

Let  $p_i = P_r(i)$  for each  $i \in [n]$ .

Then  $H = \sum_{i=1}^n p_i \log \frac{1}{p_i}$  is the (Shannon) entropy of  $P_1, \dots, P_n$ .

H measures uncertainty about the outcome.

→ information learned when the outcome is revealed.

Maximum value of H is  $\log n$ , when  $p_1 = p_2 = \dots = p_n = \frac{1}{n}$ .

$$H = \sum_{i=1}^n \frac{1}{n} \cdot \log \frac{1}{1/n} = \sum_{i=1}^n \frac{1}{n} \log n = \log n.$$

H can be arbitrarily close to zero.

$$p_1 = 1 - \epsilon(n-1), \quad p_2 = p_3 = \dots = p_n = \epsilon.$$

$$H = \underbrace{(1 - \epsilon(n-1)) \cdot \log \frac{1}{1 - \epsilon(n-1)}}_{\substack{\rightarrow 1 \\ \downarrow \\ \rightarrow 0}} + \underbrace{(n-1)\epsilon \cdot \log \frac{1}{\epsilon}}_{\substack{\rightarrow \ominus \\ \rightarrow 0}}$$

$\rightarrow 0.$