

$X_1, \dots, X_n \sim \text{independent } B(p)$

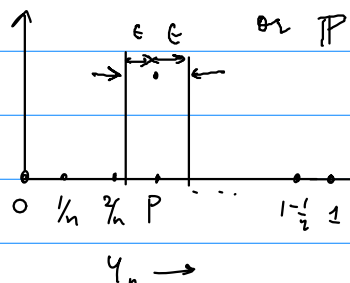
$$X_i \in \{0, 1\}$$

$$P(X_i = 1) = p$$

$$P(|Y_n - p| < \epsilon) \geq 1 - ?$$

$$\text{or } P(|Y_n - p| > \epsilon) \leq ?$$

$$Y_n = \frac{X_1 + \dots + X_n}{n}$$



$$E Y_n = E \left(\frac{X_1 + \dots + X_n}{n} \right)$$

$$= \frac{E X_1 + E X_2 + \dots + E X_n}{n} = \frac{np}{n} = p$$

$$E X_i = 0(1-p) + 1 \cdot p = p$$

$$E(Y_n - E Y_n)^2 = \text{var}(Y_n) = \text{var}\left(\frac{X_1 + \dots + X_n}{n}\right) = \frac{1}{n^2} \left[\text{var}(X_1 + \dots + X_n) \right]$$

$$= \frac{1}{n^2} \left[\text{var}(X_1) + \dots + \text{var}(X_n) \right]$$

$$\text{var}(X_i) = E[(X_i - p)^2] = (0-p)^2(1-p) + (1-p)^2 p$$

$$= \frac{np(1-p)}{n^2} = \frac{p(1-p)}{n}$$

Hoeffding's inequality (Concentration Inequalities)

$X_1, \dots, X_n \sim \text{independent random variables}$ $E X_i = \mu$
 $X_i \in [a, b]$ $Y_n = \frac{X_1 + \dots + X_n}{n}$

$$P(|Y_n - \mu| > \epsilon) \leq 2 \exp\left(-\frac{n \epsilon^2}{2(b-a)^2}\right)$$

$$P(|Y_n - \mu| < \epsilon) \geq 1 - 2 \exp\left(-\frac{n \epsilon^2}{2(b-a)^2}\right)$$

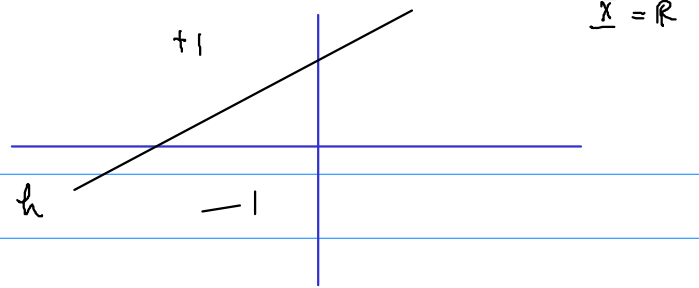
PAC Learning

PAC: Probably Approximately Correct

VC dimension

\mathcal{X} : instance space. \mathcal{Y} : label set $\{-1, 1\}$.

Hypothesis class = Set of "hypothesis".



$H = \{ \text{all linear classifiers} \}$.

Paul Erdos

