

Chebyshev's Inequality: Suppose we also know  $\text{Var}(X)$ .  
Theorem: If  $Y$  is a r.v. with  $E[Y] = \mu$ , then for any  $\alpha > 0$ ,  $P(|Y - \mu| \geq \alpha) \leq \text{Var}(Y) / \alpha^2$ .  
 Equivalently, if  $\sigma = \sqrt{\text{Var}(Y)}$ , then for any  $t > 0$ ,  $P(|Y - \mu| \geq t\sigma) \leq 1/t^2$ .

Proof: Let  $X = (Y - \mu)^2$ .  $X$  is nonnegative, so  

$$P(|Y - \mu| \geq \alpha) = P(X \geq \alpha^2) \leq \frac{E[X]}{\alpha^2}$$

$$= \frac{E[(Y - \mu)^2]}{\alpha^2} = \frac{\text{Var}(Y)}{\alpha^2}$$

Ex:  $Y$  is daily business cost,  $E[Y] = 1500$ , and  $\text{Var}(Y) = \sigma^2 = (200)^2$ .

$$P(Y \geq 2500) = P(Y - 1500 \geq 1000) = P(Y - \mu \geq 1000) \leq P(|Y - \mu| \geq 1000) \leq \frac{200^2}{1000^2} = \left(\frac{1}{5}\right)^2 = \frac{1}{25}$$

Cantelli's Inequality (one-sided Chebyshev)  
 If  $\alpha > 0$ ,  $P(Y - \mu \geq \alpha) \leq \frac{\text{Var}(Y)}{\text{Var}(Y) + \alpha^2}$ .

Ex:  $\frac{1}{26} = P(Y \geq 2500) = P(Y - 1500 \geq 1000) \leq \frac{200^2}{200^2 + 1000^2}$

$$= \frac{1}{1 + 5^2} = \frac{1}{26}$$

## Chernoff Bound

Theorem: Suppose  $X \sim \text{Bin}(n, p)$  and  $\mu = E[X]$ .

For any  $0 < \delta < 1$ ,

$$P(X \geq (1+\delta)\mu) \leq e^{-\frac{1}{3}\delta^2\mu} \text{ and}$$

$$P(X \leq (1-\delta)\mu) \leq e^{-\frac{1}{2}\delta^2\mu}$$

## Law of large numbers

Consider i.i.d. r.v.'s  $X_1, X_2, \dots$ , where

$$E[X_i] = \mu < \infty \text{ and } \text{Var}(X_i) = \sigma^2 < \infty.$$

Define sample mean  $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$ .

$$E[\bar{X}_n] = E\left[\frac{1}{n} \sum_{i=1}^n X_i\right] = \frac{1}{n} \sum_{i=1}^n E[X_i] = \frac{1}{n} \cdot n \cdot \mu = \mu.$$

$$\text{Var}(\bar{X}_n) = \text{Var}\left(\frac{1}{n} \sum_{i=1}^n X_i\right) = \frac{1}{n^2} \sum_{i=1}^n \text{Var}(X_i) = \frac{1}{n^2} \cdot n \sigma^2 = \frac{\sigma^2}{n}$$

As  $n$  increase,  $\bar{X}_n$  is more likely to be ~~ed~~ close to  $\mu$ .

Theorem (Weak Law of Large Numbers): For any  $\epsilon > 0$ ,  
as  $n \rightarrow \infty$ ,  $P(|\bar{X}_n - \mu| > \epsilon) \rightarrow 0$ .

Proof: By Chebyshev's Inequality,

$$P(|\bar{X}_n - \mu| > \epsilon) \leq \frac{\sigma^2}{n\epsilon^2} \rightarrow 0 \text{ as } n \rightarrow \infty.$$

## Probabilistic Algorithms

Background:

Given an  $n$ -digit number  $x$ , can it be factored into its prime factors efficiently?

In time some small polynomial in  $n$ . E.g.,  $O(n^2)$

Big open problem today.

Simpler question: determine if  $x$  is prime or composite?

Solovay & Strassen 1977: A probabilistic polynomial time algorithm to determine if  $x$  is prime or composite

## Quicksort (Hoare 1959)

To sort  $a_1, a_2, \dots, a_n$ : If  $n > 1$

1. Choose  $p \in \{1, 2, \dots, n\}$  uniformly and randomly.  
 $p \sim \text{Unif}(1, n)$ .

2. Let  $L = \{a_i \mid a_i < a_p\}$   
 $E = \{a_i \mid a_i = a_p\}$   
 $G = \{a_i \mid a_i > a_p\}$ .

3. Recursively sort and output  $L$

Output  $E$

Recursively sort and output  $G$ .

The expected time used by quicksort is  $O(n \log n)$ .