

# **SYSTEMS 302**

## **LECTURE 2**

- **DISCRETE RANDOM VARIABLES**
  - **Bernoulli Random Variables**
  - **Binomial Random Variables**
- **SUMS OF RANDOM VARIABLES**
  - **Means of Sums**
  - **Variance of Sums**
- **For next time:**
  - **Devore, Sections 5.1,5.2,5.5**

## **ASSERTION:**

**“The probability of heads  
on any toss is  $\frac{1}{2}$ .”**

## **QUESTION:**

**What does it mean?**

# LAPLACE HYPOTHESIS

If there is no reason to believe that one outcome is more likely than the other, then they must be treated as **equally likely** outcomes.

$$p_1 = p_2$$

$$p_1 + p_2 = 1$$

$$\Rightarrow 2p_1 = 1$$

$$\Rightarrow p_1 = \frac{1}{2} = p_2$$

# VON MISES HYPOTHESIS

There exists a **population** (universe) of ‘potential states of the world’, each including a ‘Head’ or ‘Tail’ outcome. The realized coin flip is treated as a **sample** from this population.

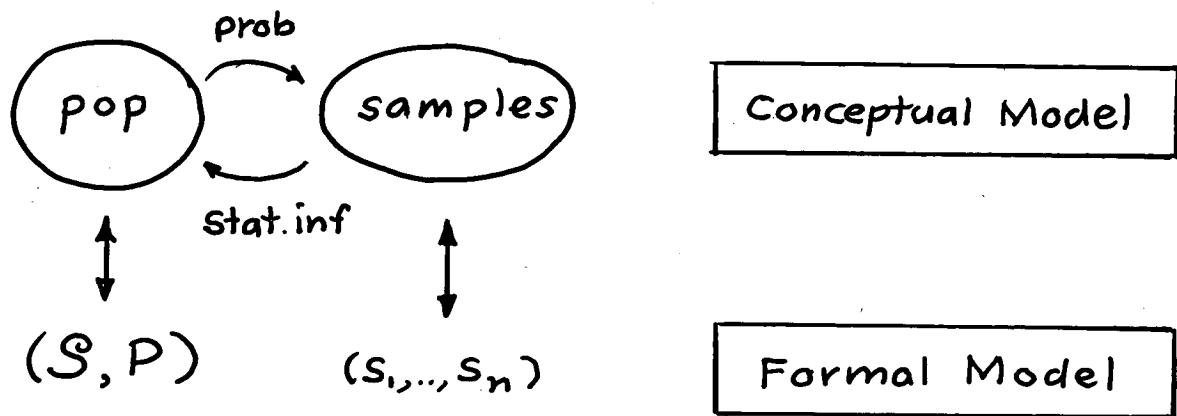
# • FORMAL STATISTICAL POPULATIONS

$s$  = possible sample (or 'state of the world')

$S$  = sample space (set of all possible 'states')

$P(s)$  = probability of sample  $s$

$(S, P)$  = statistical population (or 'probability space')



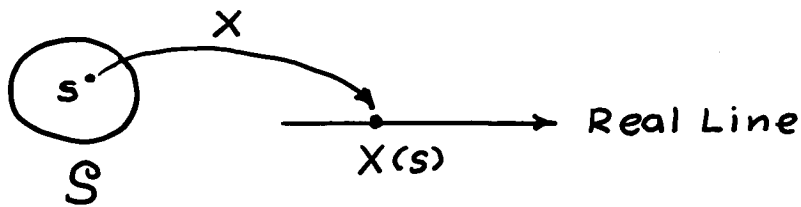
## • EXAMPLE :

$S$  = sys302 class

$s$  = individual student

$P(s)$  = probability that student  $s$  is sampled

# • DISCRETE RANDOM VARIABLES



$X =$  random variable on  $(S, P)$

$=$  numerical property of samples,  $s$ .

## • Example:

$X(s) =$  age (weight, etc.) of student,  $s$ .

$x =$  realized value of  $X$  (data!)

$\mathcal{X} = \{x_1, x_2, \dots\}$

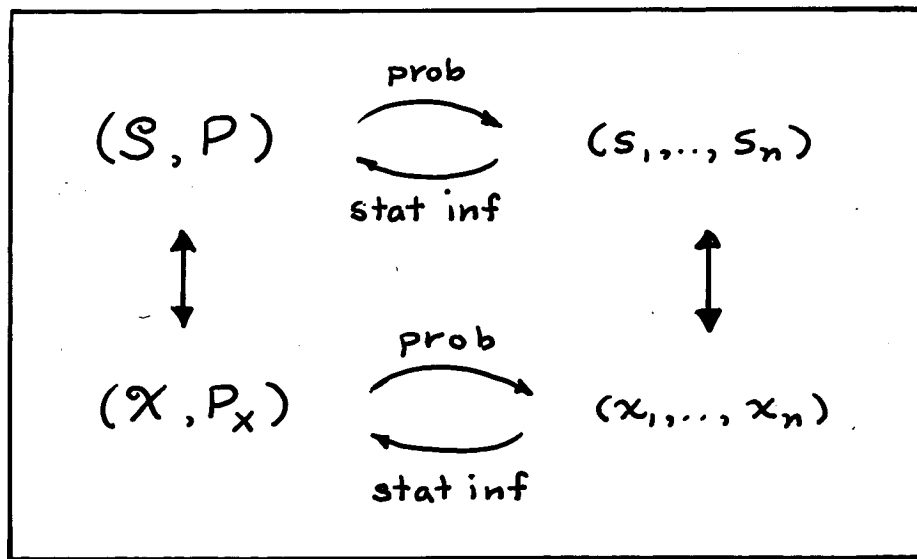
$=$  set of possible  $X$ -realizations

$P_X(x) = P(X=x) = P\{s \in S : X(s) = x\}$

$=$  probability that value  $x$  occurs

$P_X =$  probability distribution on  $\mathcal{X}$

## • THE X-POPULATION MODEL



## • EXPECTED VALUES

$$P_x = p(x) = P_X(x) \quad (\text{equivalent notation})$$

$$\mu = E(X) = \sum_x x p(x)$$

= mean of  $X$  (first moment of  $X$ )

$$\sigma^2 = E[(X-\mu)^2] = \sum_x (x-\mu)^2 p(x)$$

= variance of  $X$  (second moment of  $X$ )

➔ X-population properties

# • BERNOULLI RANDOM VARIABLES

- BERNOULLI EVENTS: Voting: 1 = 'Obama Wins'  
Testing: 1 = 'Pass'

$$X(s) = \begin{cases} 1 & , \text{ event occurs in state } s \\ 0 & , \text{ event does not occur in state } s \end{cases}$$

$$p = P(X=1) = P\{s \in \mathcal{S} : X(s) = 1\}$$

$$1-p = P(X=0) = P\{s \in \mathcal{S} : X(s) = 0\}$$

$$(X, P_X) \rightarrow \begin{array}{|c|c|} \hline 1 & p \\ \hline 0 & 1-p \\ \hline \end{array}$$

$$X \sim \text{Bernoulli}(p)$$

## • MOMENTS

$$\mu = 1 \cdot p + 0 \cdot (1-p) = \boxed{p}$$

$$\sigma^2 = (1-p)^2 p + (0-p)^2 (1-p) = \boxed{p(1-p)}$$

→ Population Values

## **BATCH SAMPLING PROBLEM:**

**Given a shipment of 1000 light bulbs with 50 defective bulbs, take an **inspection sample** of 20 bulbs.**

**Q1.** What is the chance of getting **at least one** defective bulb in the sample?

**Q2.** What is the **expected number** of defective bulbs in the sample?

**Q3.** What is the **variance** of this number?

## • BINOMIAL RANDOM VARIABLES

→ Given  $n$  independent Bernoulli random variables

$$X_i \sim \text{Bernoulli}(p), \quad i=1, \dots, n$$

the associated Binomial random variable is:

$$X = \sum_{i=1}^n X_i = \text{number of event occurrences}$$

$$P(X_1=1, \dots, X_k=1, X_{k+1}=0, \dots, X_n=0) = p^k (1-p)^{n-k}$$

$$P(X=k | n, p) = \binom{n}{k} p^k (1-p)^{n-k}, \quad k=0, 1, \dots, n$$

→  $X \sim \text{Bin}(n, p)$

### • MOMENTS

$$\mu = \sum_{k=0}^n k \binom{n}{k} p^k (1-p)^{n-k} = np$$

$$\sigma^2 = \sum_{k=0}^n (k-np)^2 \binom{n}{k} p^k (1-p)^{n-k} = np(1-p)$$