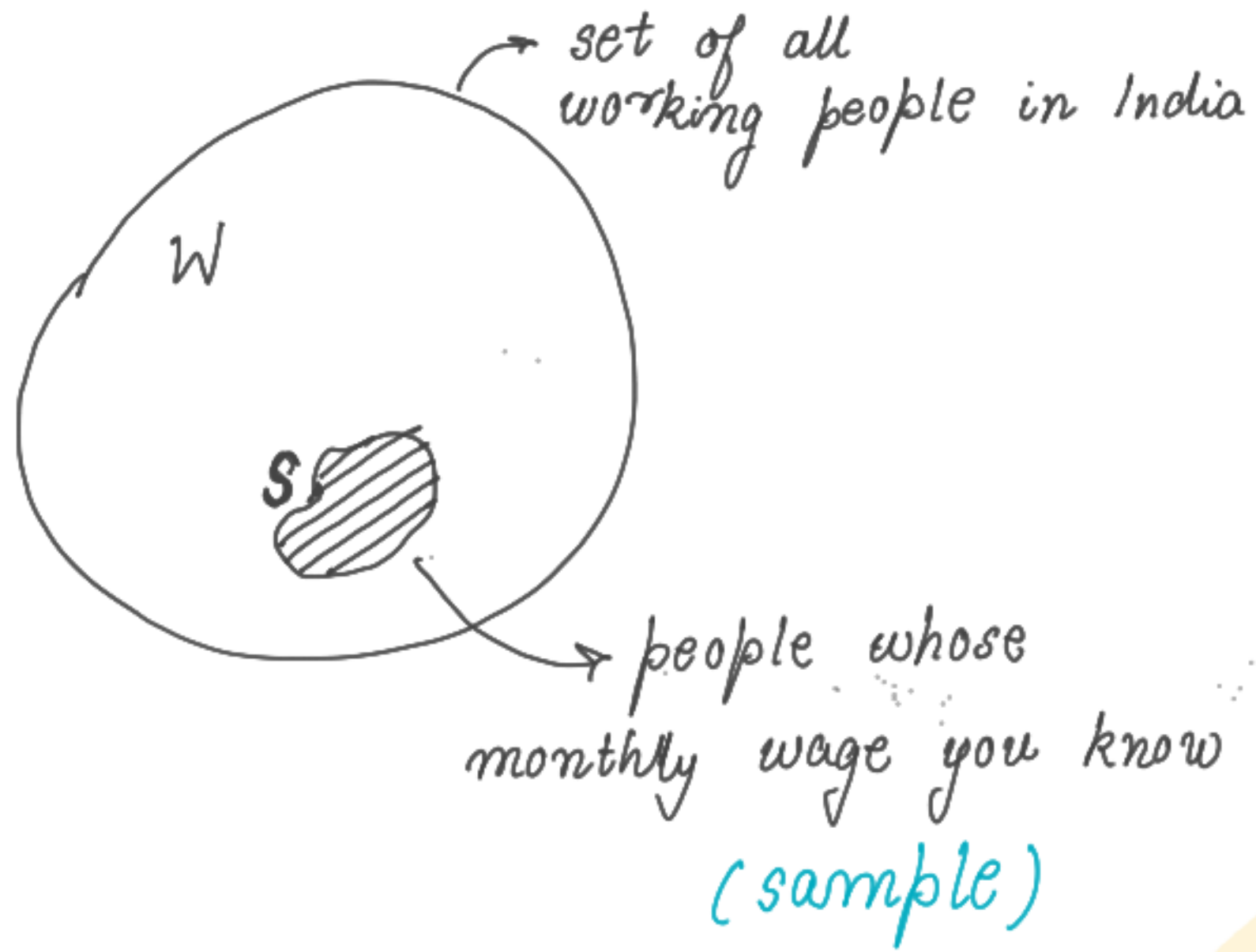


Topics to be covered:

- inferential statistics - why
- probability
- probability distributions
  - discrete
  - continuous
- statistical inference of
  - point estimate
  - interval estimate

# Why inferential statistics?

GOI wants to find <sup>estimate</sup> avg monthly wage <sup>property</sup> for the working population (population)



We want to estimate the avg. monthly wage for  $W$  by using information about avg. monthly wage in  $S$

By intelligently choosing  $S$ ; we can estimate properties of  $W$  very accurately.



A

high wage  
earners



B

low wage  
earners



Intelligently choosing 5  
would mean doing

Stratified Random Sampling

instead of Simple Random  
Sampling

# Probability

associated with a random event/experiment

all outcomes of random experiment are called sample space

an event is a subset of the sample space

RE: A batsman facing a bowler

Sample Space : { 'out', 'dot', 'one', 'two', ..., 'seven', 'extras' }

RV: # of runs scored

{ 0, 0, 1, 2, ..., 7, 1 }

RV: if the batsman got out

{ 1, 0, 0, 0, ..., 0 }

## Probability (contd...)

Random variable  
mapping from the sample space to the set of real numbers

Probability  
mapping from the set of values of random variables  
to the set  $[0, 1]$

$P(X = x)$  = probability of the event that  
the random variable  $X$  takes  
the value  $x$ .

$$1 \geq P(X = x) \geq 0$$

Axioms

of  
probability

$$\sum_x P(X = x) = 1$$



Random  
Experiment



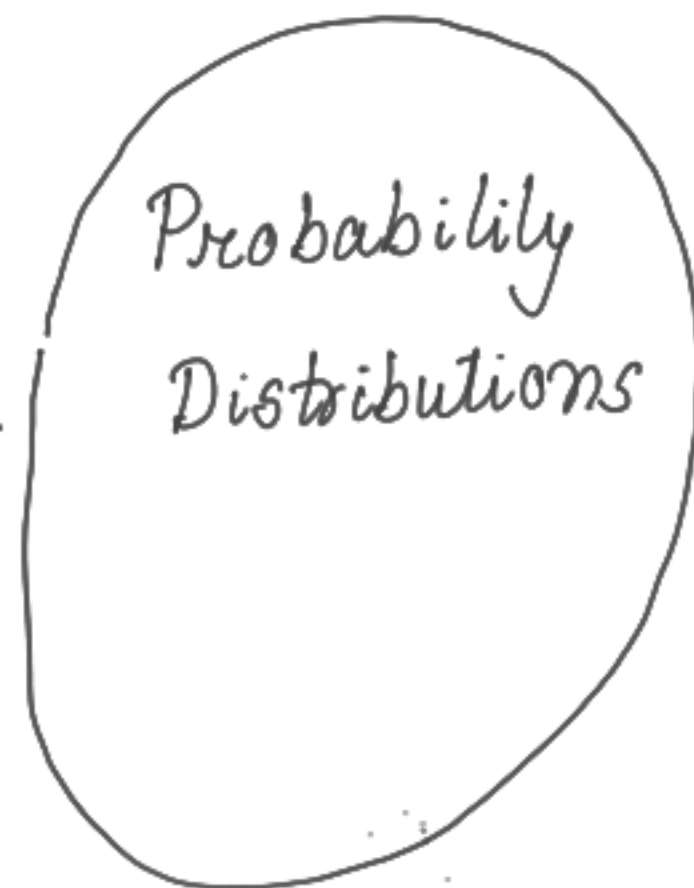
Outcomes

RV



$\mathbb{R}$

[ discrete  
continuous



$[0, 1]$

# Probability Distributions

RE:

SS:

RV:  $\{x_1, x_2 \dots x_k\} \rightarrow$  discrete

$[x_1, x_3] \rightarrow$  continuous  $\rightarrow P(X=x) = 0$

probability at a point is 0  
by def.<sup>n</sup> for cont. RV

PDF:  $P(X=x_i) = ? \rightarrow D \rightarrow \text{dist}^n \rightarrow \text{discrete}$

$P(x_i \leq X \leq x_j) = ? \rightarrow D \rightarrow \text{density} \rightarrow \text{continuous}$

$P(x-h \leq X \leq x+h) = f(x)$

$$\int_{-\infty}^{\infty} f(x) dx = 1$$

Given the PDF of any probability dist<sup>n</sup>; we want to calculate some characteristics:

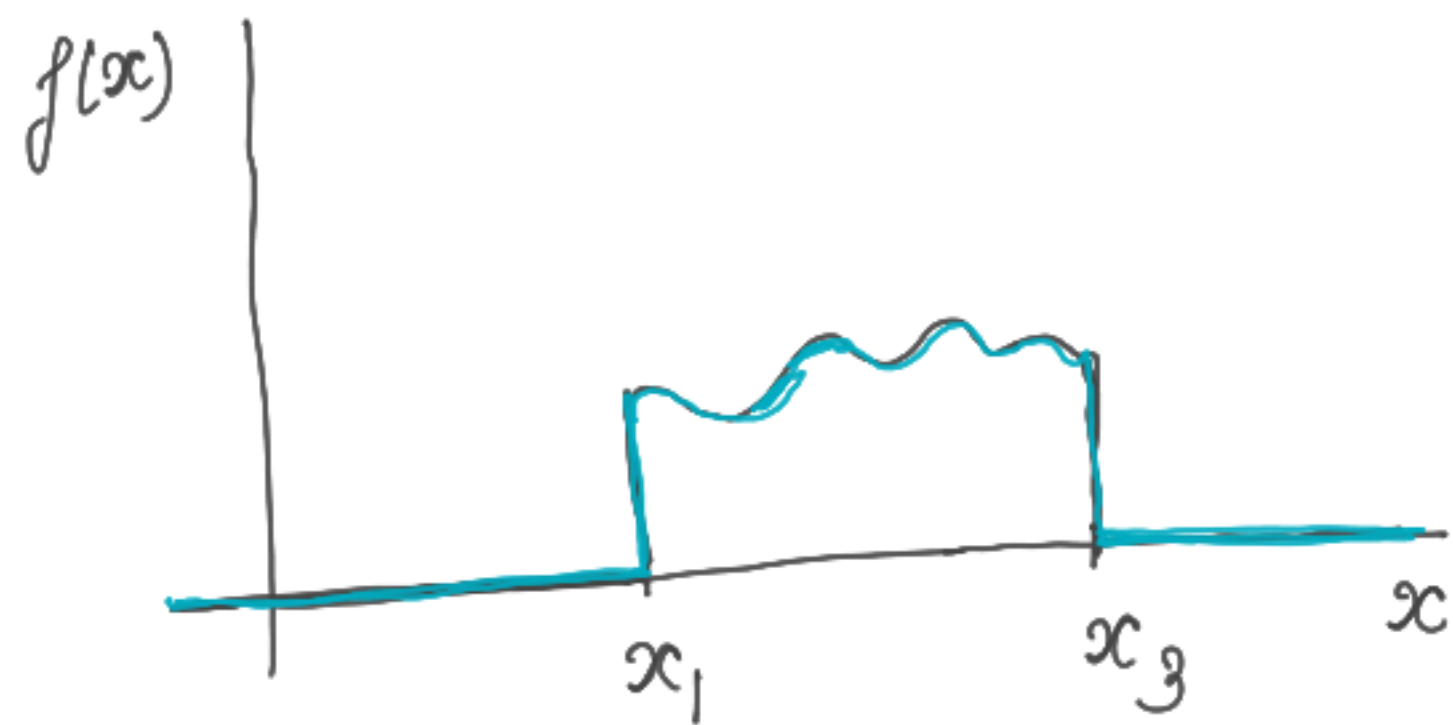
- central value of the RV  $\rightarrow$  Mean  $\rightarrow E(X)$
- extent of spread of the RV  $\rightarrow$  Variance  $\rightarrow \text{var}(X)$

$$E(X) = \sum_x x P(X=x) dx = \int_{-\infty}^{\infty} x f(x) dx$$

$$\text{var}(X) = E(X^2) - [E(X)]^2$$

$$E(X^2) = \sum_x x^2 P(X=x) dx = \int_{-\infty}^{\infty} x^2 f(x) dx.$$





$$\begin{aligned} \int_{-\infty}^{\infty} f(x) dx &= \int_{x_1}^{x_3} f(x) dx \\ \Rightarrow \int_{-\infty}^{x_1} f(x) dx + \int_{x_1}^{x_3} f(x) dx &+ \int_{x_3}^{\infty} f(x) dx \end{aligned}$$

Arrows indicate the mapping of the intervals: from  $-\infty$  to  $x_1$ , from  $x_1$  to  $x_3$ , and from  $x_3$  to  $\infty$ .

$\int_{-\infty}^{\infty}$  is just a lazy way to  
write  $\int$  over all values of  $x$ .

Random Exp : Tossing an unfair dice

SS :  $\{ \cdot, \cdot, \cdot, \cdot, \cdot, \cdot \}$

RV :  $\{ 1, 2, 3, 4, 5, 6 \}$  |  
                    unfair                      fair

X	$P(X=x)$	$P(X=x)$
1	1/21	1/6
2	2/21	1/6
3	3/21	1/6
4	4/21	1/6
5	5/21	1/6
6	6/21	1/6

$$P(X=x) \propto x \quad \sum_x P(X=x) = 1$$

$$P(X=x) = kx \Rightarrow k(1+2+3+4+5+6) = 1$$
$$\Rightarrow k = 1/21$$

$$E(X) = \sum_x x P(X=x) = \frac{1^2}{21} + \frac{2^2}{21} + \dots + \frac{6^2}{21}$$
$$= \frac{91}{21} = 4.\bar{33} \rightarrow \text{unfair.}$$

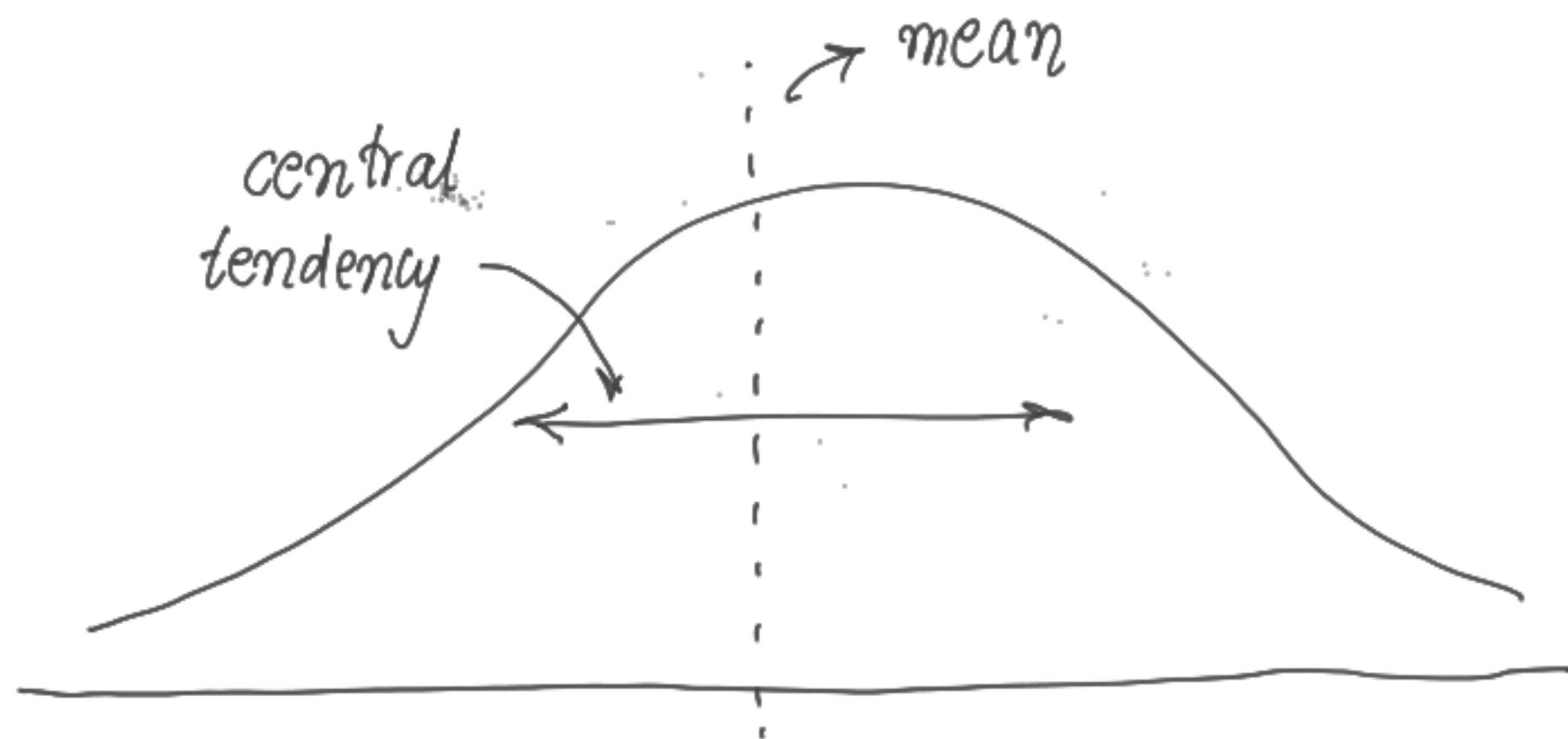
$$E(X) = \frac{21}{6} = 3.5 \rightarrow \text{fair.}$$
$$= \sum_{x=1}^6 x \cdot \left(\frac{1}{6}\right) = \frac{1}{6} \sum_{x=1}^6 x$$

$$\begin{aligned}
 E(X^2) &= \sum_x x^2 P(X=x) \\
 \text{unfair} &= \sum_{x=1}^6 \frac{x^3}{21} = \frac{1}{21} \left( \frac{6 \times 7}{2} \right)^2 = 21
 \end{aligned}$$

$$\begin{aligned}
 \text{var}(X) &= E(X^2) - E(X)^2 \\
 \text{unfair} &= 21 - 4.\bar{3}\bar{3}^2 \\
 &= 21 - 18.8 = 2.2
 \end{aligned}$$

$$\begin{aligned}
 E(X^2) &= \sum_{x=1}^6 x^2 \cdot \frac{1}{6} = 15.16 \\
 \text{fair}
 \end{aligned}$$

$$\begin{aligned}
 \text{var}(X) &= 15.16 - 12.25 \\
 \text{fair} &= 2.91
 \end{aligned}$$



Uniform Discrete

$$X: \{1, 2, \dots, k\} \rightarrow \text{RV}$$

$$\text{PDF: } P(X=x) = \frac{1}{k}$$

→ fair dice

$$E(X) = \frac{(k+1)}{2}$$

$$\text{var}(X) = \frac{k^2-1}{12}$$

Bernoulli

$$X: \{0, 1\}$$

→ binary outcome

$$\text{PDF: } P(X=x) = p^x (1-p)^{1-x}$$

$$p = P(X=1)$$

$$E(X) = p$$

$$\text{var}(X) = p(1-p)$$

Binomial

$$X: \{0, 1, \dots, n\}$$

→  $x$  success in  
 $n$  binary outcome exp.

$$p = P(X=1)$$

$$E(X) = np$$

$$\text{PDF: } P(X=x) = {}^n C_x \underline{p}^x (1-p)^{n-x}$$

$$\text{var}(X) = np(1-p)$$

Geometric

$$X: \{0, 1, \dots\}$$

discrete, countable  
but infinite

→ # of times required  
to achieve success once  
in a binary outcome setting

$$E(X) =$$

$$\text{var}(X) =$$

$$\text{PDF: } P(X=x) = (1-p)^x p$$

Poisson

$$X: \{0, 1, \dots\}$$

→ # of times an  
event occurs per unit  
time.

$$E(X) = \lambda$$

$$\text{var}(X) = \lambda$$

$$\text{PDF: } P(X=x) = e^{-\lambda} \frac{\lambda^x}{x!}$$



## Normal Dist

$$X : (-\infty, \infty)$$

$$\text{PDF: } f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

→ natural processes

$$E(X) = \mu \quad \text{var}(X) = \sigma^2$$

↳ Standard Normal Dist

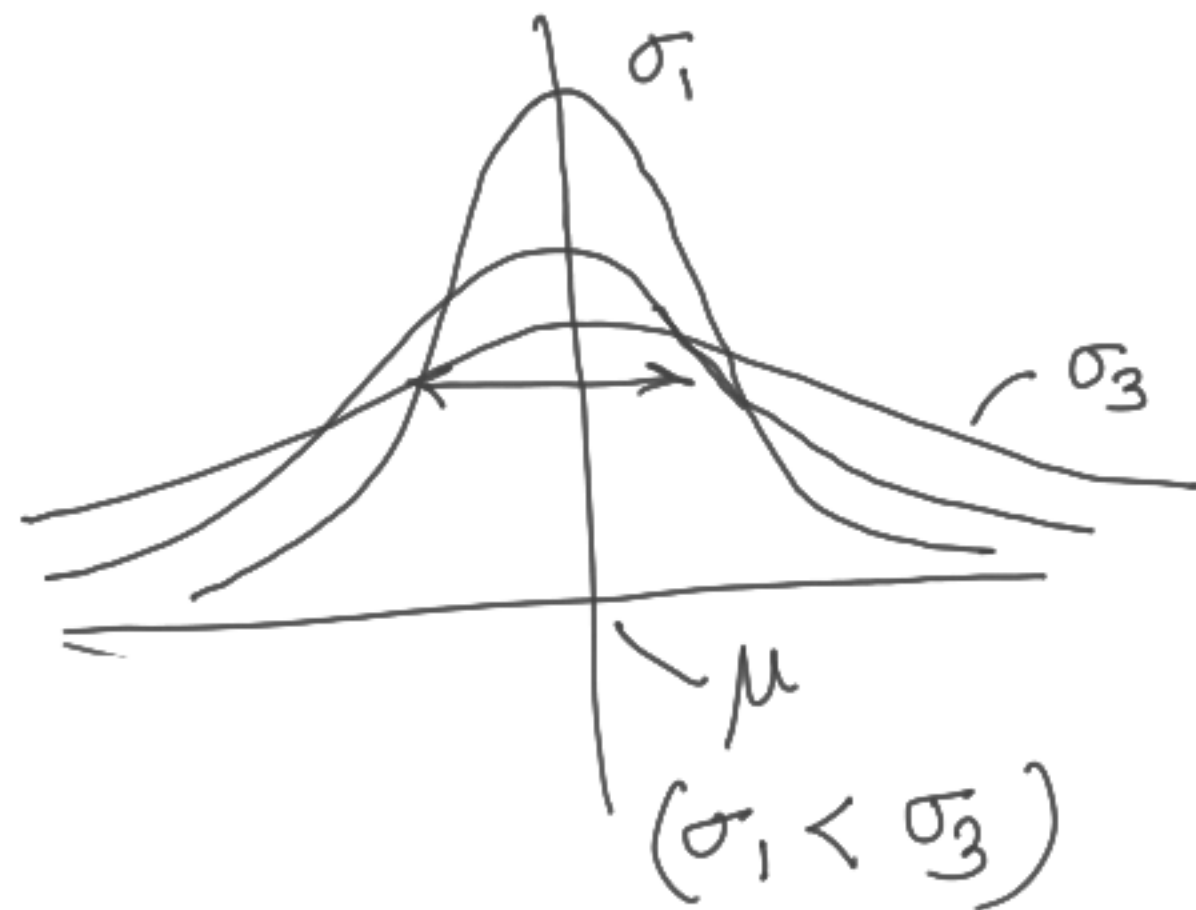
$$Z : (-\infty, \infty)$$

$$\text{PDF: } f(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$$

$$Z = \frac{x - \mu}{\sigma}$$

$$E(Z) = 0$$

$$\text{var}(Z) = 1$$



$$P(\underbrace{X < \mu + k}) = \int_{-\infty}^{\mu+k} f(x) dx$$

difficult ; and calc. redone if  $\mu, \sigma, k$  change

SND is a scale invariant sol<sup>n</sup>.

$$X < \mu + k \Rightarrow Z < k\sigma$$

$$\int_{-\infty}^{k\sigma} f(z) dz \rightarrow \text{looked from table}$$

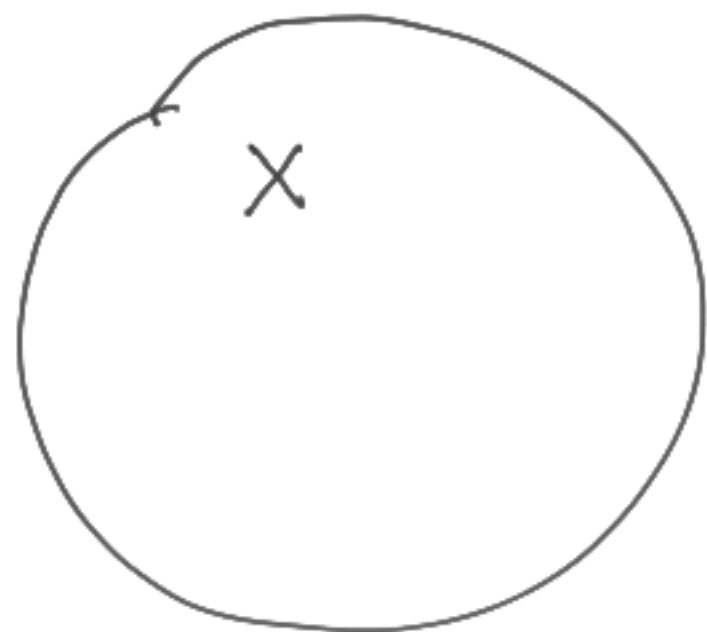
→ no need to calculate integrals.

Central Limit Theorem.

$$X \sim D$$

$$E(X) = \mu$$

$$\text{var}(X) = \sigma^2$$



interval  
estimate

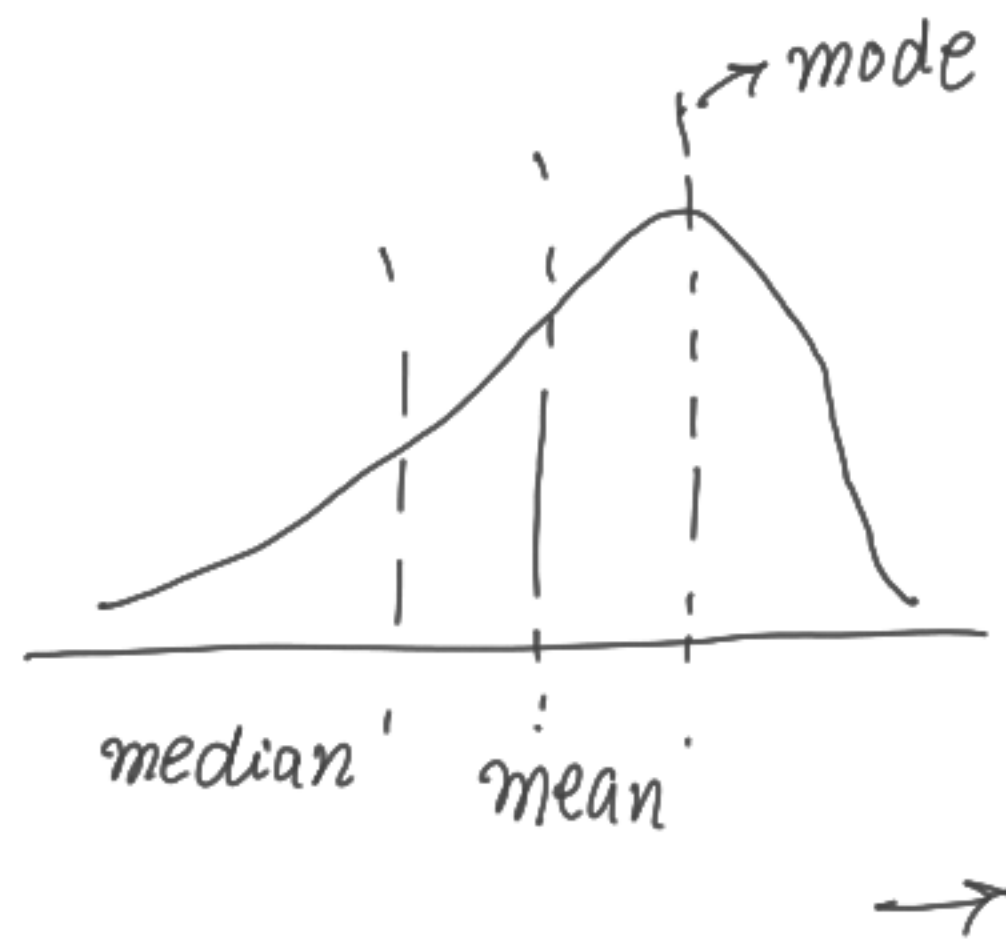
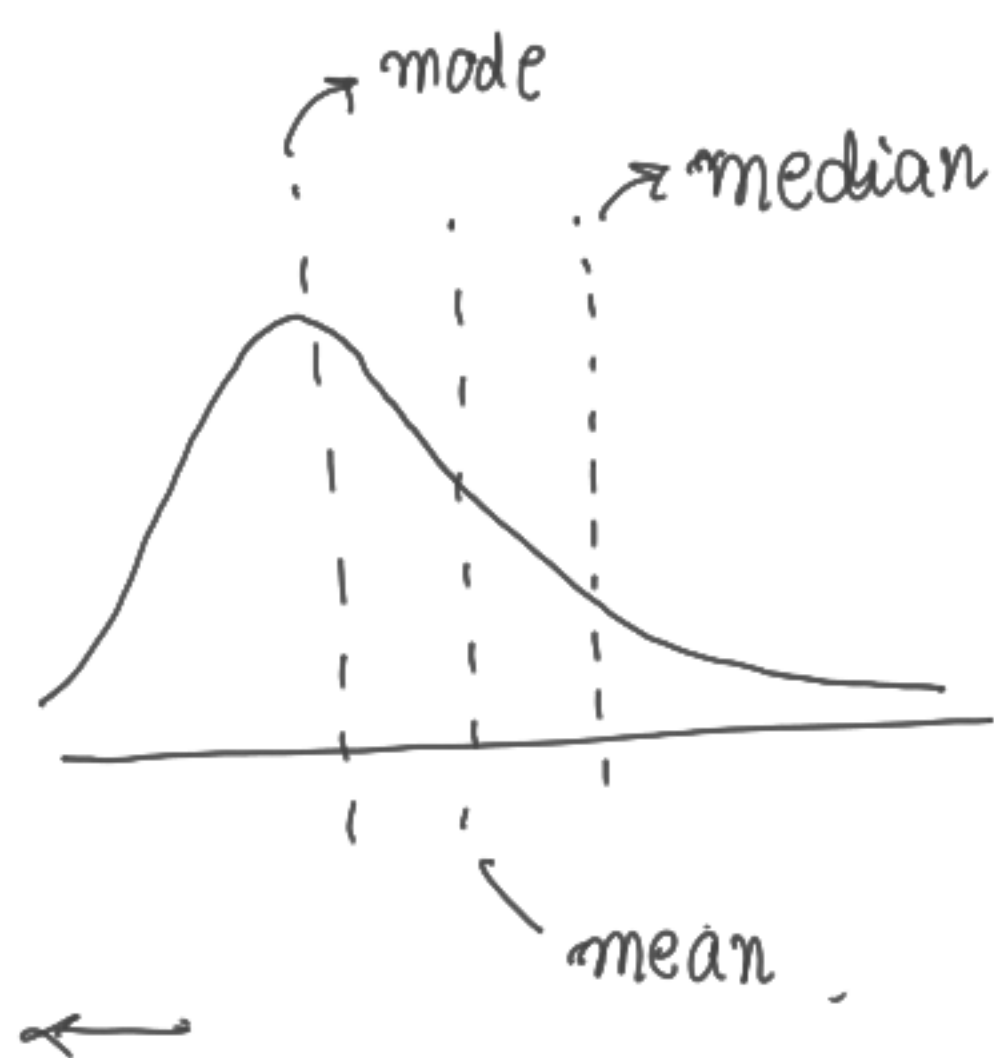
$$\sigma^2 = \frac{\sum \sigma_i^2}{k} \rightarrow t_{k-1}$$

$$\left. \begin{array}{l} \sigma_{S_1} \leftarrow S_1 \rightarrow \bar{X}_1 \\ \sigma_{S_2} \leftarrow S_2 \rightarrow \bar{X}_2 \\ \sigma_{S_3} \leftarrow S_3 \rightarrow \bar{X}_3 \\ \vdots \\ \bar{X}_k \end{array} \right\} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$

$$\mu \rightarrow \frac{\sum_{i=1}^k \bar{X}_i}{k}$$

point estimate

$$\rightarrow \frac{\sum X_i}{k} - \frac{\sum \sigma_i^2}{k} < \hat{\mu} < \frac{\sum X_i}{k} + \frac{\sum \sigma_i^2}{k}$$



skewness  $\neq 0$

$\Rightarrow$  non-symmetric PDF