

# Probability Distribution

Success / Failure  
0 / 1

Discrete

- \* - Bernoulli
- \* - Binomial
- Poisson

(Any value is valid)

Continuous

- \* - Normal / Gaussian
- Uniform
- Exponential
- logarithmic
- \* - Pareto (80:20)



0/1  
↑

① Bernoulli

+ Binary output

+ Single trial

$$P(H) = 0.6 \rightarrow P(x=0)$$

$$P(\text{Not } H) = 1 - 0.6 \rightarrow P(x=1) \\ = 0.4$$

② Binomial (multiple bernoulli)

+ Binary output

+ Multiple trial

$n$  = no. of trial

$P$  = Given probability

$X$  = ?

$\Rightarrow P(H)$  in 10 trials,  $P(H) = 0.5$



③ Normal/Gaussian : Mean  $\approx$  Median  $\approx$  Mode . (Bell Curve)  
(Symmetric Curve :  $\cap$ )

Continuous data can be understood : Mean ( $\mu$ ), Std ( $\sigma$ )

68% :  $\mu \pm 1\sigma$  : Boundary (LB, UB)

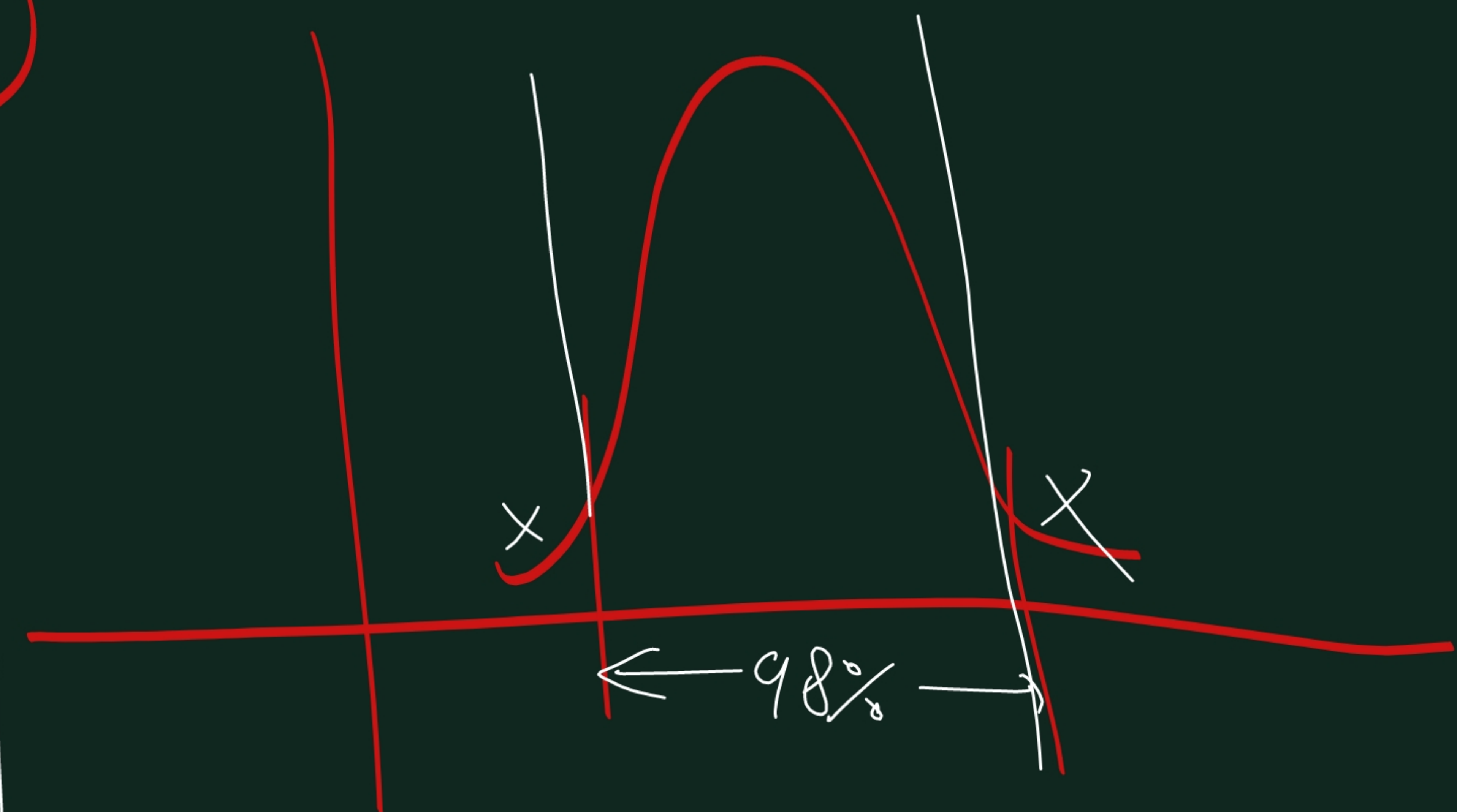
95% :  $\mu \pm 2\sigma$

98% :  $\mu \pm 3\sigma$

$$\mu = 23, \sigma = 4.2$$

$$LB = 23 - 2 * 4.2$$

$$UB = 23 + 2 * 4.2$$





④ Pareto (80 : 20)

↳ 20% data can explain 80% of distribution.

ex: if you 10 members in a family

then only 2 members are decision maker.

(Probability less  
of setting 3 Heads)

## Probability Distribution Function (scipy.stats)

① pmf() → Probability Mass Function → Discrete (=)

② pdf() → " Density " → Continuous (=)

③ cdf() → Cumulative Density Function → Discrete/Continuous

MIN MAX



# from scipy.stats import Bernoulli, Binom, Norm

Bernoulli.pmf ( )

Binom.pmf ( )

Norm.pdf ( )

## # Inferential Stats

↓  
Concluding  
Predicting  
Inferring

Population ( 100 % data )

Sample ( 30 % of data )

→ Help to conclude/infer about entire population/Data based on a sample/data.



