

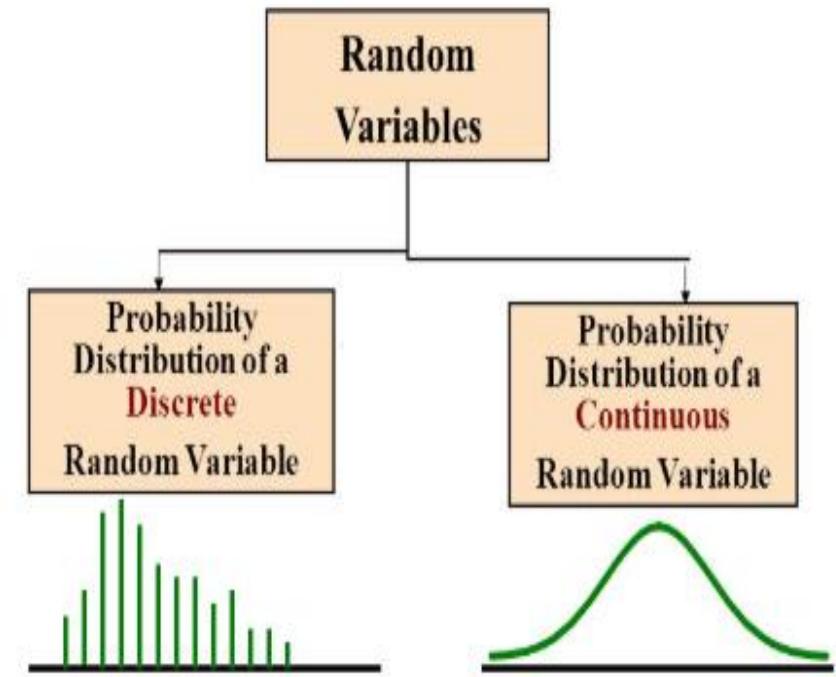
The background features abstract, overlapping green geometric shapes, primarily triangles and polygons, in various shades of green, creating a modern and dynamic visual effect. The shapes are layered, with some appearing more prominent than others, and they extend from the edges of the frame towards the center.

Data Science

Distribution

► Definition of Distribution:

- In statistics, a distribution is the way data points are spread or arranged over a range of values.
- It helps in understanding the underlying patterns and trends in the dataset.
- Discrete data: age = [22,64,12,89,56]
- Continuous data: weight = [26.4,38.2,11.9,76.3]



Uniform Distribution

Example: Rolling a die.

Here probability for each digit occurrence is $1/6$ (uniform or same)

$$\text{Mean} = \frac{(a + b)}{2}$$

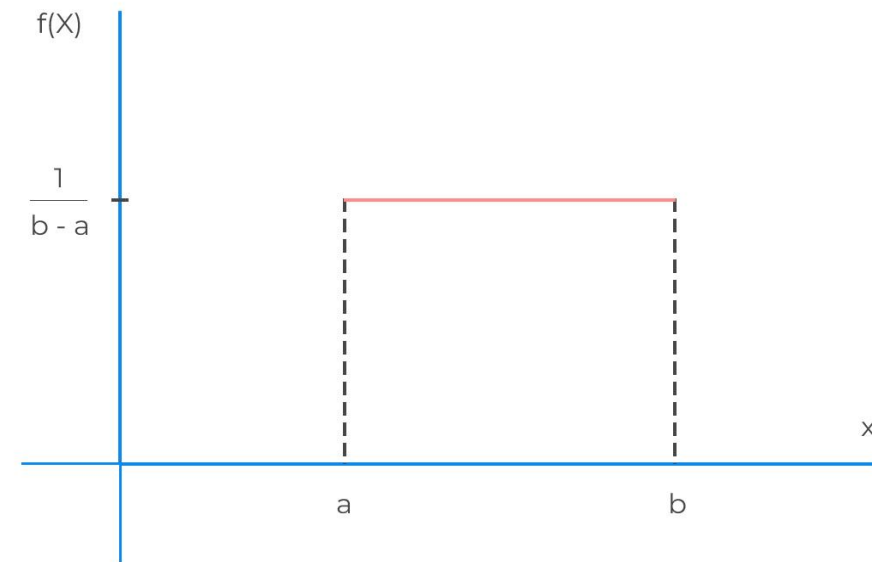
$$\text{Variance} = \frac{(b-a)^2}{12}$$

P. D. F. of Uniform Probability Distribution

$$f(x) = \begin{cases} f(x) = \frac{1}{b-a}, & \text{if } a < x < b. \\ 0, & \text{Otherwise.} \end{cases}$$



PDF of a Uniform Distribution



Area = 1

Numpy: `np.rand()`

Bernoulli Distribution

$$\text{PMF} = f(x, p) = \begin{cases} p & \text{if } x = 1 \\ q = 1 - p & \text{if } x = 0 \end{cases}$$

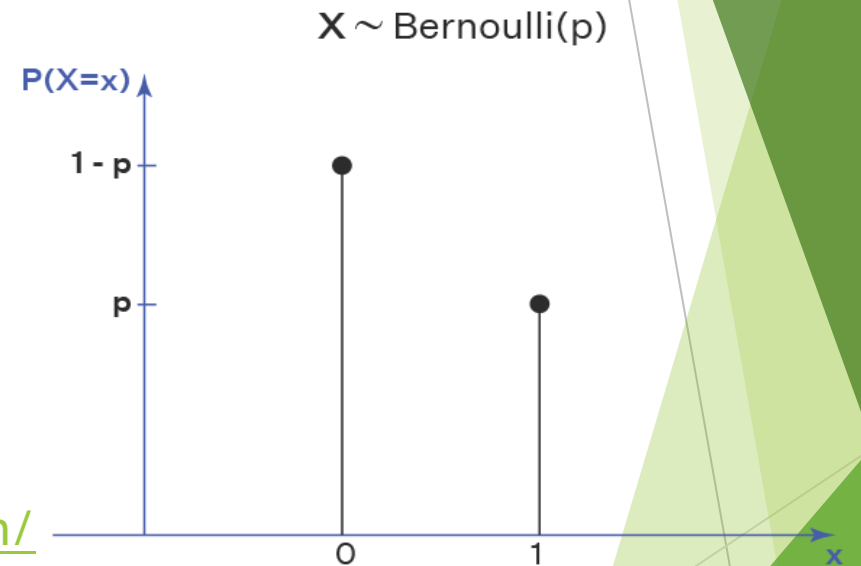
Thus, the **mean** or expected value of a Bernoulli distribution is given by

$$E[X] = p.$$

Hence, the variance of a Bernoulli distribution is $\text{Var}[X] = p(1 - p) = p \cdot q$

<https://www.cuemath.com/data/bernoulli-distribution/>

Bernoulli Distribution Graph

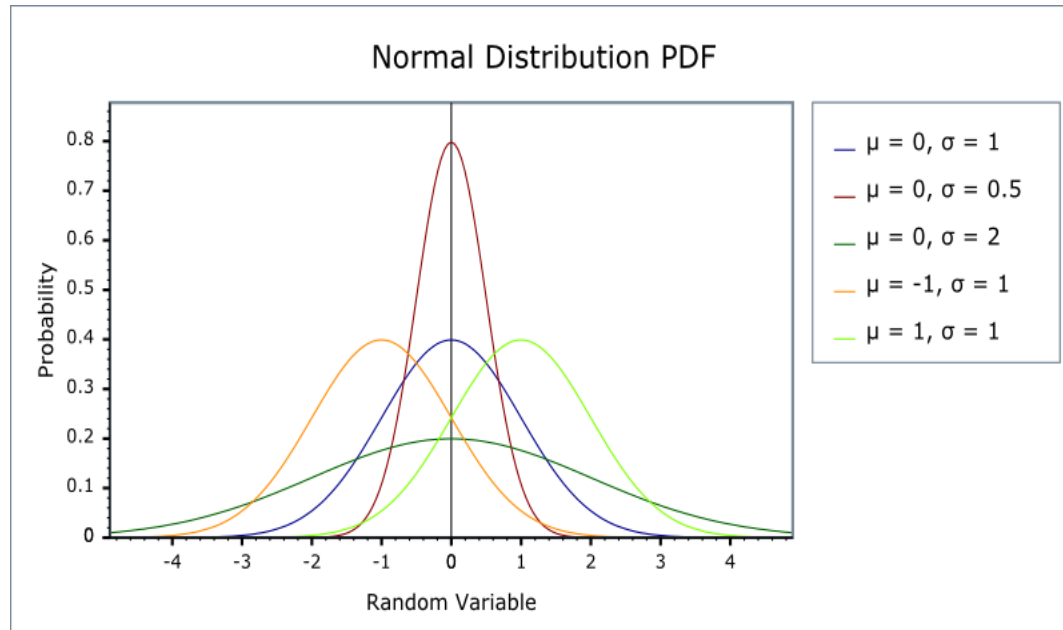
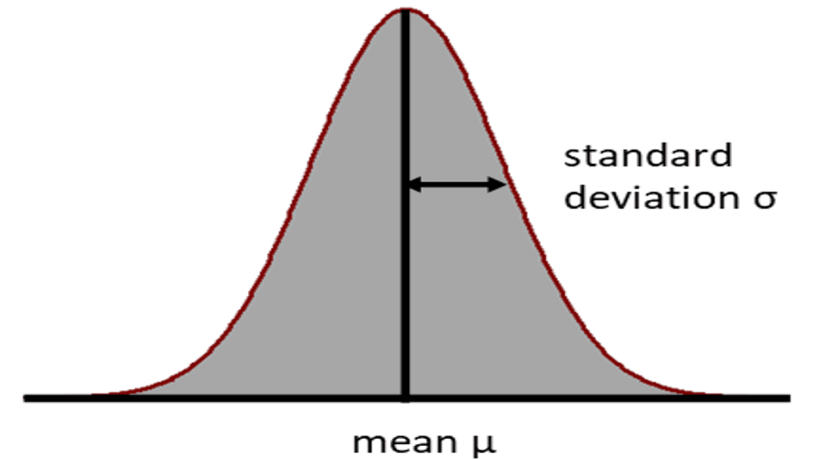


Example = Tossing a coin (heads or tails)
If $p(X=\text{head}) = 0.5$
Then $p(X=\text{tail}) = 1 - p(X=\text{head}) = 1 - 0.5 = 0.5$

Normal (Gaussian) Distribution

The normal distribution is an arrangement of data points in which most values form a cluster in the middle of the range and the rest taper off symmetrically toward either extreme ends

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$



Example: Height or weight of people in country

Normal (Gaussian) Outlier detection

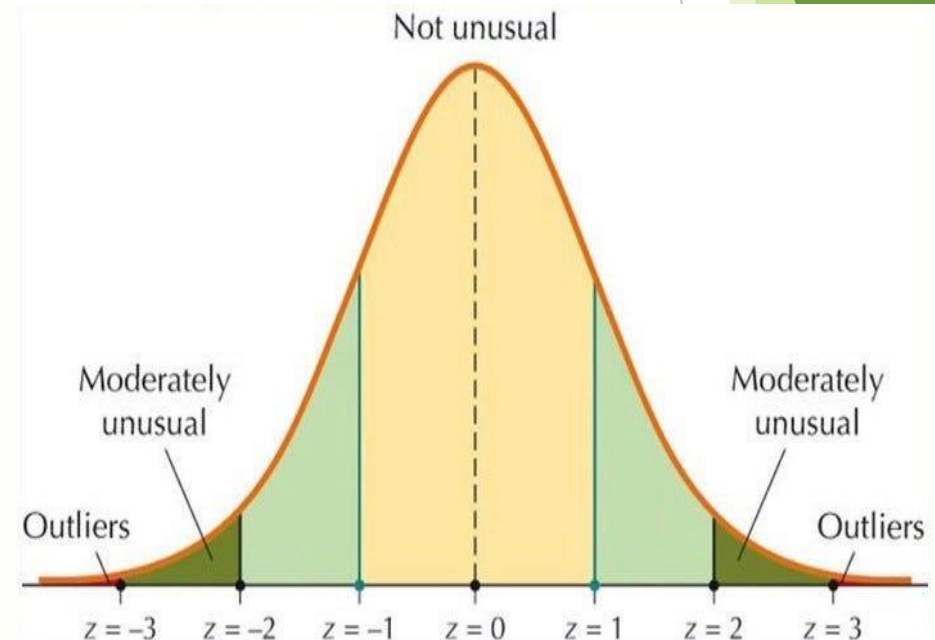
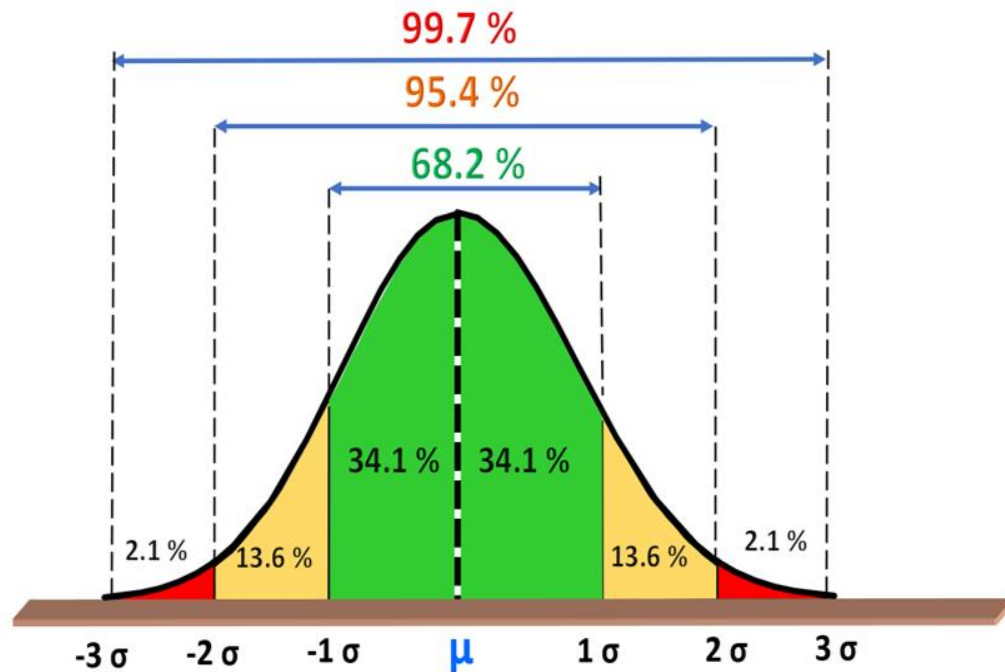
→ The 68-95-99 rule:

→ 68% of the data is within 1 standard deviation of the mean.

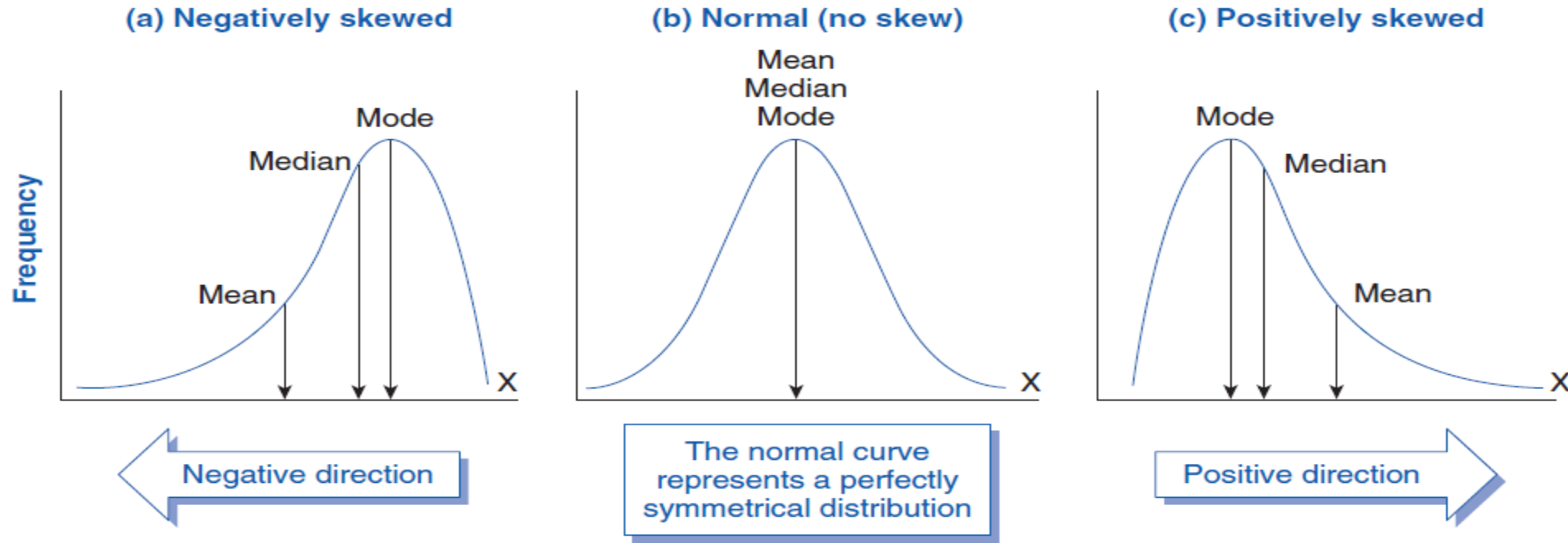
→ 95% of the data is within 2 standard deviation of the mean.

→ 99.7% of the data is within 3 standard deviation of the mean.

Numpy:
`np.randn()`



Skewed distribution



Left skewed:
 $\text{Mean} < \text{Median} < \text{Mode}$

Example: Exam scores where most students scored high but a few scored very low.

$\text{Mode} == \text{Median} == \text{Mean}$

Example: Height of individuals in a country

Right skewed:
 $\text{Mode} < \text{Median} < \text{Mean}$

Example: Income distribution where most earn less, and a few earn much more.