

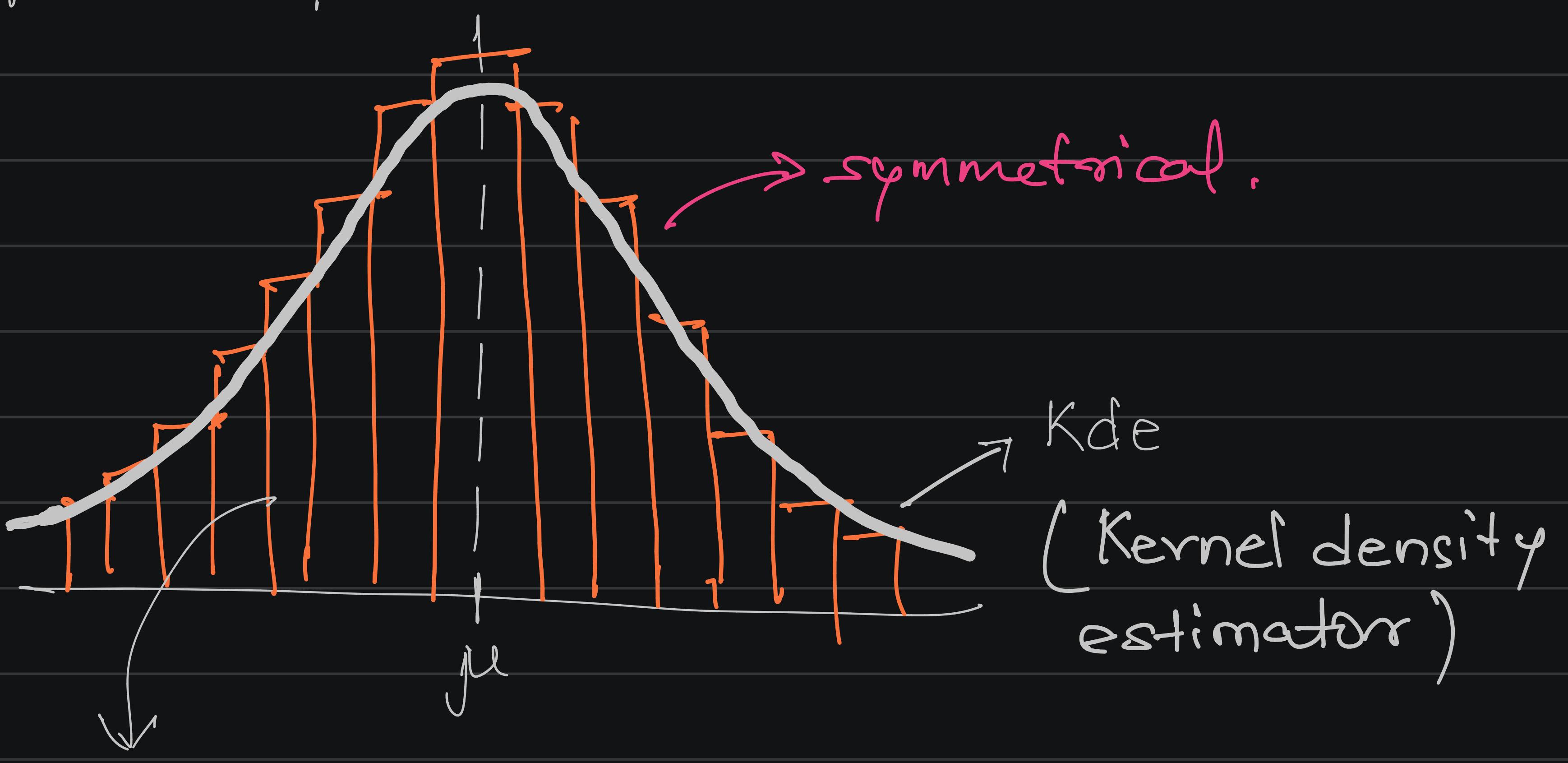
# Day 5

Agenda.

- (1) Normal distribution .
- (2) Standard Normal distribution -
- (3) Z - Score
- (4.) Standardization & Normalization



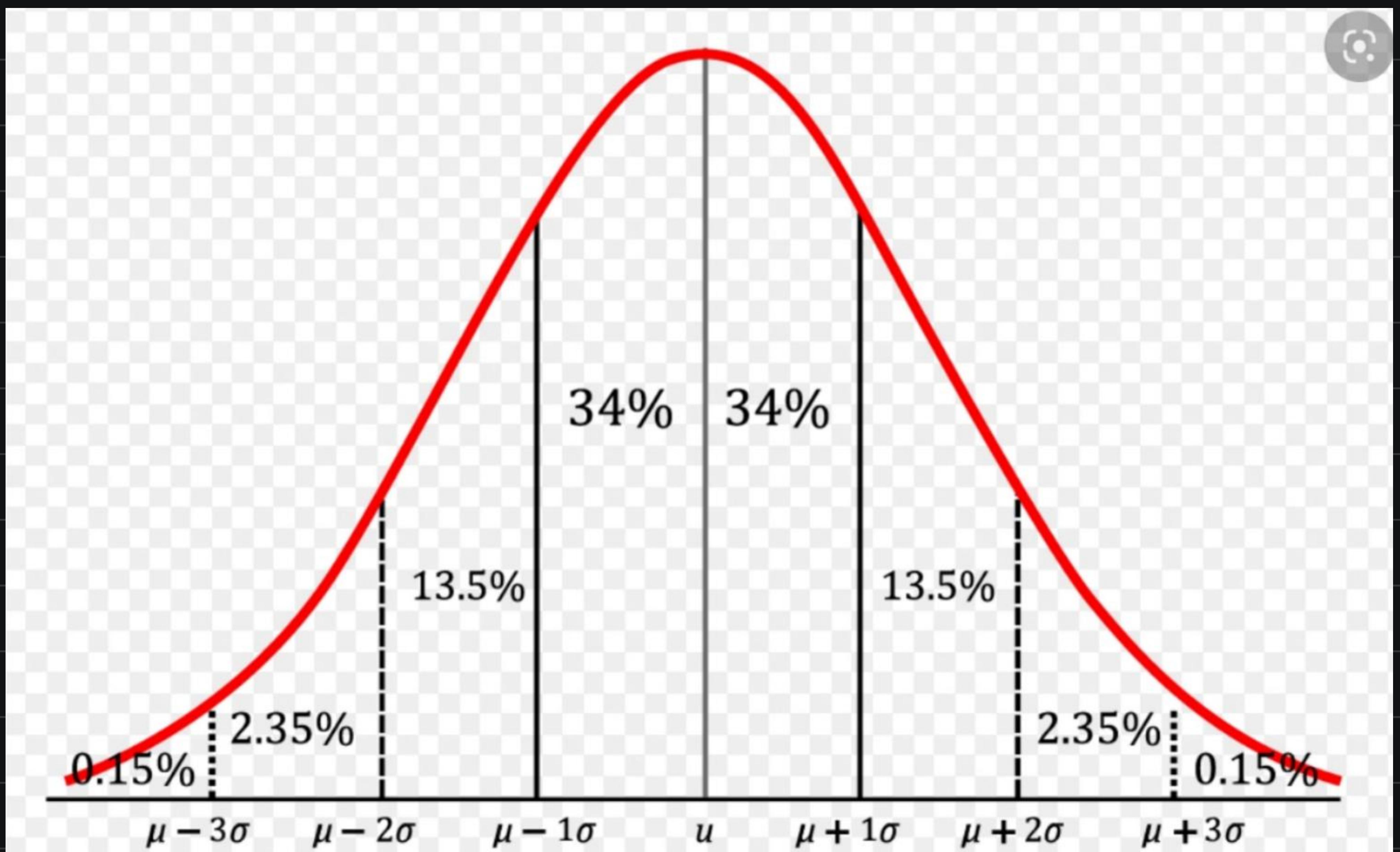
Gaussian / Normal distribution .



$$\text{Area under curve} = 1 (100\%)$$

e.g. Age, Weight, Height .  $\rightarrow$  follow  
Normal distribution .

## (\*) Empirical Rule of Normal distribution.



→ Assumptions

(i) 68% of data falls within 1 SD (Standard deviation)

(ii) 95% of data falls within 2 SD.

(iii) 99.7% of data falls within 3 SD

68 - 95 - 99.7% rule

→ Empirical Rule.

~~Q-Q~~

Q-Q plot → Can show if a distribution  
is Gaussian or not.

(2.)

Standard Normal Distribution.

if  $X$  belongs to Gaussian Distribution  $(\mu, \sigma)$

↓ can be converted into. using Z Score

$Y$   $\sim$  Standard Normal distribution  
 $(\mu = 0, \sigma = 1)$

e.g.  $X = \{1, 2, 3, 4, 5\}$

$$\mu = 3$$

$$\sigma = 1.41$$

$$Z\text{-Score} = \frac{x_i - \mu}{\sigma / \sqrt{n}}$$

$n = 1$ , applied on every element

⇒ Standard Error.

$$\text{So, } Z \text{ score} = \frac{x_i - \mu}{\sigma}$$

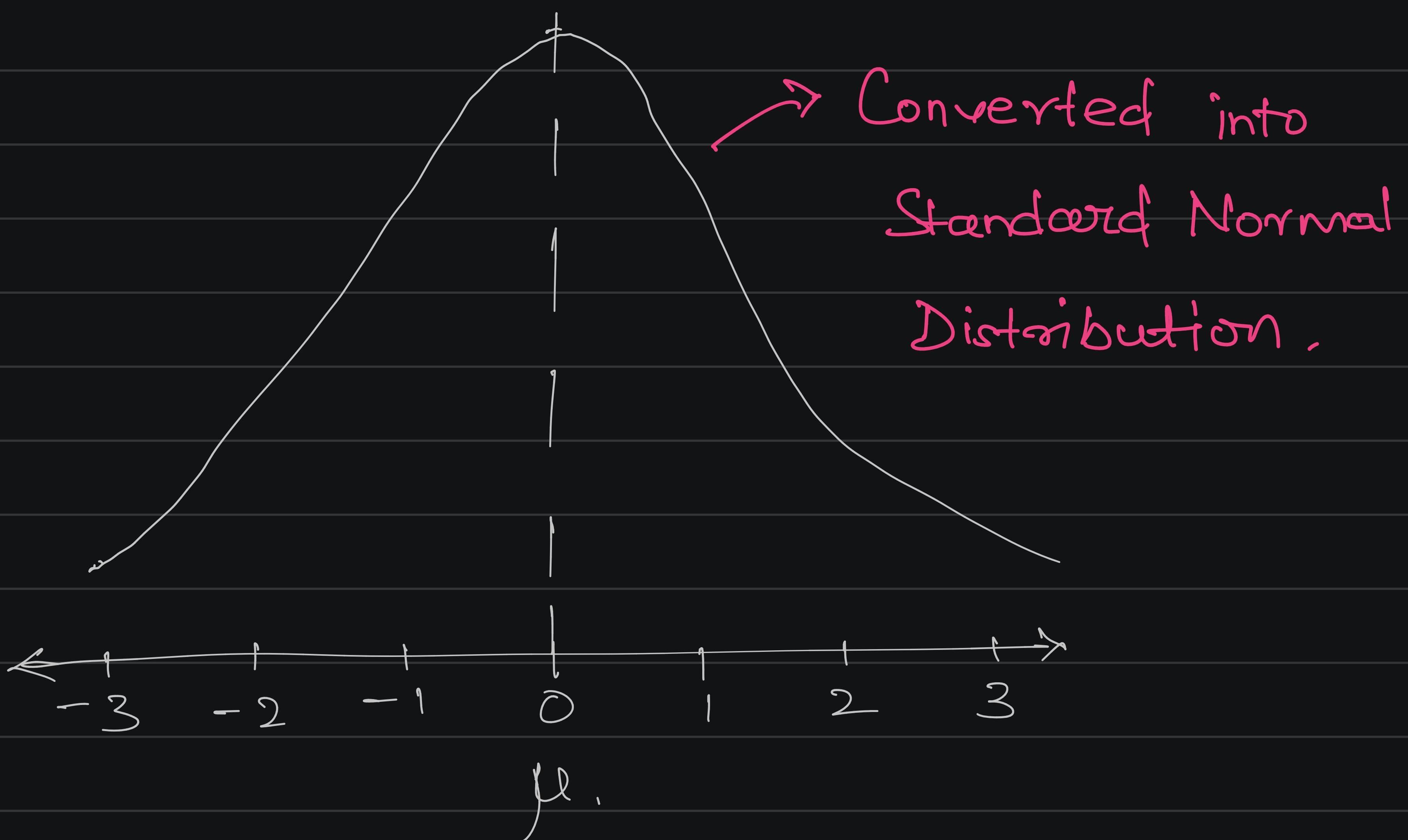
  $y = \{-1.414, -0.707, 0, 0.707, 1.414\}$

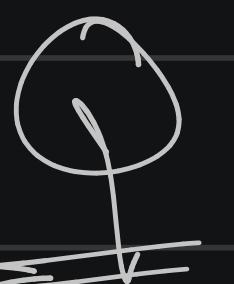
for 1<sup>st</sup> element  $Z\text{-score} = \frac{1-3}{1.414} = -1.414$

for 2<sup>nd</sup> element  $Z\text{-score} = \frac{2-3}{1.414} = -0.707$

for 3<sup>rd</sup> element  $Z\text{-score} = \frac{3-3}{1.414} = 0$

Distribution of  $y$ .



 Why do we need to convert ?

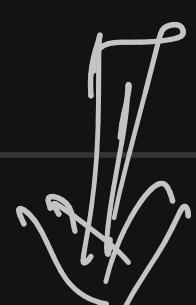
3 features →	Age (year)	height (kg)	height (cm)
	24	72	150
	26	78	160
	32	84	165
	33	92	170
	34	87	150
	28	83	180

29.

80.

175.

Units are different



Value ranges differ a lot.



Not favorable of ML models .



\* Apply Z-score on all Features to bring them all to the same scale (-3 to 3)

$$\mathcal{X} \mu = 0 \quad \mathcal{X} \sigma = 1$$

## Feature Scaling

① Standardization :- { z-score applied }

$$\mu = 0, \sigma = 1$$

$$\text{data} = \{-3 \text{ to } +3\} \quad 99.7\% \text{ data}$$

2 In Normalization → We give the range  
within which we need to  
convert the values.

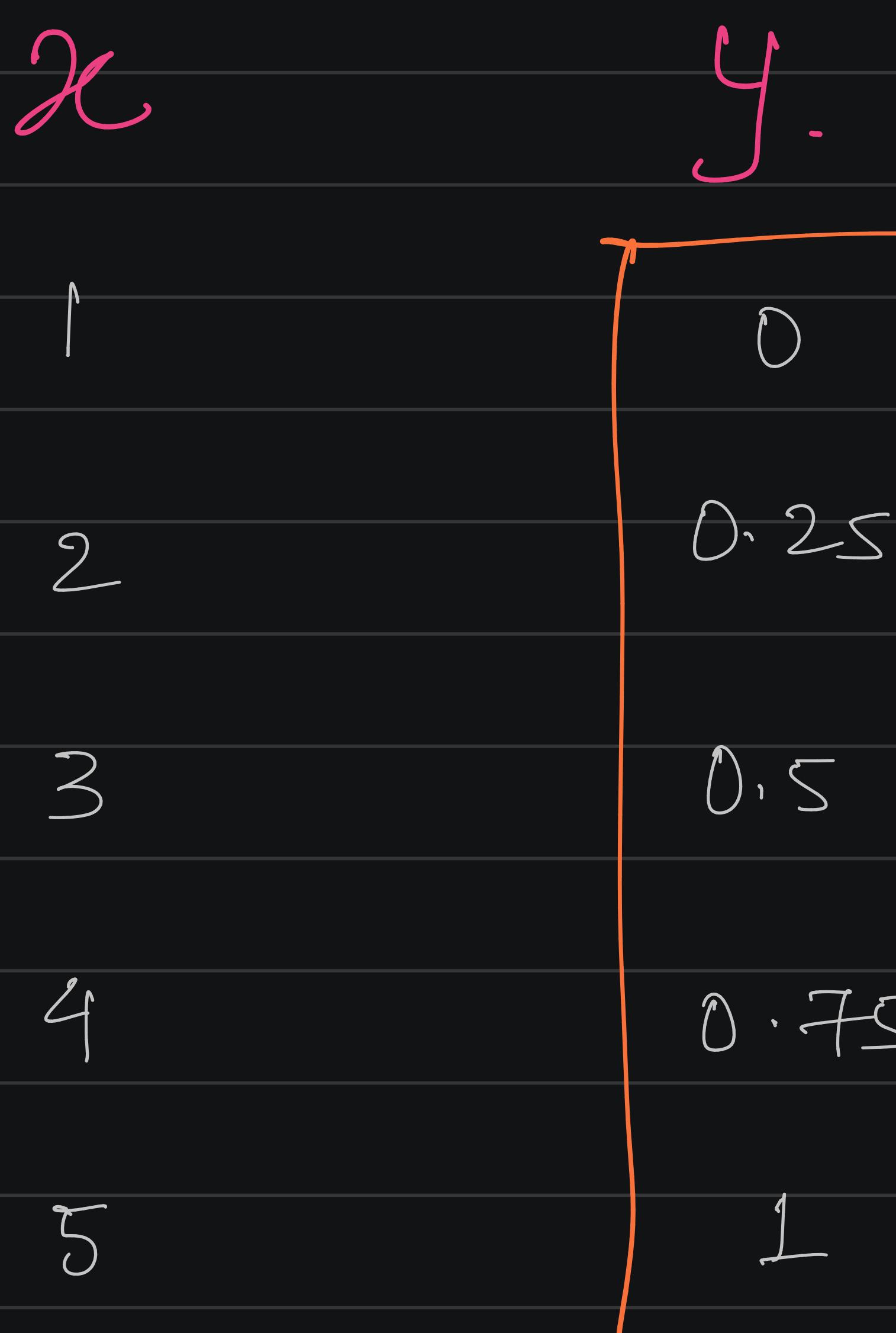
e.g. (1) Min-max scaler → transform values between  
0 to 1.

Formula,

$$x_{\text{scaled}} = \frac{x - x_{\min}}{x_{\max} - x_{\min}}$$

where,  
 $x_{\max} = 1$

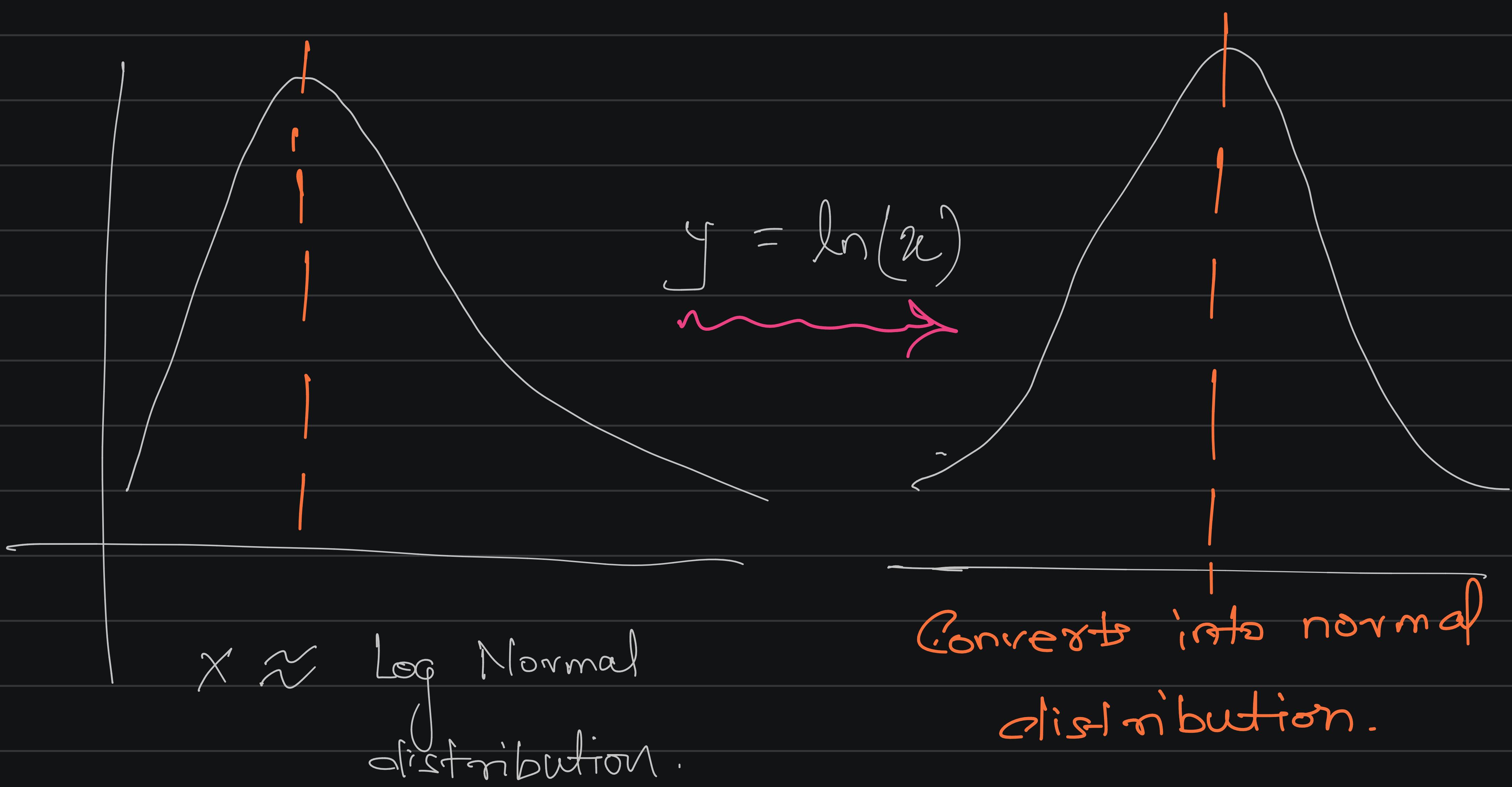
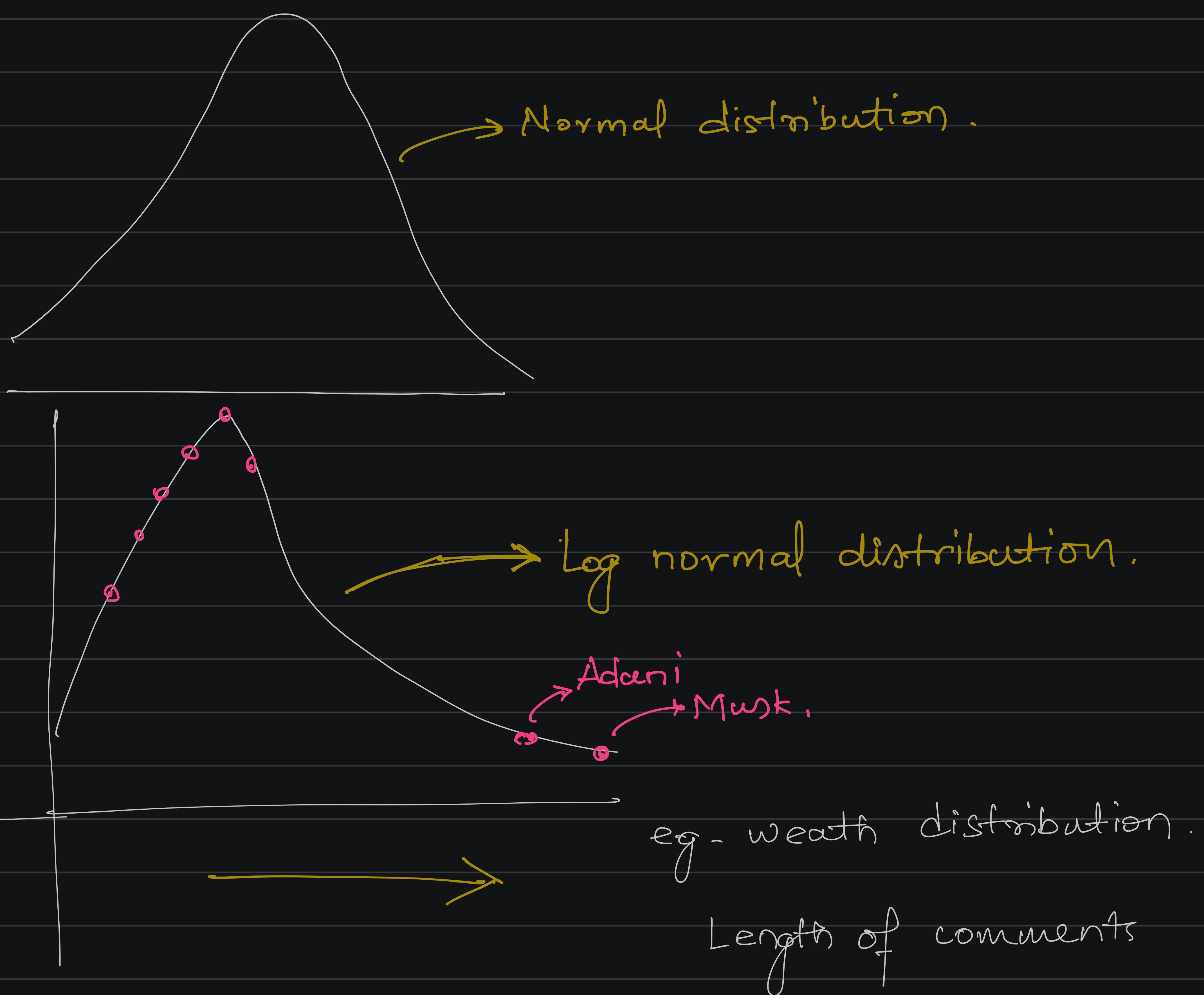
$x_{\min} = 0$



All values  
converted  
to 0 to 1.

Applied in Deep Learning

## Log Normal Distribution.

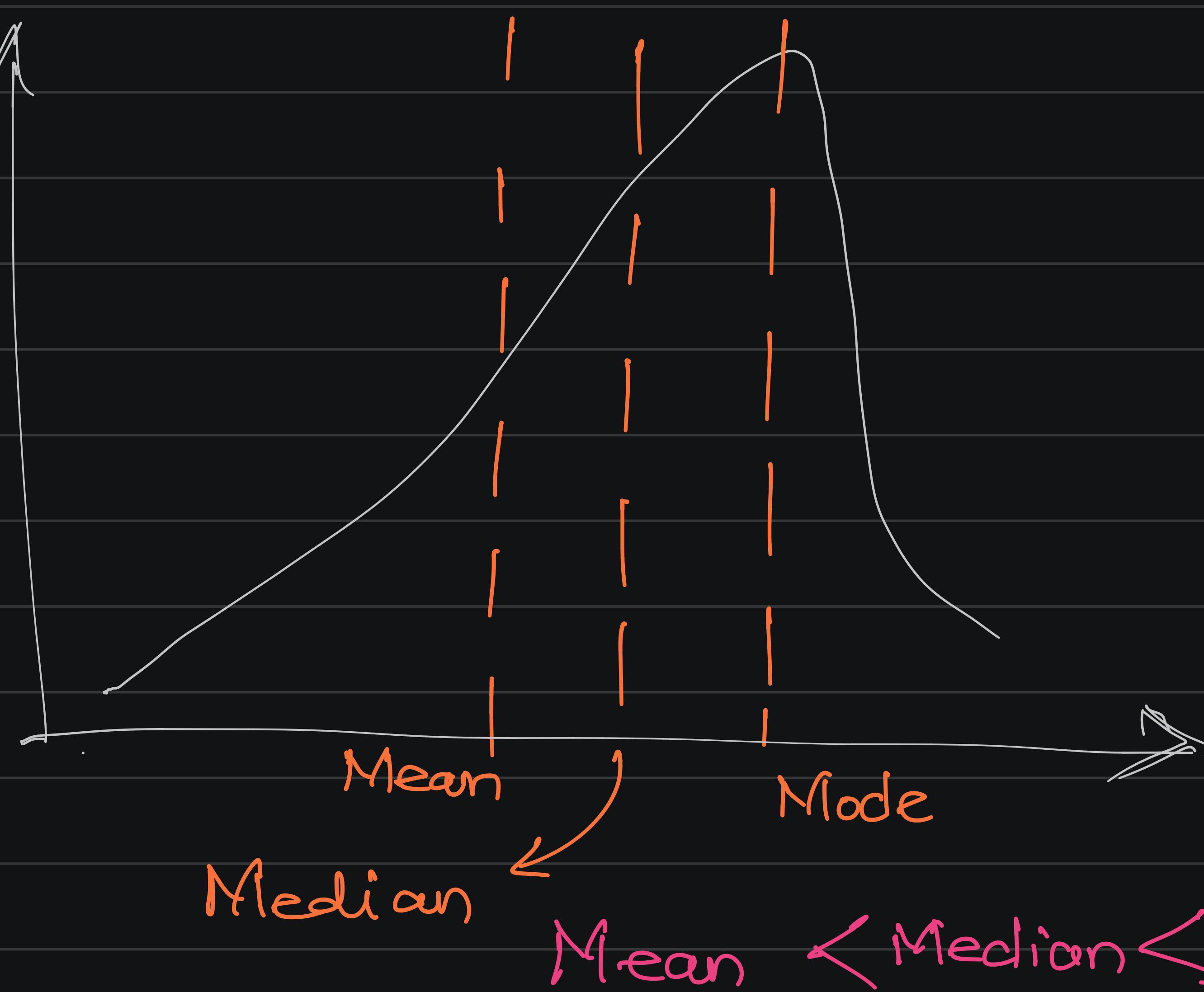


 = What is the relationship between mean, median and mode for below distributions

(1)



(2)



Mean < Median < Mode.



(2.) Use Z-table to find area under curve.

<b>z</b>	<b>0</b>	<b>0.01</b>	<b>0.02</b>	<b>0.03</b>	<b>0.04</b>	<b>0.05</b>	<b>0.06</b>	<b>0.07</b>	<b>0.08</b>	<b>0.09</b>
<b>+0</b>	.50000	.50399	.50798	.51197	.51595	.51994	.52392	.52790	.53188	.53586
<b>+0.1</b>	.53983	.54380	.54776	.55172	.55567	.55966	.56360	.56749	.57142	.57535
<b>+0.2</b>	.57926	.58317	.58706	.59095	.59483	.59871	.60257	.60642	.61026	.61409
<b>+0.3</b>	.61791	.62172	.62552	.62930	.63307	.63683	.64058	.64431	.64803	.65173
<b>+0.4</b>	.65542	.65910	.66276	.66640	.67003	.67364	.67724	.68082	.68439	.68793
<b>+0.5</b>	.69146	.69497	.69847	.70194	.70540	.70884	.71226	.71566	.71904	.72240
<b>+0.6</b>	.72575	.72907	.73237	.73565	.73891	.74215	.74537	.74857	.75175	.75490
<b>+0.7</b>	.75804	.76115	.76424	.76730	.77035	.77337	.77637	.77935	.78230	.78524
<b>+0.8</b>	.78814	.79103	.79389	.79673	.79955	.80234	.80511	.80785	.81057	.81327
<b>+0.9</b>	.81594	.81859	.82121	.82381	.82639	.82894	.83147	.83398	.83646	.83891
<b>+1</b>	.84134	.84375	.84614	.84849	.85083	.85314	.85543	.85769	.85993	.86214
<b>+1.1</b>	.86433	.86650	.86864	.87076	.87286	.87493	.87698	.87900	.88100	.88298
<b>+1.2</b>	.88493	.88686	.88877	.89065	.89251	.89435	.89617	.89796	.89973	.90147
<b>+1.3</b>	.90320	.90490	.90658	.90824	.90988	.91149	.91308	.91466	.91621	.91774
<b>+1.4</b>	.91924	.92073	.92220	.92364	.92507	.92647	.92785	.92922	.93056	.93189

0.59  $\Rightarrow$  59% → Area under curve for head.

$$\text{Area under curve} = (1 - 0.59) \\ (\text{tail}) = 41\%$$



