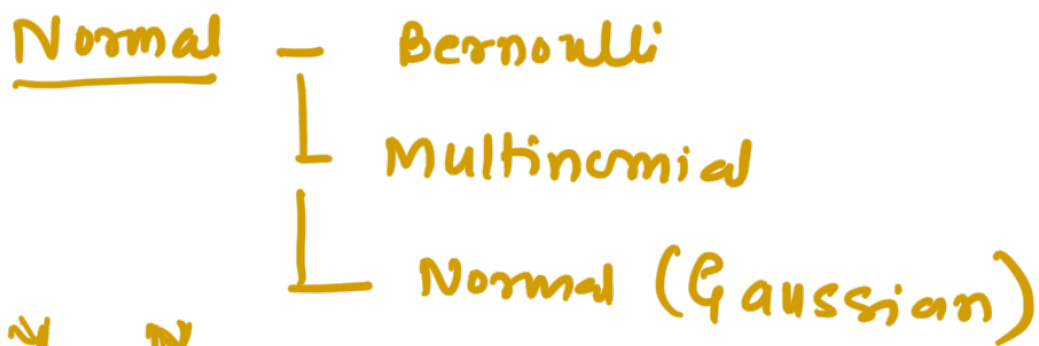


Distribution



Bernoulli (Binomial)

$$P(\text{Success}) = p$$

$$P(\text{failure}) = q = 1 - p$$

$$X = 1 \text{ [Success]}$$

$$X = 0 \text{ [failure]}$$

$$P(X=x) = p^x \cdot (1-p)^{1-x}$$

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

MultiNominal Distribution

— Discrete Count (Categorical) (Range)

$$P(X_1=x_1, \dots, X_k=x_k)$$

$$P(X=x_1, \dots, x_k) = \frac{n!}{x_1! \dots x_k!} p_1^{x_1} \dots p_k^{x_k}$$

eg.

BC	O	A	B	AB
P	0.4	0.42	0.10	0.02

Categories
Multiclass

$$4 \sim 50 \text{ AR} \Delta R? \quad \uparrow \frac{10}{20} \rightarrow 4 = 4/20$$

$$\sigma = \frac{1}{20} \quad \downarrow \quad \downarrow \quad \downarrow \quad \downarrow$$

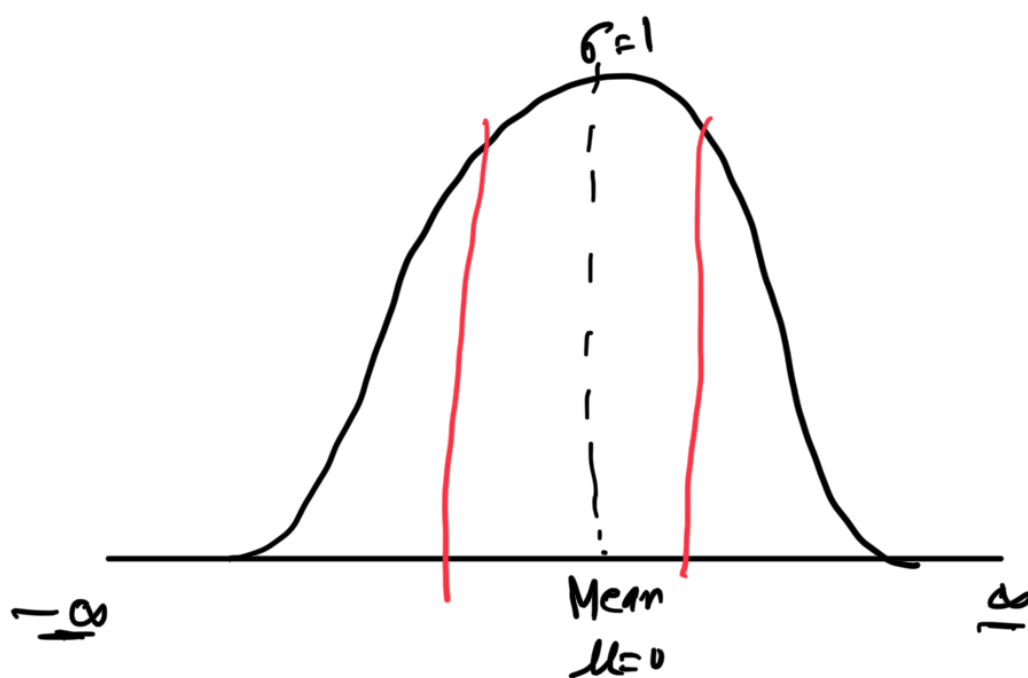
$$\frac{4}{20} \quad \frac{6}{20} \quad \frac{5}{20} \quad \frac{5}{20}$$

$$\begin{array}{|c|} \hline 6 \\ \hline 5 \\ \hline 20 \\ \hline \end{array} \quad \begin{array}{l} 6 = 6/20 \\ 5 = 5/20 \\ 5 = 5/20 \end{array}$$

Gaussian - Normal < Discrete
Continuous

— prob density function (pdf)

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



<u>Data</u> →	<u>Discrete</u> ✓	<u>Continuous</u>
---------------	-------------------	-------------------

x_1 ✓ p_1
 x_2 ✓ p_2
 x_3 ✓ p_3
 \vdots
 x_n ✓ p_n

pdf $\neq f(x)$

$$\sum_{i=1}^n p_i = 1$$

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

μ

$$\sum_{i=1}^n x_i p_i$$

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

σ

σ

n

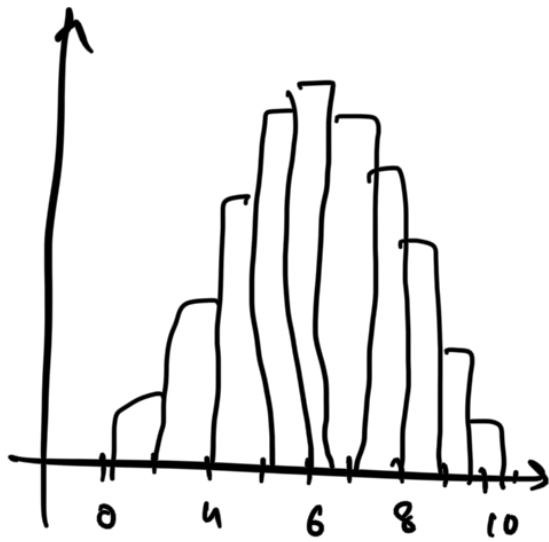
2

\uparrow

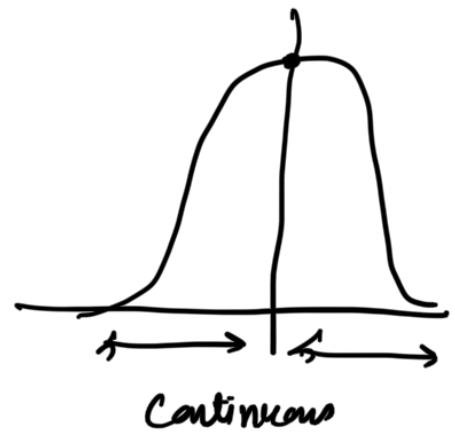
2

Variance $\sigma^2 = \sum_{i=1}^n (x_i - \mu)^2 p_i$

$\int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx$

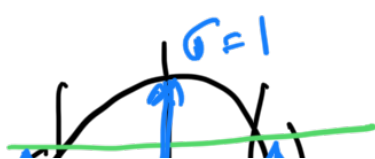
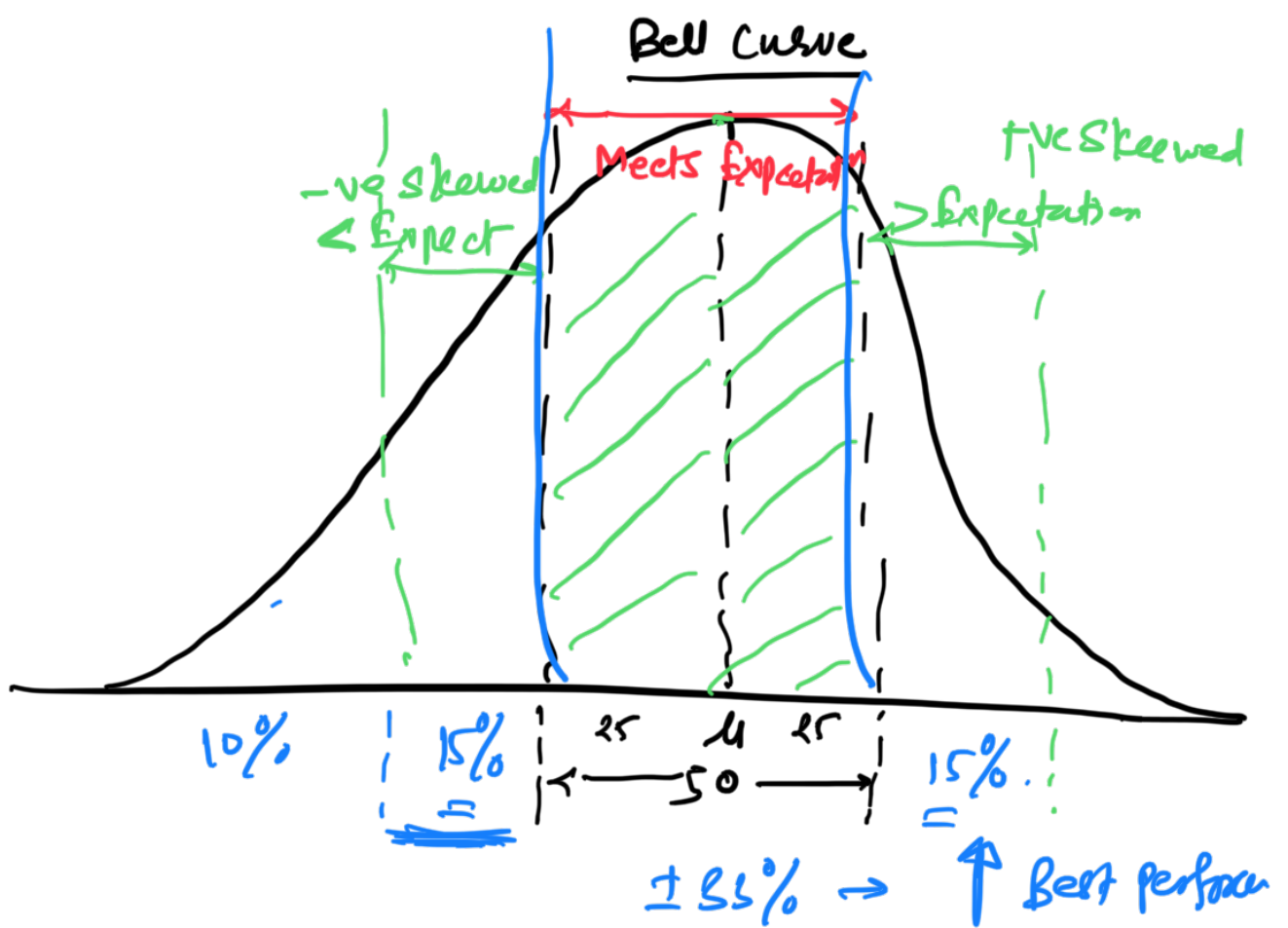
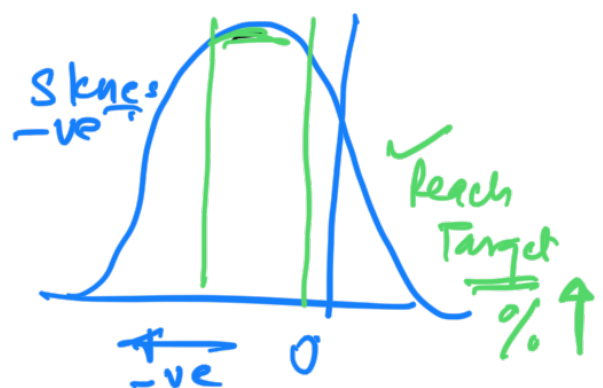
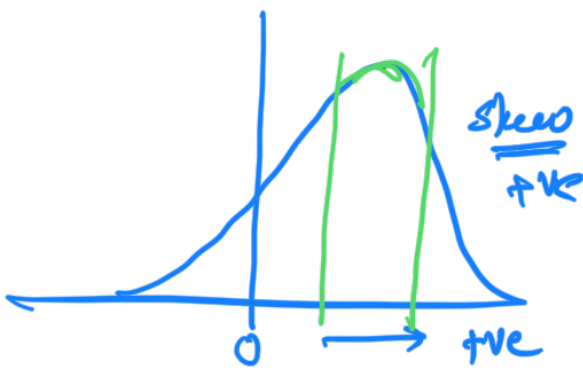
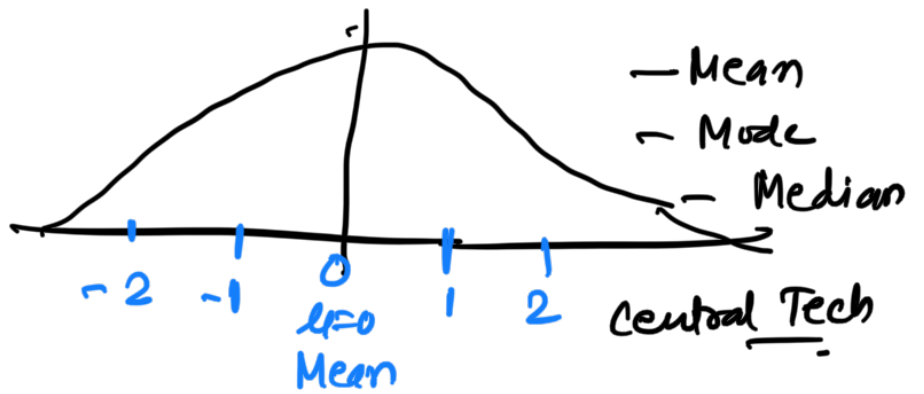


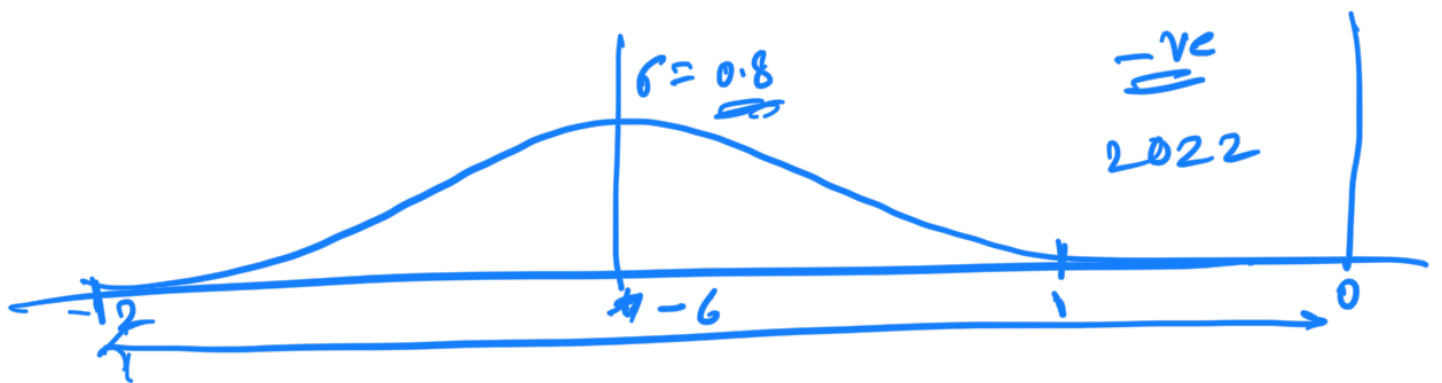
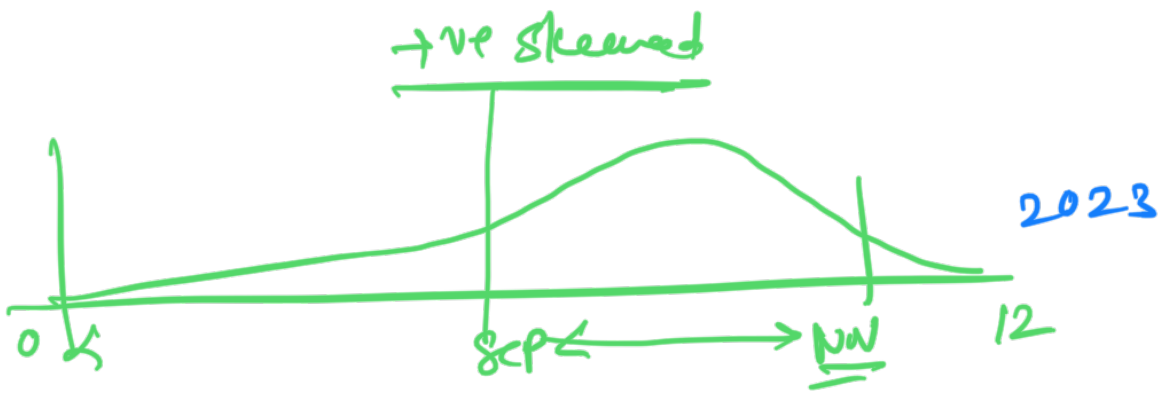
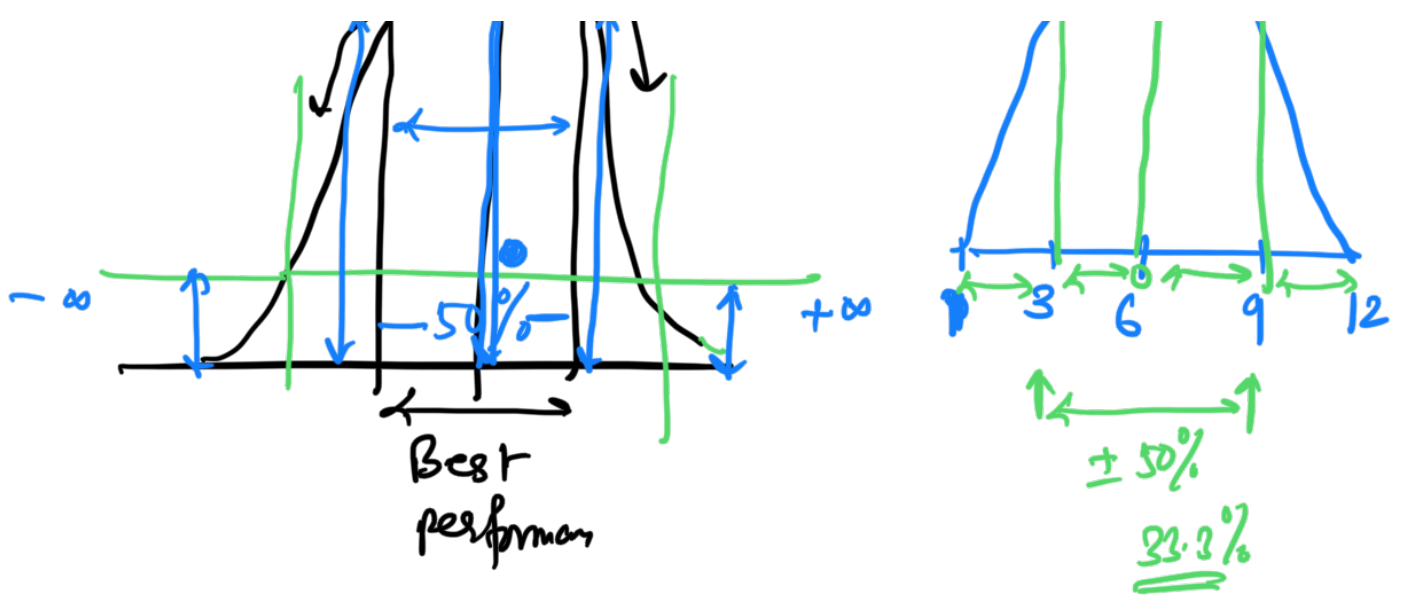
Discrete



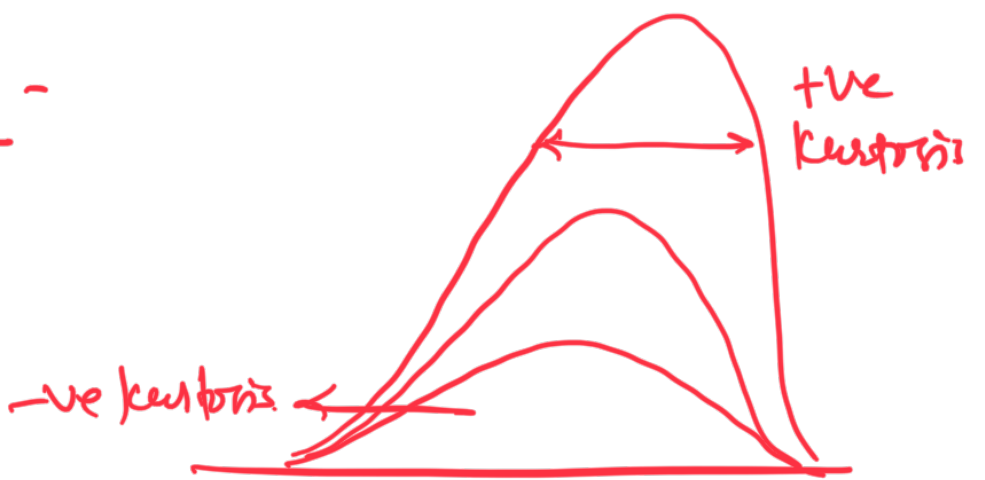
Statistics →

Case → data

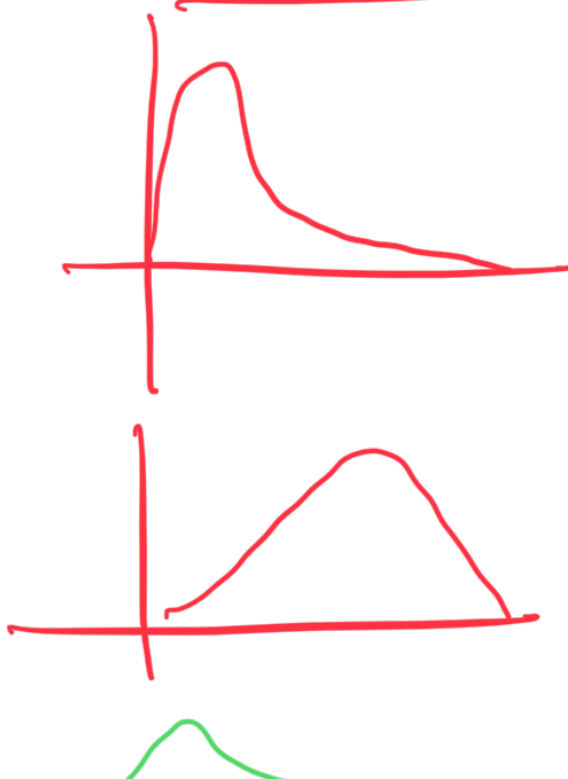




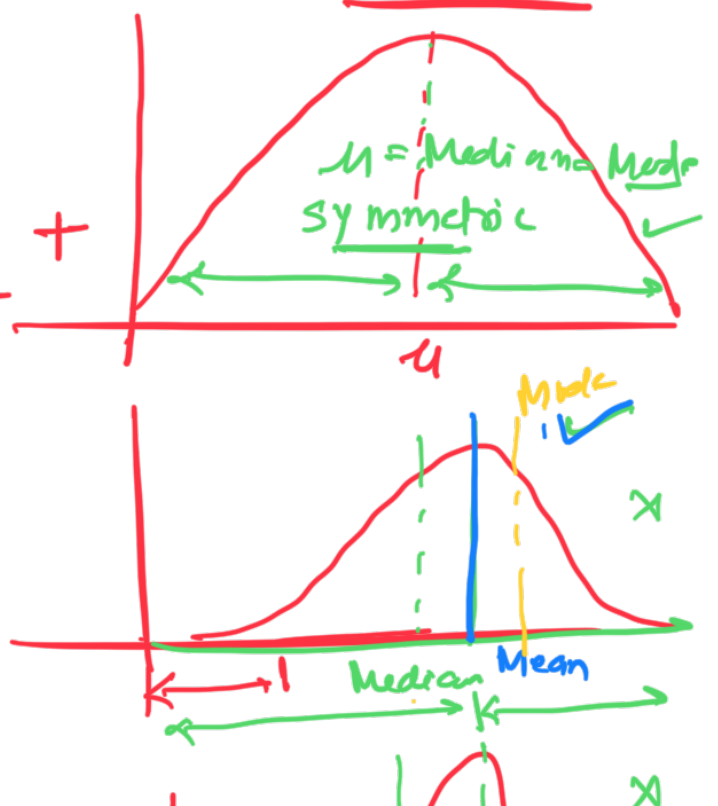
Decision Tree -



Skewness



Kurtosis

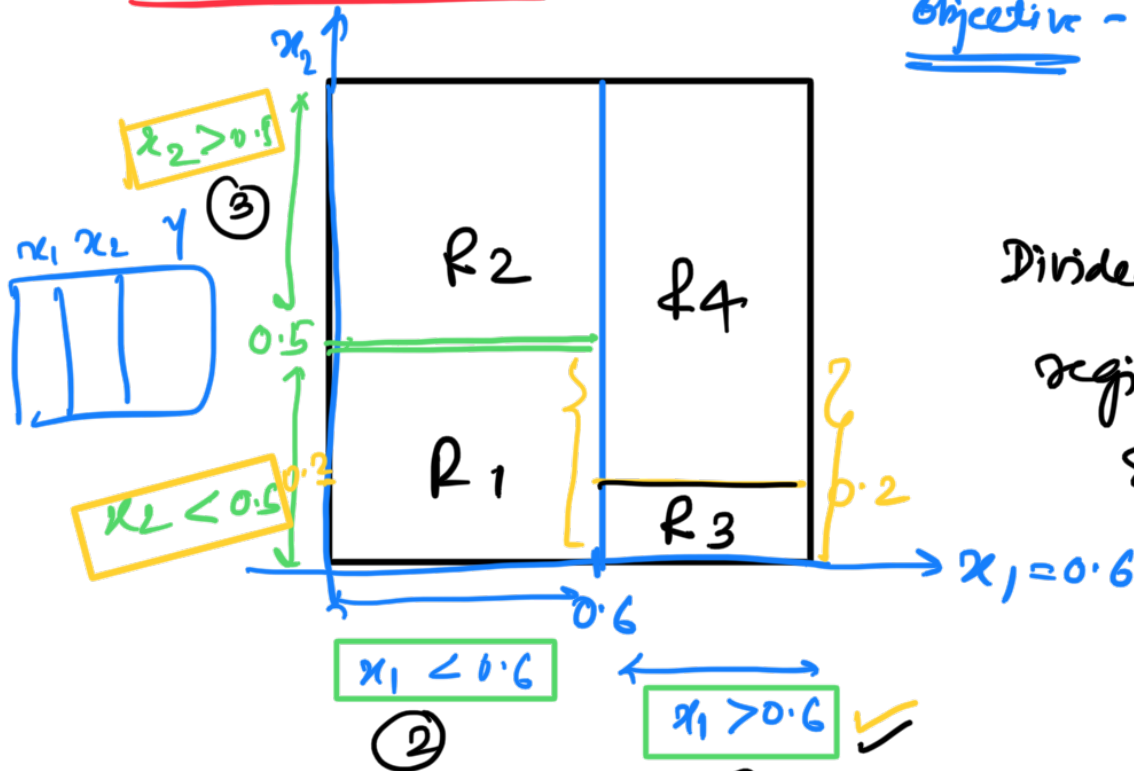


\times Mean $\rightarrow \checkmark$
 \checkmark Median $\rightarrow \checkmark$
 Mode



Mean < Mode < Median \Rightarrow \pm Skewed

Decision Tree



Objective - Partition of 2D space

Divided my interval regions into Subregion

Decision Tree

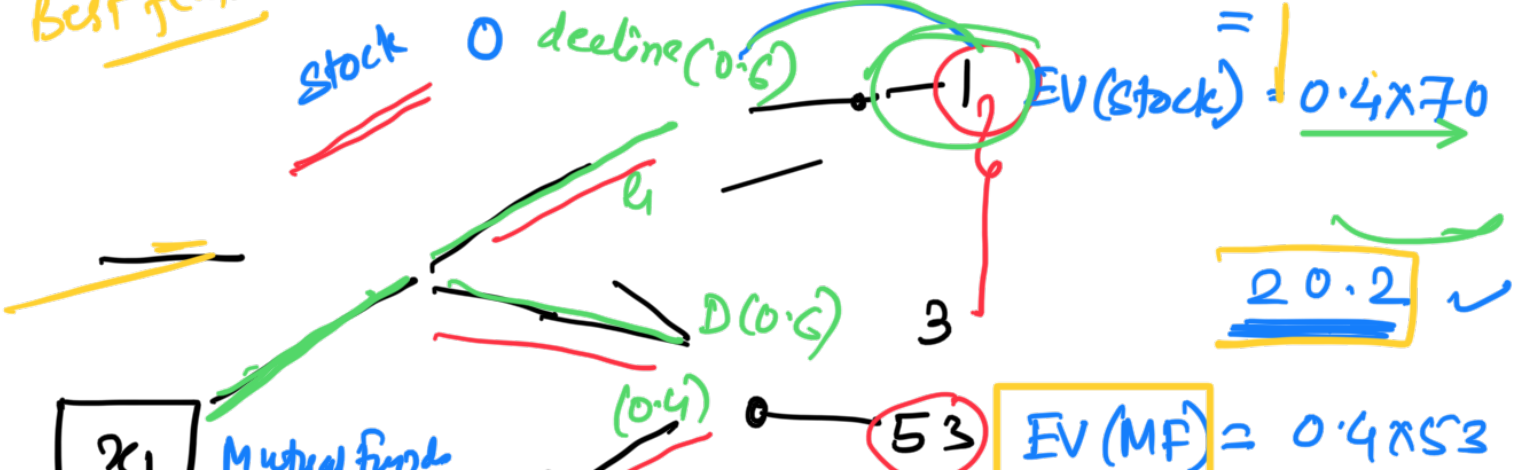
① $R = \{R_1, R_2, R_3, R_4\}$
 ② $x_2 = 0.5$
 ③ $x_2 < 0.2$ ✓
 ④ $x_2 > 0.2$ ✓

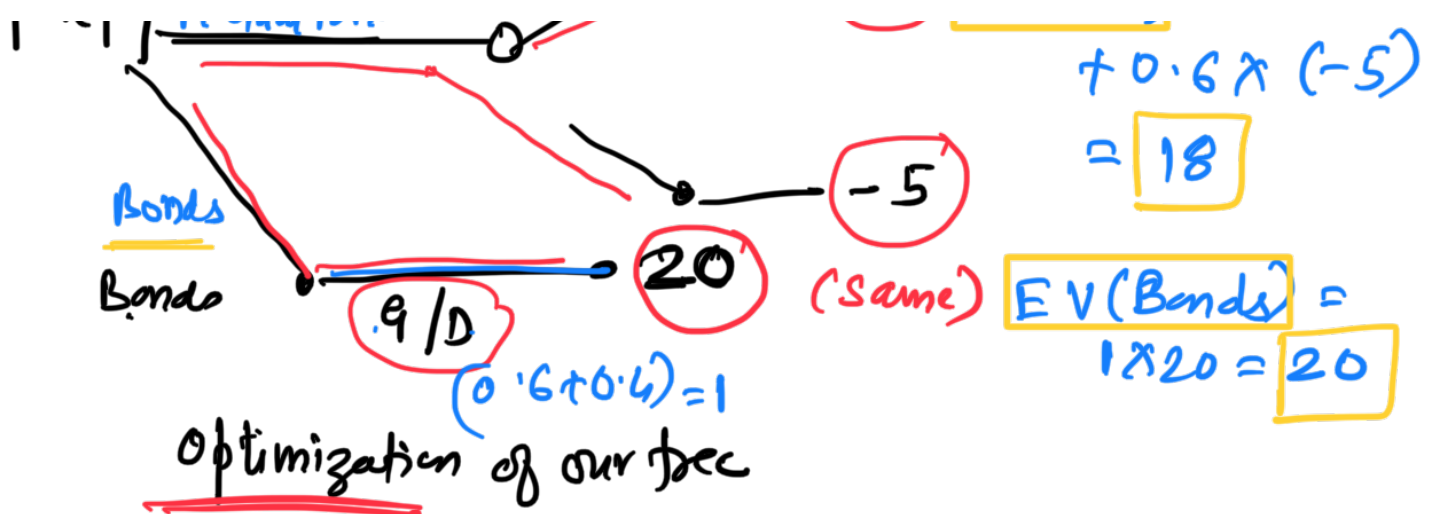
Eg \Rightarrow Payoff Tables

	Mutual funds	Stocks																		
Decision	<table border="1"> <tr> <th>Probability</th><th>Growing (0.4)</th><th>Declining (0.6)</th></tr> <tr> <th>Stocks</th><td>70</td><td>-5</td></tr> <tr> <th>Mutual funds</th><td>53</td><td>-3</td></tr> </table>	Probability	Growing (0.4)	Declining (0.6)	Stocks	70	-5	Mutual funds	53	-3	<table border="1"> <tr> <th>Probability</th><th>Growing (0.4)</th><th>Declining (0.6)</th></tr> <tr> <th>Stocks</th><td>70</td><td>-5</td></tr> <tr> <th>Mutual funds</th><td>53</td><td>-3</td></tr> </table>	Probability	Growing (0.4)	Declining (0.6)	Stocks	70	-5	Mutual funds	53	-3
Probability	Growing (0.4)	Declining (0.6)																		
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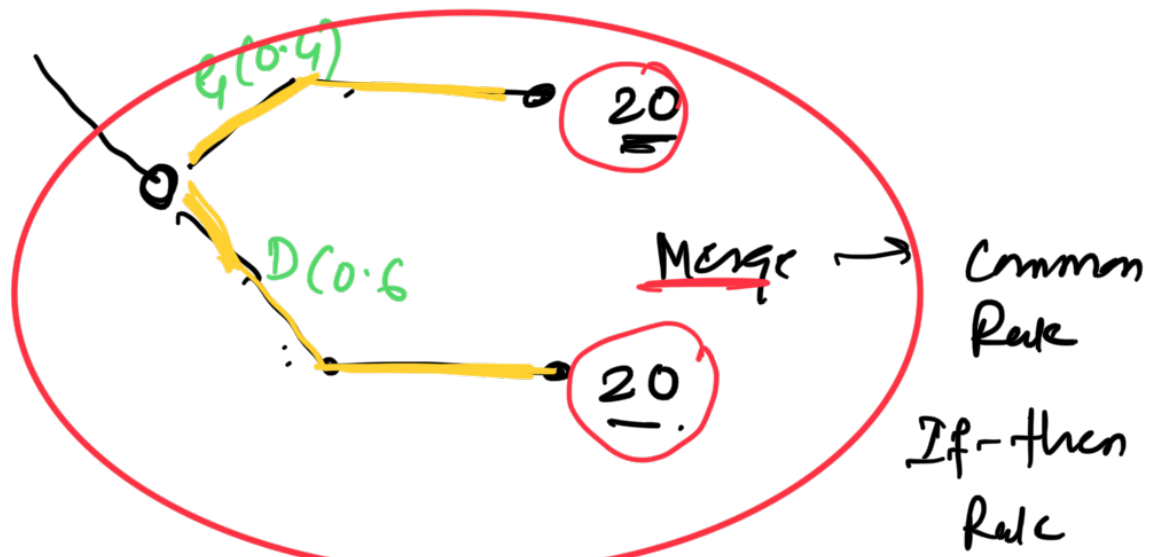
① Calculate feature Value
 $MF = 53 \times 0.4 + (-3) \times 0.6$
 $= 20.2$
 Bonds $\Rightarrow 20$

Best feature $\rightarrow EV \equiv \text{Max}$
 Stock \Rightarrow decline (0.6) \rightarrow 1
 EV(Stock) $= 0.4 \times 70 + 0.6 \times -3$
 $= 20.2$ ✓

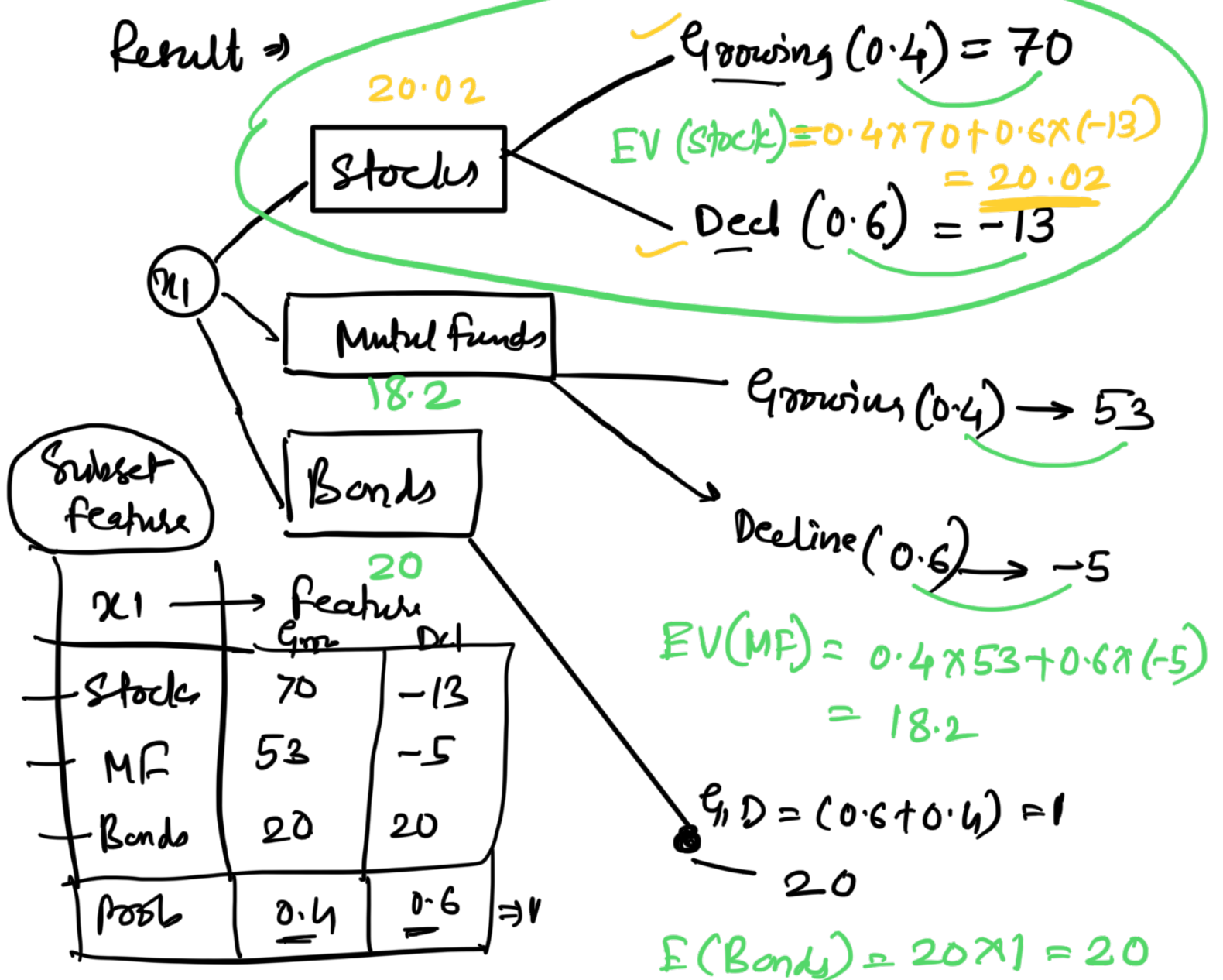




Tree Pruning



Result ⇒

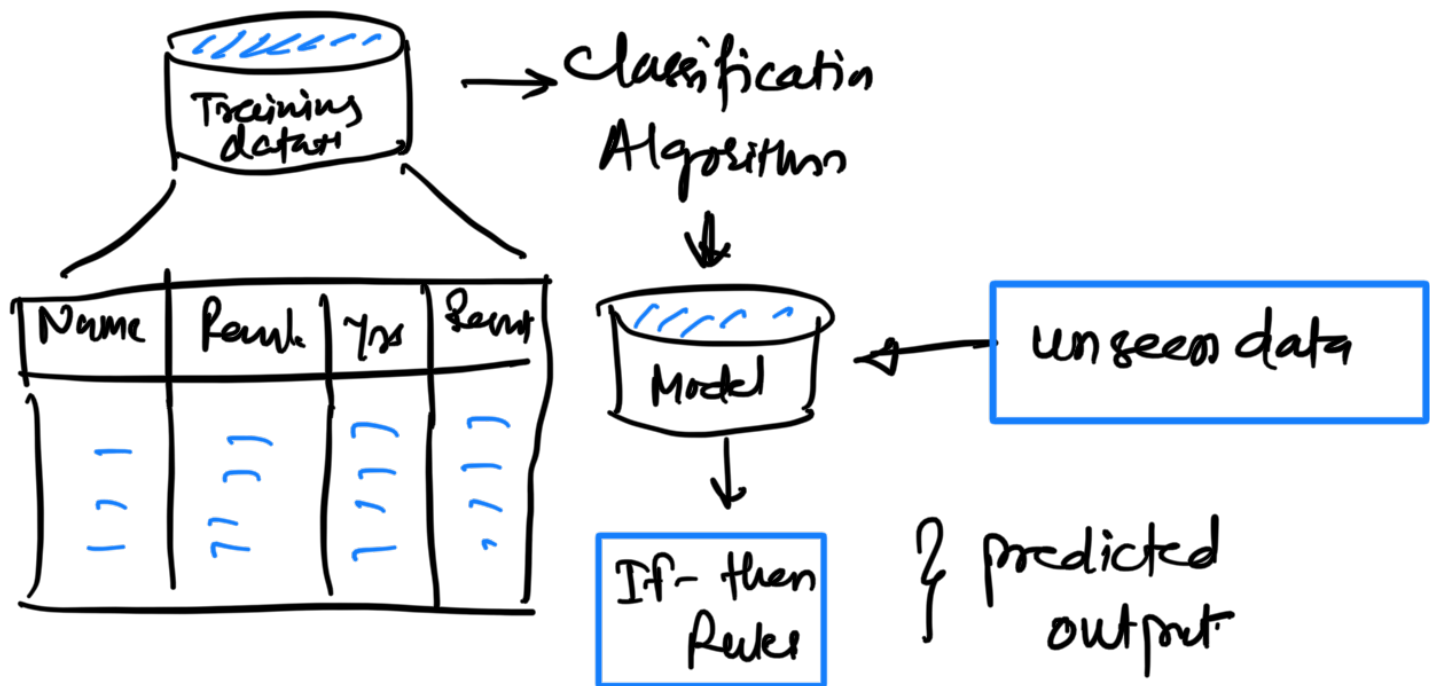


Stock, MF, Bonds?

Stocks = 20.02 → Stock > MF & Bonds
MF = 18
Bonds = 20

Important feature

Purpose → Identify the most important features in subset list



Root Node → Start of DT

Node → Condition spe

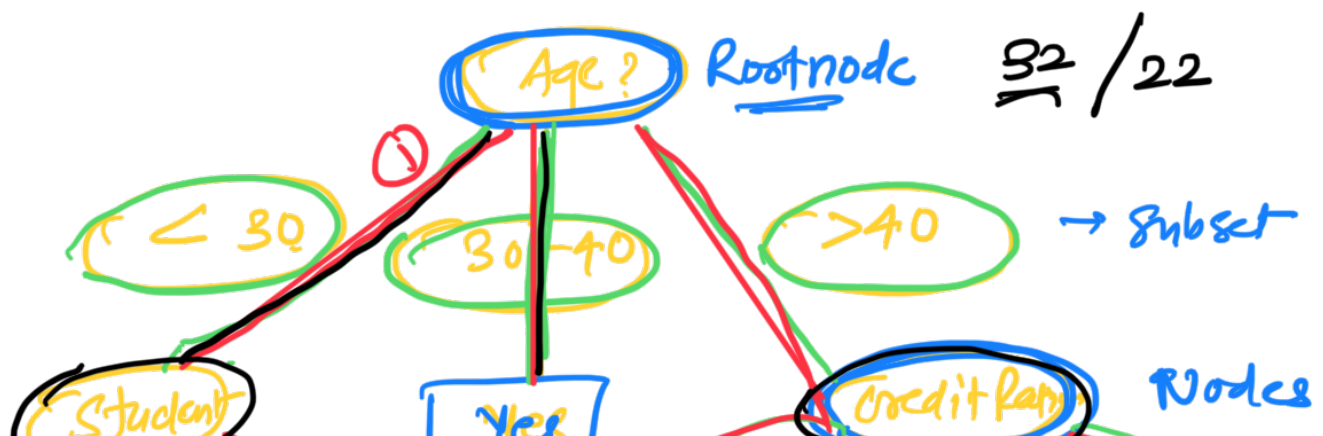
Age → < 30 , $30-40$, > 40 —

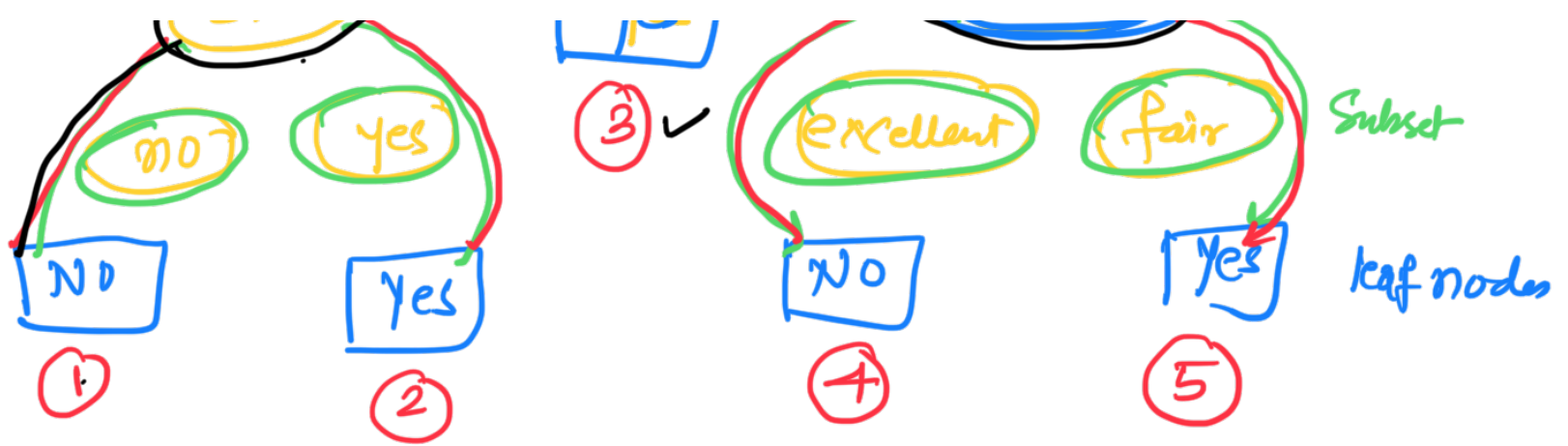
Ex

Age	Income	Student	Credit ratings	buys computer

x_1 Age = $\{< 30, 30-40, > 40\}$ ✓
 x_2 Income = $\{high, medium, low\}$ ✓
 x_3 Student = $\{yes, no\}$ ✓
 x_4 Credit = $\{fair, excellent\}$ ✓

$y \leftarrow \text{buy} = \{yes, no\}$ ✓
 yes — $P(1)$ ✓
 no — $P(0)$ ✓





→ Age = 2, Student = no, Income = high.
unknown

Credit rating = fair ⇒ buys = ? yes
No