

De Morgan's Law:

Union over $n = \bigcup_n$

$$1) \left(\bigcup_n A_n \right)^c = \bigcap_n A_n^c$$

$$\begin{aligned} \text{Suppose } x &\in \left(\bigcup_n A_n \right)^c \\ &= x \notin \left(\bigcup_n A_n \right) \\ &= x \notin A_n \quad \forall n \\ &= x \in A_n^c \quad \forall n \\ &= x \in \bigcap_n A_n^c \end{aligned}$$

$$(2) \left(\bigcap_n A_n \right)^c = \bigcup_n A_n^c$$

Sigma Fields or Sigma Algebra of sets

- 1) $\mathcal{S} \in \mathcal{f} \leftarrow$ sigma field. 3.) If $A_1, A_2 \in \mathcal{f}$ then $\bigcup_n A_n \in \mathcal{f}$
- 2.) If $A \in \mathcal{f}$ then $A^c \in \mathcal{f}$

Consider the experiment, rolling a die

$$\mathcal{S} \in = \{1, 2, 3, 4, 5, 6\}$$

$$\text{If } A = \{ \text{No. showing up} > 2 \}$$

$$A = \{3, 4, 5, 6\}$$

$$A^c = \{1, 2\}$$

Consequences of properties 1-(3)

$$(1) \quad \phi \in \mathcal{f} \quad (\text{since } \phi = \emptyset^c)$$

$$(2) \quad A_1, A_2, \dots \in \mathcal{f}, \text{ then } \bigcap_n A_n \in \mathcal{f}$$

$$\text{Since } A_1, A_2, \dots \in \mathcal{f}$$

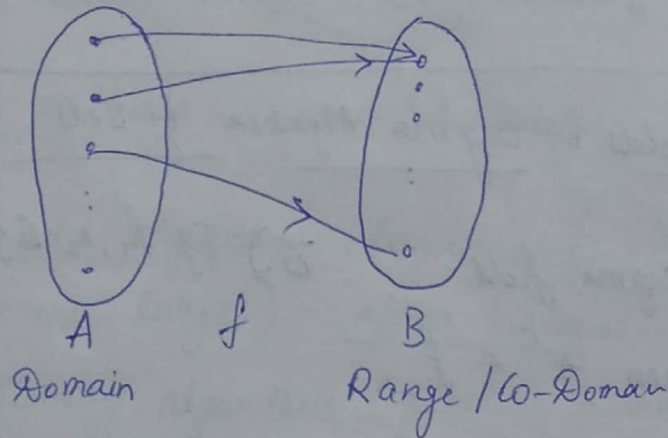
$$\Rightarrow A_1^c, A_2^c, \dots \in \mathcal{f} \quad (\text{I})$$

$$\Rightarrow \bigcup_n A_n^c \in \mathcal{f} \quad (\text{II})$$

$$\Rightarrow \left(\bigcup_n A_n^c \right)^c \in \mathcal{f} \quad (\text{III})$$

$$\Rightarrow \bigcap_n A_n \in \mathcal{f} \quad [\text{By De Morgan's Law}]$$

Function:



Probability:

$$P: \mathcal{f} \rightarrow [0, 1]$$

\mathcal{f} - sigma field.

Properties:

P1. $P(S) = 1$

P2. If $A, B \in \mathcal{F}$ and $A \cap B = \emptyset$,

then $P(A \cup B) = P(A) + P(B)$

Corollary:

~~P3~~ If $A, B \in \mathcal{F}$

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

Eg:-

Consider a ~~fair~~ fair die

Event $A: \{x > 2\}$

$$P(A) = P(3) + P(4) + P(5) + P(6) = 4\left(\frac{1}{6}\right) = \frac{2}{3}$$

$$P(B) = P(2) + P(1) = 2\left(\frac{1}{6}\right) = \frac{1}{3}$$

$$P(A^c) = 1 - P(A) \quad \text{Since } (A \cup A^c = S)$$

→ Simplest σ -field

$$\mathcal{F} = \{S, \emptyset\}$$

→ Power Set: (2^S)

$$\mathcal{F} = \{\text{All possible subsets of } S\}$$

~~If S contains, then number of no of su~~

If S contain n elements, then no of subset = 2^n

Random Variables:

$$X: S \longrightarrow \mathbb{R}$$

$$\left\{ \omega \in S \mid X(\omega) = c, \right\} \in \mathcal{F} \\ c \in \mathbb{R}$$

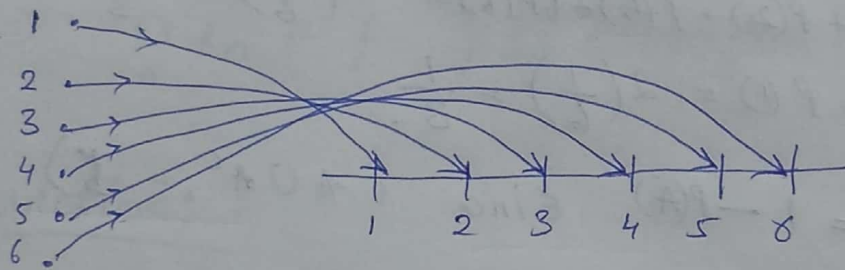
Where 'S' is the sample space i.e., the set that contains all the outcomes of expt

Consider the die rolling experiment:

$X =$ ~~No~~ the shows up on the face

$$S = \{1, 2, 3, 4, 5\}$$

$$X(1) = 1 \quad ; \quad X(2) = 2 \quad ; \quad X(3) = 3 \quad \dots \quad X(6) = 6$$



Eg: $P(X=1) = P(\{\omega \in S \mid X(\omega) = 1\})$

$\omega=1$ $P(1) = 1/6.$

Conditional Probabilities:

Suppose A & B are two events

$$P(A|B) = \begin{cases} \frac{P(A \cap B)}{P(B)} & \text{if given } P(B) > 0 \\ \text{Undefined.} & \text{otherwise} \end{cases}$$

Consider a die, that is rolled an infinite number of times,

(i) $A = \text{A six occurs on every roll}$

$$P(A) = \lim_{n \rightarrow \infty} \left(\frac{1}{6}\right)^n = 0 \quad \leftarrow \text{zero prob.}$$

$$\begin{aligned} \text{(ii) } P(3 | \text{odd outcome}) &= \frac{P(\{3\} \cap \{1, 3, 5\})}{P(\{1, 3, 5\})} = \frac{\frac{1}{6}}{\frac{3}{6}} = \frac{1}{3} \\ &= \frac{\frac{1}{6}}{\frac{1}{2}} = \underline{\underline{\frac{1}{3}}} \end{aligned}$$

$$\begin{aligned} \text{(iii) } P(X \text{ is odd} | X \text{ is multiple of 3}) \\ &= \frac{P(A \cap B)}{P(B)} = \frac{P(\{3\} \cap \{3, 6\})}{P(\{3, 6\})} = \frac{\frac{1}{6}}{\frac{2}{6}} = \underline{\underline{\frac{1}{2}}} \end{aligned}$$

We say that events A & B are independent events if

$$\Rightarrow P(A|B) = P(A)$$

$$\Rightarrow \frac{P(A \cap B)}{P(B)} = P(A)$$

$$\Rightarrow \boxed{P(A \cap B) = P(A)P(B)}$$

When we have more than 3 events, A, B, C are independent if:

$$\Rightarrow P(A \cap B \cap C) = P(A)P(B)P(C)$$

$$\Rightarrow P(A \cap B) = P(A)P(B); P(B \cap C) = P(B)P(C); P(C \cap A) = P(C)P(A)$$

If A, B & C are events,

$$P(A \cap B | C) = P(A | B \cap C) \cdot P(B | C)$$

$$\frac{P(A \cap B)}{P(B)}$$

~~not~~

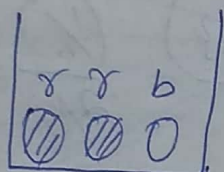
taking LHS

$$\frac{P(A \cap (B \cap C))}{P(C)} = \frac{P(A | B \cap C) \cdot P(B \cap C)}{P(C)}$$

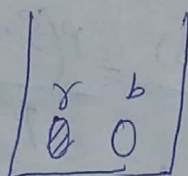
$$= P(A | B \cap C) \cdot P(B | C)$$

Eg:

Urn #1



Urn #2



Expt: Pick one of the urns at random & pick a ball from that urn

$$S = \{1r, 1b, 2r, 2b\}$$

Alternatively, $S = S_1 \times S_2$

$$\text{Where } S_1 = \{1, 2\}, S_2 = \{r, b\}$$

Let X_1 be out of urn S_1

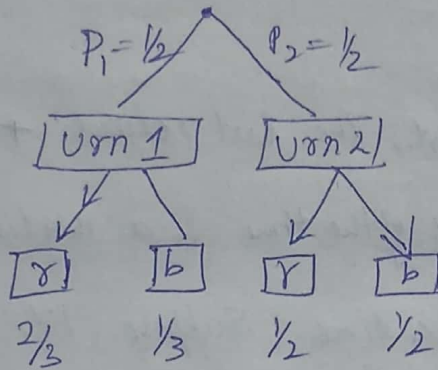
X_2 be the outcome of ~~urn~~ ball S_2

$$P(1r) = P(X_1=1, X_2=r) = P(X_1=1) \cdot P(X_2=r | X_1=1)$$

← prob of selectg urn
 → prob of r given urn 1

$$= \frac{1}{2} \cdot \frac{2}{3} = \frac{1}{3}$$

$$\begin{aligned} P(2b) &= P(X_1=2) \cdot P(b|X=2) \\ &= \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4} \end{aligned}$$



Bayes Theorem: Suppose, A_1, A_2, \dots is a collection of events such that $A_i \cap A_j = \emptyset \forall i \neq j$

$$\text{w.k.t } \bigcup_i A_i = S$$

Suppose B is another event,

$$\begin{aligned} \text{Then } P(B) &= P(B \cap S) \\ &= P(B \cap \bigcup_i A_i) \\ &= P(\bigcup_i (B \cap A_i)) \quad \text{--- } (*) \end{aligned}$$

Since A_i are disjoint,
 $B \cap A_i, i \geq 1$ are also disjoint
 $\bigcup_i (B \cap A_i) = B$

From (*)

$$\begin{aligned} P(B) &= \sum_i P(B \cap A_i) \\ &= \sum_i P(B|A_i) \cdot P(A_i) \end{aligned}$$

$$\text{Consider } P(A_j | B) = \frac{P(A_j \cap B)}{P(B)}$$

$$= \frac{P(A_j) P(B | A_j)}{\sum_i (P(B | A_i) P(A_i))}$$

Example:

A hospital conducts a drug test. The test returns +ve result for a drug user 99% of the time & a negative result for a non-user 95% time. Suppose 1% of population uses drug.

Q: What is the prob that an individual is a drug user given that he tests positive.

Soln :-

Case

$S = \{ \text{User}+, \text{user}-, \text{non-user}+, \text{non user}- \}$

$$P(+ | \text{User}) = 0.99$$

$$P(- | \text{nonuser}) = 0.95$$

$$P(u | +) = \frac{P(+ | \text{User}) \cdot P(\text{User})}{P(+ | \text{User}) \cdot P(\text{User}) + P(+ | \text{non-user}) \cdot P(\text{non user})}$$

$$= \frac{0.99 \times 0.01}{0.99 \times 0.01 + (0.05)(0.99)}$$

$$= \frac{99 \times 1}{99 \times 1 + 99 \times 5} = \frac{1}{6}$$

Expectations:

Suppose X is a random variable,

$E[X]$ or μ_x is called the expected value of mean of X .

For the case of a discrete random value x ,

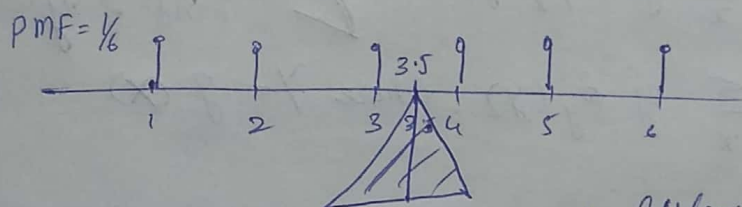
$$E[X] = \sum_x x P(X=x)$$

Probability mass function (PMF)
of a random value x

$$PMF(x) = P(X=x)$$

1) For a die roll expt, a fair dice,

$$E[X] = 1 \cdot \frac{1}{6} + 2 \cdot \frac{1}{6} + 3 \cdot \frac{1}{6} + \dots + 6 \cdot \frac{1}{6} = \frac{21}{6} = 3.5$$



$E[X] \leftarrow$ A fulcrum where it is the center of mass.

2) Given a random variable $X \in \{-2, -1, 0, 1, 2, 3, 4, 5\}$

$$\Rightarrow P(X=x) = \frac{1}{8} \quad \forall x \in X$$

$E[X^2]$ Suppose $Y = X^2$

$$E[Y] = \sum_y y \cdot P(y)$$

$$Y = \{0, 1, 4, 9, 16, 25\}$$

$$P_Y(y) \begin{cases} \frac{1}{8} & y=0 \\ \frac{2}{8} & y=1 \\ \frac{2}{8} & y=4 \\ \frac{1}{8} & y=9 \\ \frac{1}{8} & y=16 \\ \frac{1}{8} & y=25 \end{cases}$$

$$E[Y] = 0 \cdot \frac{1}{8} + 1 \cdot \frac{2}{8} + 4 \cdot \frac{2}{8} + 9 \cdot \frac{1}{8} + 16 \cdot \frac{1}{8} + 25 \cdot \frac{1}{8}$$

$$A4 \text{ College Book} = \frac{60}{8}$$

Kavitha

Suppose $Y = g(x)$, $E[Y] = \sum_y y \cdot p_y(y)$

Alternatively

$$E[Y] = E[X^2]$$

$$= \sum_x x^2 \cdot p_x(x)$$

$$= \frac{1}{8} \sum_x x^2 = \frac{1}{8} (4 + 1 + 0 + 1 + 4 + 9 + 16 + 25)$$

$$= \frac{60}{8}$$

i.e. $E[Y] = \sum_x g(x) p_x(x)$ useful when we know pmf of x

$$= \sum_y y p_y(y) \text{ where } Y = g(x)$$

Properties of Expectations:

1. $E[X + c] = E[X] + c$ ($c \in \mathbb{R}$ is some constant)

2. $E[Y + X] = E[X] + E[Y]$

3. $E[aX] = aE[X]$ ($a \in \mathbb{R}$ is some constant)

This shows that E is a linear operator.

Conditional Expectation:

conditioning is based on an Event. i.e.,

$$E[X|Y=y] = \sum_x x P(X=x|Y=y)$$

Eg 1:

Consider the 2 urn experiment, with additional conditions such that,

we get 10 dollars for observing a red ball & \$5 for observing a black ball.

$$S = \{1r, 1b, 2r, 2b\}$$

$$E[\text{gain}(X)] = \sum_x \text{gain}(x) \cdot P_x(x)$$

$$= 10 \cdot P(X=1r) +$$

$$= \text{gain}(1r) \cdot P(1r) + \text{gain}(1b) \cdot P(1b) + \text{gain}(2r) \cdot P(2r) + \text{gain}(2b) \cdot P(2b)$$

$$= 10 \cdot \frac{1}{3} + 5 \cdot \frac{1}{6} + 10 \cdot \frac{1}{4} + 5 \cdot \frac{1}{4} \approx \underline{\underline{7.9167}}$$

Eg 2: Suppose we pick ~~urn~~ Urn 1. What is the expected gain.
Soln:

$$E[\text{gain}(X)|Y=1] = \sum_x \text{gain}(X=x) \cdot P(X=x|Y=1)$$

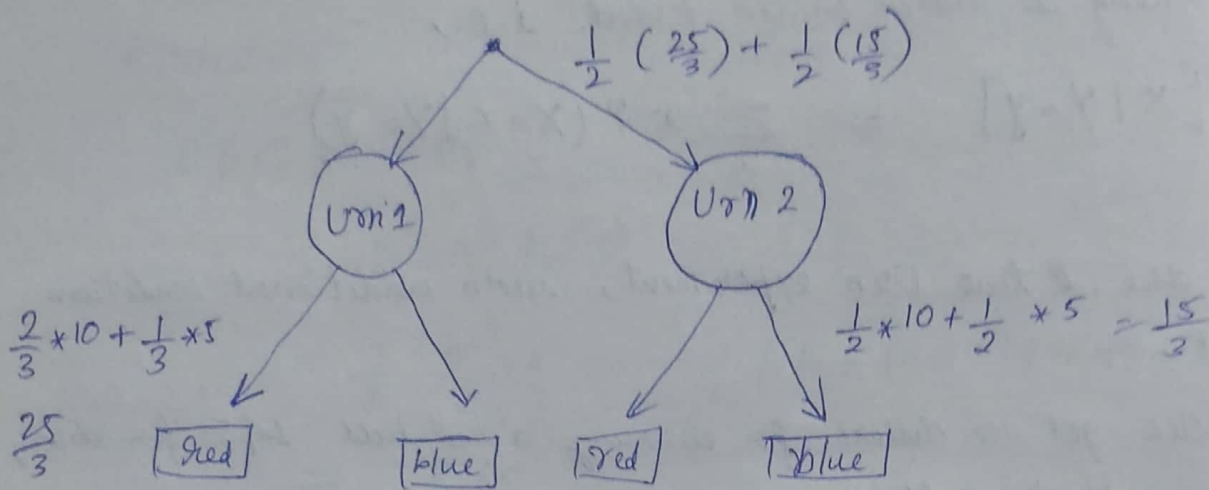
Where Y is random variable corresponding to pickup Urn

$$= \text{gain}(1r) P(X=1r|Y=1) + \text{gain}(1b) P(X=1b|Y=1) + \text{gain}(2b) P(X=2b|Y=1) + \text{gain}(2r) P(X=2r|Y=1)$$

$$= 10 \cdot \frac{2}{3} + 5 \cdot \frac{1}{3} = \underline{\underline{8.3}}$$

\therefore The gain is high if we pick Urn 1 instead of Urn 2

Law of total Expectation:



$$E[X] = \sum_y E[X|Y=y] P(Y=y) \quad \text{--- (2)}$$

$$E[X] = \sum_x x P(X=x) \quad \text{--- (1)}$$

N.K.T

$$P(X=x) = P(\{X=x\} \cap S)$$

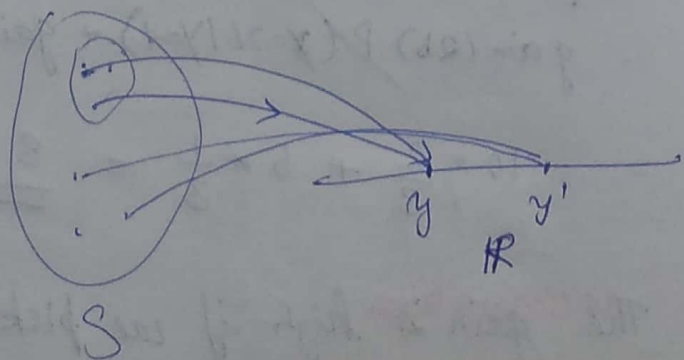
$$= P(\{X=x\} \cap \bigcup_y \{Y=y\})$$

$$= P(\bigcup_y \{X=x\} \cap \{Y=y\})$$

Events $\{X=x\}$ & $\{Y=y\}$ (are disjoint because Y is a f^n of X)

Since

$$\{Y=y\} \Rightarrow \{\omega \mid Y(\omega) = y\}$$



continuing ① & ②

$$\begin{aligned}
 &= \sum_y x \sum_x P(X=x | Y=y) P(Y=y) \\
 &= \sum_y \left[\sum_x x P(X=x | Y=y) \right] P(Y=y) \\
 &= \sum_y E[X | Y=y] \cdot P(Y=y)
 \end{aligned}$$

Markov Chain:

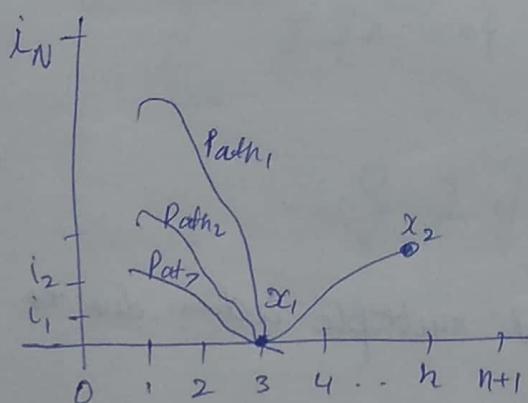
Consider a seq. $\{X_n\}$ of random variables on a common prob. (S, \mathcal{F}, P)

$$S = \{i_1, i_2, \dots, i_N\}$$

$$P(X_{n+1}=j | X_n=i_n, X_{n-1}=i_{n-1}, \dots, X_0=i_0)$$

$$= P(X_{n+1}=j | X_n=i) \quad \left(\begin{array}{l} \text{next state depends on } \text{prev state} \\ \text{irrespective of path.} \end{array} \right)$$

$$\approx P(i, j) \quad \text{transition prob.}$$



x_2 depends on x_1 only.