

Inferential Analysis

Part 2

Probability

- Probability is a branch of mathematics that deals with calculating the likelihood of a given event's occurrence.
 - Probability is the measure of the likeliness that an event will occur.
 - Probability is quantified as a number between 0 and 1 (where 0 indicates impossibility and 1 indicates certainty).
 - The higher the probability of an event, the more certain that the event will occur.
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- Probability is written as,

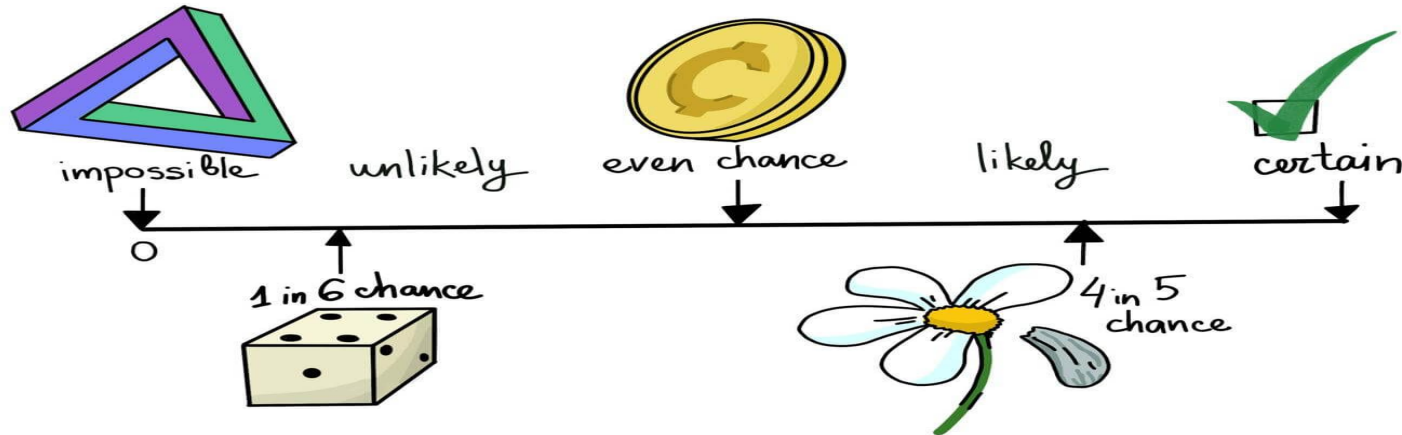
$$\text{Probabilit (Event happening)} = \frac{\text{Number of ways it can happen}}{\text{Total Number of Outcomes}}$$

Key Terms in Probability

- **Probability:** A number between 0 and 1 that indicates the chance or likelihood of an event happening.
- **Probability Experiment:** A situation where number of trials are conducted to determine probability.
- **Sample spaces:** Set of all possible of outcomes of any trial in a probability experiment.
- **Sample point:** One element of the sample spaces.
- **Outcome:** Result of the one trial in probability experiment.
- **Equally likely outcomes:** Outcomes with the same probability.
- **Expected frequency:** Number of times a specific outcomes is expected to occur when a probability experiment or a trial is repeated number of times.
- **Observed frequency:** Number of times that specific outcome did occur.
- **Event:** One or more favourable outcomes of probability experiment.
- **Complement of an Event:** All outcomes that are not the Event.

Key Terms in Probability

- **Independent Events:** Events that don't affect or are not affected by another event or events. Each event has exactly the same probability.
- **Dependent Events:** Events that affect or are affected by another event or events.
- **Mutually Exclusive Events:** Events that cannot happen at same time. The probability of mutually exclusive events is Zero.

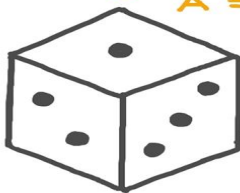


Types of Events

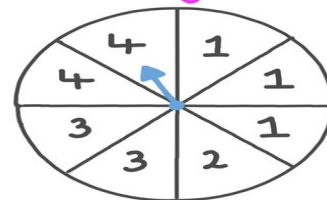
- **Dependent Event:** Dependent event also called "Conditional", where one event is affected by other events
- Example: • Drawing 2 Cards from a Deck. After taking one card from the deck there is one card less than the previous, so the probability changes!
- **Independent Event:** One event is not affected by any other events.
- Example: You toss a coin three times and it comes up "Heads" each time. what is the chance that the next toss will also be a "Head"? The chance is simply $1/2$, or 50%, just like ANY OTHER toss of the coin. What it did in the past will not affect the current toss.

DEPENDENT AND INDEPENDENT EVENTS

A = rolling an odd number
 $A = \{1, 3, 5\}$



B = spinner landing on 1



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

Conditional Probability

- A Conditional probability is a probability whose sample space has been limited to only those outcomes that fulfill a certain condition.
- A rule that can be used to determine a conditional probability from unconditional probabilities is:

$$P(A \mid B) = P(A \cap B) / P(B)$$

where:

- $P(A \mid B)$ = Conditional probability that event A will occur given that event B has occurred already
- $P(A \cap B)$ = Unconditional probability that event A and event B both occur
- $P(B)$ = Probability that event B occurs
- The usual notation for "event A occurs given that event B has occurred" is " $A \mid B$ " (A given B).
- The symbol $|$ is a vertical line and does not imply division.

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

Probability of A and B

Probability of A given B

Probability of B

Bayes theorem

- Bayes theorem shows the relationship between one conditional probability and its inverse
- It provides a mathematical rule for revising an estimate of experience and observation

Related terms in Bays theorem

P(A):The probability of event A not concerning its associated event B.

This is also called as Prior probability of A

P(B):The probability of event B not concerning its associated

This is also called as Prior probability of A

P(B|A) : Conditional probability of B given A.

This is also called as likelihood

P(A|B) : Conditional probability of A given B.

This is also called as posterior probability

The diagram illustrates Bayes' Theorem with the following components:

- LIKELIHOOD** (orange text): the probability of "B" being TRUE given that "A" is TRUE. An arrow points from this text to the $P(B|A)$ term in the numerator.
- PRIOR** (green text): the probability of "A" being TRUE. An arrow points from this text to the $P(A)$ term in the numerator.
- POSTERIOR** (green text): the probability of "A" being TRUE given that "B" is TRUE. An arrow points from this text to the $P(A|B)$ term in the denominator.
- The probability of "B" being TRUE** (pink text): An arrow points from this text to the $P(B)$ term in the denominator.

The equation is presented as:

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

Each term in the equation is enclosed in a colored box: $P(A|B)$ is in a yellow box, $P(B|A)$ is in an orange box, $P(A)$ is in a green box, and $P(B)$ is in a pink box.

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1.1 Discrete Distributions

	Notation ¹	$F_X(x)$	$f_X(x)$	$\mathbb{E}[X]$	$\mathbb{V}[X]$	$M_X(s)$
Uniform	$\text{Unif}\{a, \dots, b\}$	$\begin{cases} 0 & x < a \\ \frac{ x - a + 1}{b - a} & a \leq x \leq b \\ 1 & x > b \end{cases}$	$\frac{I(a < x < b)}{b - a + 1}$	$\frac{a + b}{2}$	$\frac{(b - a + 1)^2 - 1}{12}$	$\frac{e^{as} - e^{-(b+1)s}}{s(b - a)}$
Bernoulli	$\text{Bern}(p)$	$(1 - p)^{1-x}$	$p^x (1 - p)^{1-x}$	p	$p(1 - p)$	$1 - p + pe^s$
Binomial	$\text{Bin}(n, p)$	$I_{1-p}(n - x, x + 1)$	$\binom{n}{x} p^x (1 - p)^{n-x}$	np	$np(1 - p)$	$(1 - p + pe^s)^n$
Multinomial	$\text{Mult}(n, p)$		$\frac{n!}{x_1! \dots x_k!} p_1^{x_1} \dots p_k^{x_k} \quad \sum_{i=1}^k x_i = n$	np_i	$np_i(1 - p_i)$	$\left(\sum_{i=0}^k p_i e^{s_i} \right)^n$
Hypergeometric	$\text{Hyp}(N, m, n)$	$\approx \Phi\left(\frac{x - np}{\sqrt{np(1 - p)}}\right)$	$\frac{\binom{m}{x} \binom{m-x}{n-x}}{\binom{N}{x}}$	$\frac{nm}{N}$	$\frac{nm(N - n)(N - m)}{N^2(N - 1)}$	
Negative Binomial	$\text{NBin}(n, p)$	$I_p(r, x + 1)$	$\binom{x + r - 1}{r - 1} p^r (1 - p)^x$	$r \frac{1 - p}{p}$	$r \frac{1 - p}{p^2}$	$\left(\frac{p}{1 - (1 - p)e^s} \right)^r$
Geometric	$\text{Geo}(p)$	$1 - (1 - p)^x \quad x \in \mathbb{N}^+$	$p(1 - p)^{x-1} \quad x \in \mathbb{N}^+$	$\frac{1}{p}$	$\frac{1 - p}{p^2}$	$\frac{p}{1 - (1 - p)e^s}$
Poisson	$\text{Po}(\lambda)$	$e^{-\lambda} \sum_{i=0}^x \frac{\lambda^i}{i!}$	$\frac{\lambda^x e^{-\lambda}}{x!}$	λ	λ	$e^{\lambda(e^s - 1)}$

1.2 Continuous Distributions

	Notation	$F_X(x)$	$f_X(x)$	$\mathbb{E}[X]$	$\mathbb{V}[X]$	$M_X(s)$
Uniform	$\text{Unif}(a, b)$	$\begin{cases} 0 & x < a \\ \frac{x-a}{b-a} & a < x < b \\ 1 & x > b \end{cases}$	$\frac{I(a < x < b)}{b-a}$	$\frac{a+b}{2}$	$\frac{(b-a)^2}{12}$	$\frac{e^{sb} - e^{sa}}{s(b-a)}$
Normal	$\mathcal{N}(\mu, \sigma^2)$	$\Phi(x) = \int_{-\infty}^x \phi(t) dt$	$\phi(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left\{-\frac{(x-\mu)^2}{2\sigma^2}\right\}$	μ	σ^2	$\exp\left\{\mu s + \frac{\sigma^2 s^2}{2}\right\}$
Log-Normal	$\ln \mathcal{N}(\mu, \sigma^2)$	$\frac{1}{2} + \frac{1}{2} \operatorname{erf}\left[\frac{\ln x - \mu}{\sqrt{2}\sigma}\right]$	$\frac{1}{x\sqrt{2\pi}\sigma^2} \exp\left\{-\frac{(\ln x - \mu)^2}{2\sigma^2}\right\}$	$e^{\mu + \sigma^2/2}$	$(e^{\sigma^2} - 1)e^{2\mu + \sigma^2}$	
Multivariate Normal	$\text{MVN}(\mu, \Sigma)$		$(2\pi)^{-k/2} \Sigma ^{-1/2} e^{-\frac{1}{2}(x-\mu)^T \Sigma^{-1}(x-\mu)}$	μ	Σ	$\exp\left\{\mu^T s + \frac{1}{2}s^T \Sigma s\right\}$
Student's t	$\text{Student}(\nu)$	$I_x\left(\frac{\nu}{2}, \frac{\nu}{2}\right)$	$\frac{\Gamma(\frac{\nu+1}{2})}{\sqrt{\nu\pi}\Gamma(\frac{\nu}{2})} \left(1 + \frac{x^2}{\nu}\right)^{-(\nu+1)/2}$	0	0	
Chi-square	χ_k^2	$\frac{1}{\Gamma(k/2)} \gamma\left(\frac{k}{2}, \frac{x}{2}\right)$	$\frac{1}{2^{k/2}\Gamma(k/2)} x^{k/2} e^{-x/2}$	k	$2k$	$(1-2s)^{-k/2} s < 1/2$
F	$F(d_1, d_2)$	$I_{\frac{d_1 x}{d_1 x + d_2}}\left(\frac{d_1}{2}, \frac{d_1}{2}\right)$	$\frac{\sqrt{\frac{(d_1 x)^{d_1} d_2^{d_2}}{(d_1 x + d_2)^{d_1+d_2}}}}{xB\left(\frac{d_1}{2}, \frac{d_1}{2}\right)}$	$\frac{d_2}{d_2 - 2}$	$\frac{2d_2^2(d_1 + d_2 - 2)}{d_1(d_2 - 2)^2(d_2 - 4)}$	
Exponential	$\text{Exp}(\beta)$	$1 - e^{-x/\beta}$	$\frac{1}{\beta} e^{-x/\beta}$	β	β^2	$\frac{1}{1-\beta s} (s < 1/\beta)$
Gamma	$\text{Gamma}(\alpha, \beta)$	$\frac{\gamma(\alpha, x/\beta)}{\Gamma(\alpha)}$	$\frac{1}{\Gamma(\alpha)\beta^\alpha} x^{\alpha-1} e^{-x/\beta}$	$\alpha\beta$	$\alpha\beta^2$	$\left(\frac{1}{1-\beta s}\right)^\alpha (s < 1/\beta)$
Inverse Gamma	$\text{InvGamma}(\alpha, \beta)$	$\frac{\Gamma(\alpha, \frac{\beta}{x})}{\Gamma(\alpha)}$	$\frac{\beta^\alpha}{\Gamma(\alpha)} x^{-\alpha-1} e^{-\beta/x}$	$\frac{\beta}{\alpha-1} \alpha > 1$	$\frac{\beta^2}{(\alpha-1)^2(\alpha-2)^2} \alpha > 2$	$\frac{2(-\beta s)^{\alpha/2}}{\Gamma(\alpha)} K_\alpha(\sqrt{-4\beta s})$
Dirichlet	$\text{Dir}(\alpha)$		$\frac{\Gamma(\sum_{i=1}^k \alpha_i)}{\prod_{i=1}^k \Gamma(\alpha_i)} \prod_{i=1}^k x_i^{\alpha_i-1}$	$\frac{\alpha_i}{\sum_{i=1}^k \alpha_i}$	$\frac{\mathbb{E}[X_i](1-\mathbb{E}[X_i])}{\sum_{i=1}^k \alpha_i + 1}$	
Beta	$\text{Beta}(\alpha, \beta)$	$I_x(\alpha, \beta)$	$\frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$	$\frac{\alpha}{\alpha+\beta}$	$\frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$	$1 + \sum_{k=1}^{\infty} \left(\prod_{r=0}^{k-1} \frac{\alpha+r}{\alpha+\beta+r}\right) \frac{s^k}{k!}$
Weibull	$\text{Weibull}(\lambda, k)$	$1 - e^{-(x/\lambda)^k}$	$\frac{k}{\lambda} \left(\frac{x}{\lambda}\right)^{k-1} e^{-(x/\lambda)^k}$	$\lambda \Gamma\left(1 + \frac{1}{k}\right)$	$\lambda^2 \Gamma\left(1 + \frac{2}{k}\right) - \mu^2$	$\sum_{n=0}^{\infty} \frac{s^n \lambda^n}{n!} \Gamma\left(1 + \frac{n}{k}\right)$
Pareto	$\text{Pareto}(x_m, \alpha)$	$1 - \left(\frac{x_m}{x}\right)^\alpha x \geq x_m$	$\alpha \frac{x_m^\alpha}{x^{\alpha+1}} x \geq x_m$	$\frac{\alpha x_m}{\alpha-1} \alpha > 1$	$\frac{x_m^\alpha}{(\alpha-1)^2(\alpha-2)} \alpha > 2$	$\alpha(-x_m s)^\alpha \Gamma(-\alpha, -x_m s) s < 0$