

Probability Distribution

Distribution : The possible values a variable can take and How frequently they occur.

- probabilities measure the likelihood of an outcome, depending on how often it is featured in the sample space.

Two characteristics:

- mean (average) μ
- Variance (spread out) σ^2

data kinds

- Population data "all the data"
- Sample data "part of it"

- sample mean = \bar{X}

- sample variance = S^2

- sample STD = S

- population STD = σ

القيمة المتوقعة

$$\sigma^2 = E((Y - \mu)^2) \rightarrow E(Y^2) - \mu^2$$

Types of probability Distribution

1. Discrete distribution

- finite number of outcomes (Die, picking a card)

2. Continuous distribution

- infinitely many outcomes (time, Distance)

variable
 $X \sim N(\mu, \sigma^2)$ characteristics (May vary)
 Type

Discrete distributions

- uniform Distribution
 - all outcomes are equally likely.
 - ↳ Equiprobable.
 - (picking a card, coin)
- Bernoulli Distribution
 - Events with only two possible outcomes
 - ↳ true
 - ↳ false
- Binomial Distribution
 - * Two outcomes per iteration
 - * Many iterations.
 - (Flipping a coin 3 times)
- Poisson Distribution
 - Test out how unusual an event frequency is for a given interval

Continuous distributions

- Normal distribution
 - often observed in nature
- Chi-Squared
 - Asymmetric.
 - only consists of non-negative value, always starts from zero on the left
- Doesn't often mirror real life events
- The curve is skewed to left
- used in Hypothesis Testing to help to determine how Goodness of fit
- Exponential distribution
 - Events are rapidly changing early on.
- logistic distribution
 - useful in Forecast analysis
 - useful for determining a cut-off point for a successful outcome

* Uniform distribution $U()$

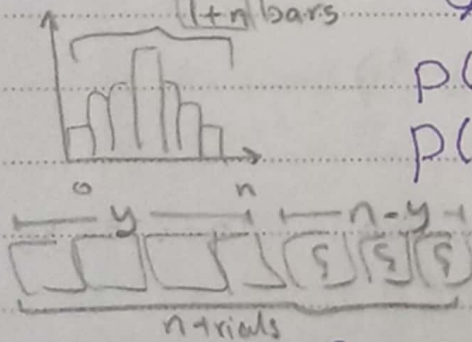
- Each outcome is equally likely.
- Expected value Provides us no relevant information because all outcomes have the same probability
- Both the mean and the variance are uninterpretable.
- No predictive power

* Binomial distribution $B()$

- It is a sequence of identical Bernoulli events

$B(n, p)$ ← عدد المحاولات
احتمال نجاح
كل مرة

$$\text{Bern}(p) = B(1, p)$$



$$p(\text{desired outcome}) = p$$

$$p(\text{alternative outcome}) = 1-p$$

- The number of ways in which 4 out of the 6 trials

$$\text{can be successful} = C_4^6$$

$$p(y) = \binom{n}{y} \cdot p^y \cdot (1-p)^{n-y}$$

Expected value

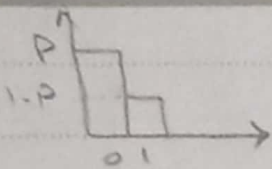
$$E(x) = x_0 \cdot p(x_0) + x_1 \cdot p(x_1) + \dots + x_n \cdot p(x_n)$$

$$\sigma^2 = E(y^2) - E(y)^2 = n \cdot p \cdot (1-p)$$

* Bernoulli distribution $\text{Bern}(p)$

events with :
 → 1 Trial
 → 2 possible outcomes

p is known OR past data indicating some experimental probability



$$1 \rightarrow p$$

$$E(x) \rightarrow p$$

$$0 \rightarrow 1-p$$

$$\sigma^2 = p(1-p)$$

* Poisson distribution $P_0(\lambda)$

Deals with the frequency with which an event occurs

$(p)^x$

$(\lambda)^x$

عدد مرات حدوث الشيء خلال فترة زمنية أو مساحة محددة

$$P(Y) = \frac{\lambda^y e^{-\lambda}}{y!}$$

$$E(y) = y_0 \cdot P(y_0) + y_1 \cdot P(y_1) + \dots$$

$$\mu = \sigma^2 = \lambda$$

* Normal distribution $N(\mu, \sigma^2)$

Appearing in nature.

ex: The size of a full grown male lion. 150 : 250 kg

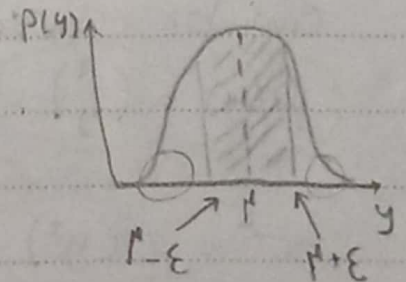
Distinct characteristics:

The most of the data is centered around the mean

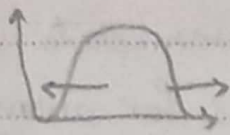
$$E(x) = \mu$$

$$\text{Var}(x) = \sigma^2 \leftarrow \text{usually given}$$

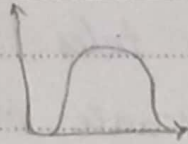
$$\text{Var}(x) = E(x^2) - E(x)^2$$



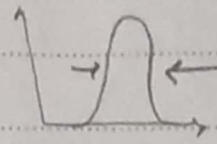
transformation: A way in which we can alter every element of a distribution to get a new distribution with similar characteristics.



$$y = F\left(\frac{x}{\sigma}\right)$$



$$y = F(x)$$



$$y = F(x, c)$$

Standardizing: A special kind of transformation in which $E(X) = 0$, $Var(X) = 1$ we get Standard Normal distribution.

Z-score table: is statistical measure that describes a value's relationship to the mean of a group of values.

How to standardize?

$$y = F(x) \xrightarrow{\mu=0} y = F(x - \mu) \xrightarrow{\sigma=1} y = F\left(\frac{x - \mu}{\sigma}\right)$$

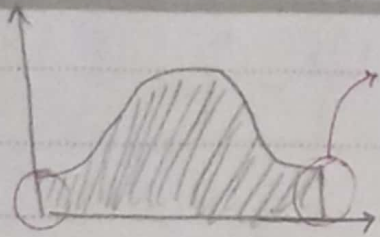
$$\left. \begin{array}{l} Z \sim N(0, 1) \\ y \sim N(\mu, \sigma^2) \end{array} \right\} \Rightarrow Z = \frac{y - \mu}{\sigma} \leftarrow \begin{array}{l} \text{Transformation} \\ \text{we use in} \\ \text{Standardizing} \end{array}$$

$$y: \mu + 2.3\sigma \Rightarrow Z = 2.3$$

Student's T distribution $t(k)$ degrees of freedom

- small sample size approximation of a normal distribution

- Certain characteristics + sufficient data = Normal distribution
- Certain characteristics + sufficient data = Student's T distribution



fatter tails to accommodate the occurrence of values far away from the mean

if $k > 2$

$$- E(Y) = \mu$$

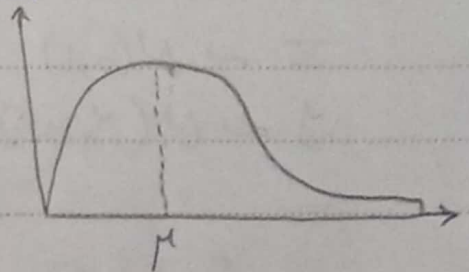
$$- \text{Var}(Y) = \frac{s^2 * k}{k-2}$$

Application : - Frequently used when conducting statistical analysis.

- Hypothesis testing with limited data.
- CDF table (T-table)

Chi-Squared distribution $\chi^2(k)$

- statistical analysis when doing :
 - Hypothesis testing
 - Computing confidence intervals



$$\bullet \sqrt{x} \sim t(k)$$

$$\bullet Y \sim t(k) \rightarrow Y^2 \sim \chi^2(k)$$

$$E(X) = k$$

$$V(X) = 2k$$

Conditional probability

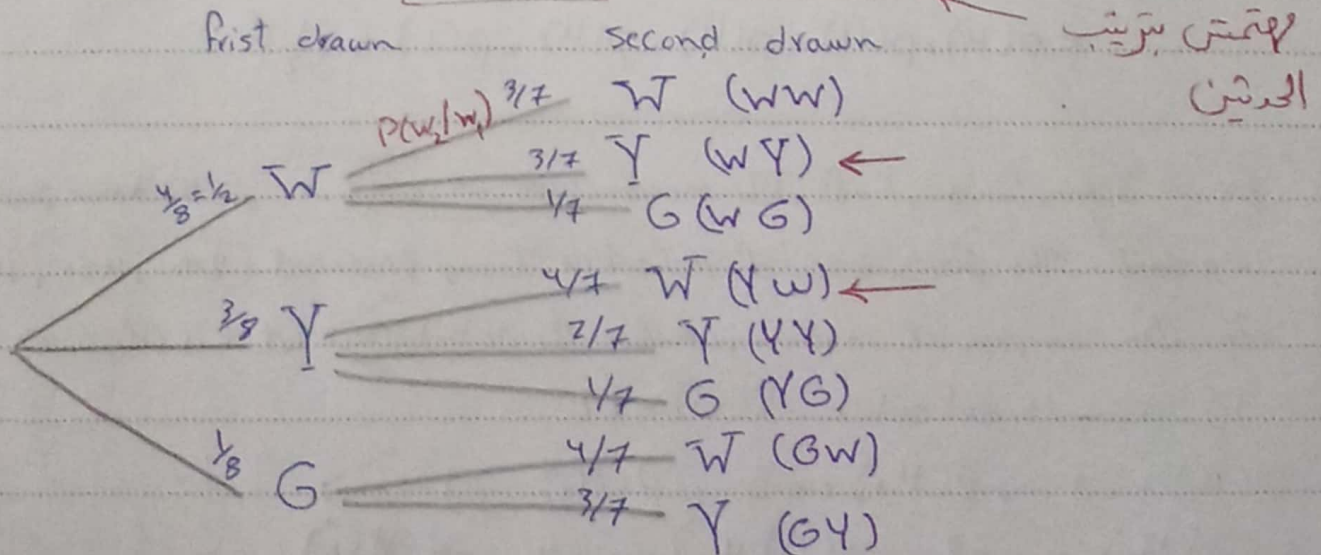
$$E.F.: S = \{1, 2, 3, 4, 5, 6\}, A = \{1, 3, 5\}, B = \{1, 2\}$$

$$P(B/A) = \frac{1}{3}$$

$$* P(E|F) = \frac{P(E \cap F)}{P(F)}$$

$$\text{Product rule} \rightarrow P(E \cap F) = P(F) \cdot P(E|F)$$

EX 1. From a box containing four white, three yellow, and one green ball, two balls are drawn one at a time without replacing the first ball before the second is drawn. Use a tree diagram to find the probability that one white and one yellow ball are drawn.



Sol:

$$\begin{aligned}
 P(W \cap Y) &= P(Y_{1st}) \cdot P(W_{2nd} | Y_{1st}) + P(W_{1st}) \cdot P(Y_{2nd} | W_{1st}) \\
 &= \frac{3}{8} \cdot \frac{4}{7} + \frac{1}{2} \cdot \frac{3}{7} = \frac{3}{7}
 \end{aligned}$$

Independent Events

$$P(E|P) = P(E)$$

$$P(E \cap F) = P(E) \cdot P(F) \rightarrow \text{Criterion for independent Events}$$

$$P(E_1 \cap E_2 \cap \dots \cap E_n) = P(E_1) \cdot P(E_2) \cdot \dots \cdot P(E_n)$$

Ex: - A fair coin is tossed repeatedly until the first tail appear. what is the probability of getting the first tail at the 5th trial?

$$H: \text{Head}, P(H) = \frac{1}{2}$$

$$T: \text{Tail}, P(T) = \frac{1}{2}$$

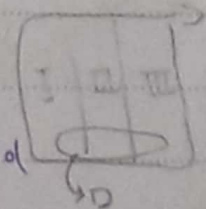
$$\begin{aligned} P(\text{1st tail appears in 5th trial}) &= P(H_1 \cap H_2 \cap H_3 \cap H_4 \cap T_5) \\ &= P(H) \cdot P(H) \cdot P(H) \cdot P(H) \cdot P(T) = \left(\frac{1}{2}\right)^5 \end{aligned}$$

Ex: - 3 machines, I, II, III manufacture $\left(\frac{1}{10}, \frac{2}{10}, \frac{7}{10}\right)$ of the total production in a plant. The percentage of defective items produced $\left(\frac{1}{10}, \frac{2}{10}, \frac{3}{10}\right)$ For item chosen at random, what is the probability that it is defective?

$$D: \text{item is defective} \quad P(D) = ??$$

$$P(M_1) = 0.4, P(M_2) = 0.5, P(M_3) = 0.1$$

$$P(D|M_1) = 0.02, P(D|M_2) = 0.04, P(D|M_3) = 0.01$$



$$\begin{aligned} P(D) &= P(D \cap M_1) + P(D \cap M_2) + P(D \cap M_3) \\ &= 0.4 \cdot 0.02 + 0.5 \cdot 0.04 + 0.1 \cdot 0.01 \end{aligned}$$

Bayes' Theorem

$$P(A_j | E) = \frac{P(A_j) \cdot P(E | A_j)}{P(E)}$$

الحدث
الذي حصل

$$= \frac{P(A_j) \cdot P(E | A_j)}{P(A_1) \cdot P(E | A_1) + P(A_2) \cdot P(E | A_2) \dots}$$