

Hypothesis Testing

① Null Hypothesis

- Well accepted / default
- Represented by H_0

$H_0 \rightarrow T_1 \& T_2$ (Same)

$$\mu_1 = \mu_2$$

② Alternate Hypothesis

- Opposite to H_0
- Represented by H_A

$H_A \rightarrow T_1 \& T_2$ (not same)

$$\mu_1 \neq \mu_2$$

②

Fig. 1

Eg:-

Consider a pharma company is working in India & it wants to launch a new drug in India, we want check whether it will succeed or not.

Population:- want to survey on

→ How many no. of people using that drug or not

Steps

① Population:- group of people want to survey

Sample :- parts of Population

Hypothesis [Claim that says it's true]

- ① Null (H_0)
- ② Alternative (H_A)



Popn. Sample

①
②

Now, Same company launch New Drug & they want to find whether this new drug will work or not, so we will do hypothesis

③

① Null Hypothesis

Consider old drug 20% used in population we are assuming New one will also work with 20% used then it says that Null Hypothesis

9

is True, As $H_0 = H_1$

Otherwise its Alternating Hypothesis
Only one Hypothesis will remain at a time

$$H_0 \neq H_1$$

Test to determine Null Hypothesis is correct or Not

1) χ^2 test (Chi)

2) t-Student test

3) Fisher Z test

Level of Significance (α) [5% 1%]

Confidence (C) [95% 99%]

$$\alpha + C = 1$$

A random sample of 50 items gives the mean 6.2 and variance 10.24. Can it be regarded as drawn from normal population with mean 5.4 at 5% level of Significance?

$$\bar{X} = 6.2, \quad S^2 = 10.24, \quad n = 50$$

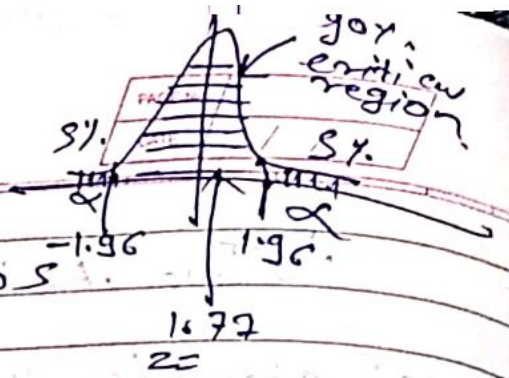
$$\text{Null Hypothesis } H_0: \mu = 5.4$$

$$\text{Alternative Hypothesis } H_1: \mu \neq 5.4$$

Test Statistic

$$Z = \left| \frac{\bar{X} - \mu}{\frac{S}{\sqrt{n}}} \right| = \left| \frac{6.2 - 5.4}{\frac{\sqrt{10.24}}{\sqrt{50}}} \right| = 1.77$$

$$Z = 1.77$$



(4) Level of Significance, $\alpha = 0.05$

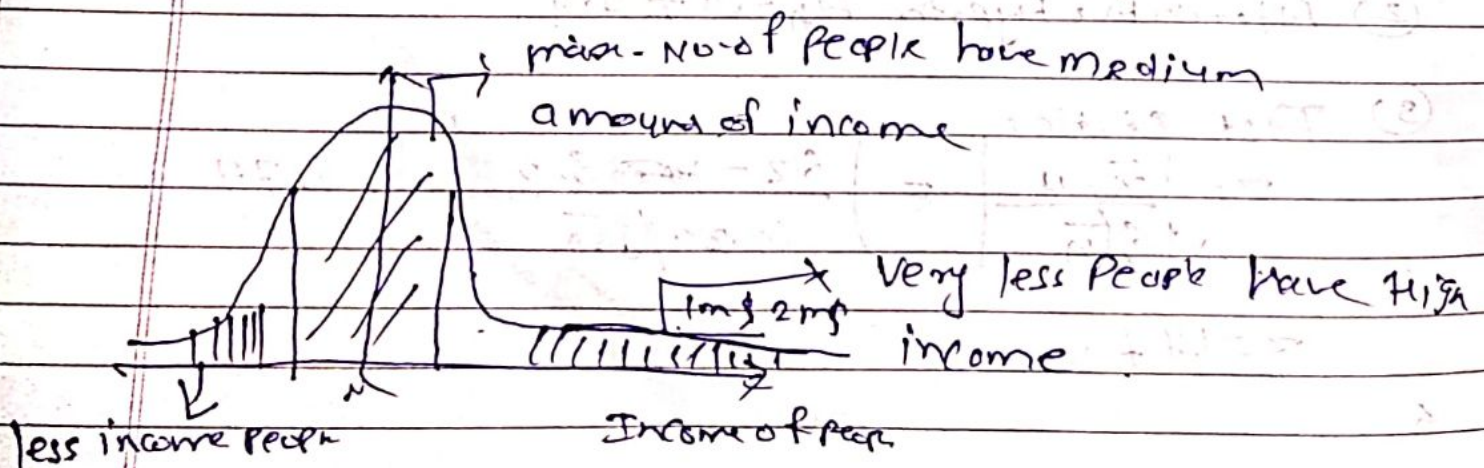
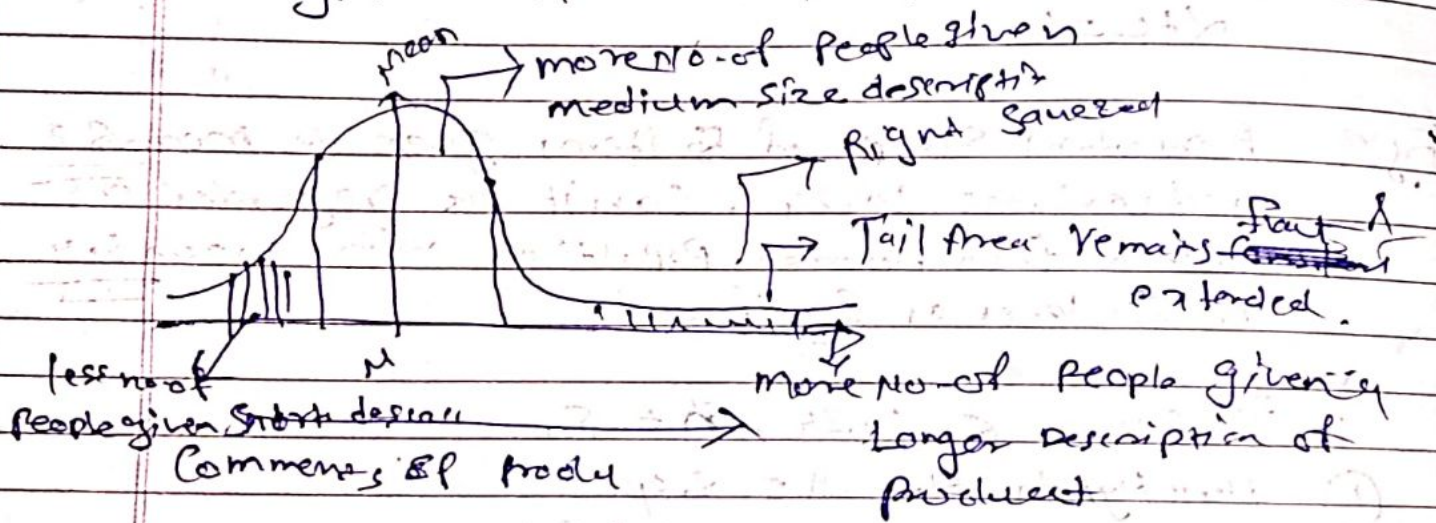
(5) Critical value:
if $\alpha = 0.05$ then $z = 1.96$

(6) Decision

$$1.77 < 1.96$$

So Null Hypothesis is accepted.

Log Normal Distribution



Std. Normal Distribⁿ $\mu = 0, \sigma = 1$

PAGE NO.	
DATE	/ /

→ Random variable

X in Log Normal distribution

if $\log(x)$ is normally distributed $\rightarrow [\ln(x) \sim N(\mu, \sigma)]$

$$X = \{x_1, x_2, x_3, \dots, x_n\}$$

$$\log(x) = \{\ln(x_1), \ln(x_2), \ln(x_3) \dots \ln(x_n)\}$$

Why we learn Distribution.

E.g.:-

Marketing Expense \rightarrow convert to Log Normal \rightarrow question data

$$15,000 \rightarrow \ln(15000) \sim N(\mu, \sigma)$$

$$16,000 \rightarrow \ln(16000) \sim N(\mu, \sigma)$$

$$17,000 \rightarrow \ln(17000) \sim N(\mu, \sigma)$$

$$18,000 \rightarrow \ln(18000) \sim N(\mu, \sigma)$$

$$\rightarrow \text{S.D.} \left[\frac{\sigma}{\sqrt{n}} \right]$$

Strat. Normal

Distrib.

All values will be scatter down

So due to this accuracy of model increase

Whole process / known as Log Normalisation
Model Accuracy will be increased.

$$\text{Variance} = \frac{1}{n} \sum_{i=1}^n (x_i - \mu)^2 \rightarrow \text{mean}$$

You can't Build a Time Series model, until you know
Time Series Stationary.

PAGE NO.	
DATE	/ /

$$y = 140, 140, 140, 90, 140$$

1 2 3

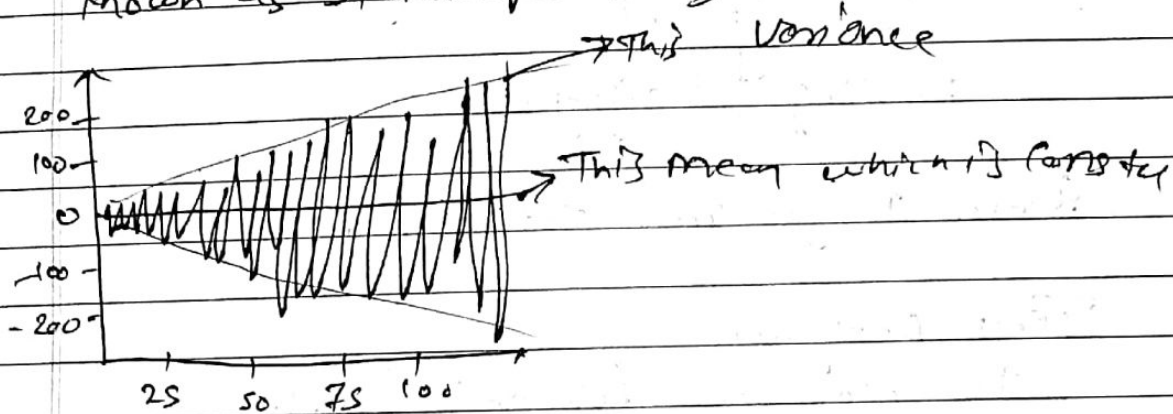
Stationary: Probability of Stat values repeat over the Time

Joint probability of a Series doesn't change over time - mean & Variance remain constant over time.

No trend in the Series

$$F(x_t) = F(y_{t+k})$$

F is a joint probability distribution known as Strict probability.



Mean Stationary Process

- Constant mean
- Constant variance
- Constant Auto Covariance \rightarrow It's Covariance between two ^{diff} timepoints between same Series

$$y_t \rightarrow y_{t+k} \text{ same}$$

\rightarrow Covariance between same variable with lags.

$$\text{Auto}(t_1, t_2) = \text{Auto}(t_3, t_4) = \text{Auto}(t_5, t_6)$$