

Assumptions of LP -

- 1) Linearity
- 2) Normality of residuals
- 3) Homoscedasticity
- 4) No auto correlation
- 5) NO or little multicollinearity

1) Linearity -

↳ There is linear relationship b/w dependent & independent var

→ what if assumption fails -

- Bias in parameter estimates
- Reduce predictive accuracy
- Invalid hypothesis test & confidence interval

• How to check assumptions -

1) Scatter plot

↳ If 2D or 3D then we can see

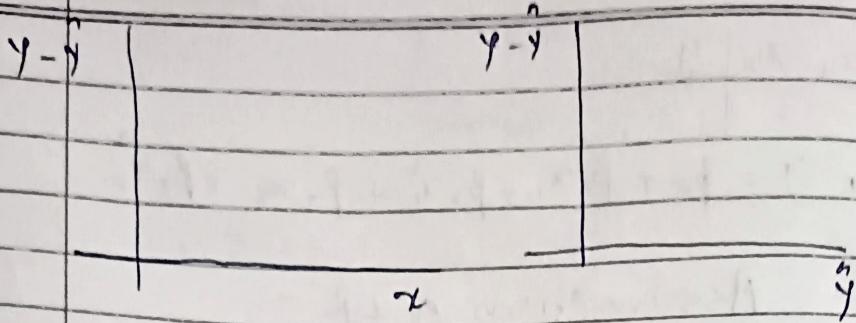
but if 4D or more $x_1, x_2, x_3 | y$

then make scatter plot individually

$$x_1 \rightarrow y, x_2 \rightarrow y, x_3 \rightarrow y$$

2) Residual plot

↳ Residuals $y - \hat{y}$



↳ If linearity holds then there will be random scatter and if there is any kind of pattern then linearity fails.

3) Polynomial Representation -

- ↳ apply linear Reg and then check R^2 score.
- ↳ apply polynomial if there is significant improvement means non-linearity

⇒ What if assumption fails -

i) Apply transformation

↳ on dep. & indep. variables

ii) Polynomial Reg -

↳ apply on indep. variable & check R^2 score.

2) Normality of Residuals -

↳ The error terms (residuals) are assumed to follow normal distribution with mean zero & constant variance.

$$e_i \sim N(0, \sigma^2)$$

• what if assumption fails

i) Inaccurate hypothesis test -

↳ f-tests & t-tests assume that residuals follow normal dist

→ If it doesn't follow then hypothesis tests may be inaccurate

- 2) Invalid confidence interval
- 3) Bad model performance

• How to check assumptions -

- 1) Plot histogram or kde plot for residual
 $y, \hat{y} \Rightarrow y - \hat{y}$
 ↳ nistsp or kde

- 2) Draw Q-Q plot

- 3) Statistical test

↳ Shapiro-wilk, omnibus,
 Jarque-Bera.

→ Omnibus test-

- 1) H_0 : residuals are normally dist.
 H_A : " " " not "

- 2) fit linear reg model

- 3) calculate Residuals ($y - \hat{y}$)

- 4) calculate skewness

- 5) " " " kurtosis

- 6) $\chi^2 = n \left[\frac{(\text{skewness})^2}{6} + \frac{(\text{kurtosis})^2}{24} \right]$

$n \rightarrow$ no. of observations

χ^2 follows χ^2 dist. So plot it in graph
 $(df=2)$

→ compare with p-value & decide whether reject or accept hypothesis

→ what to do when assumption fails -

↳ feature Selection

↳ Robust regression

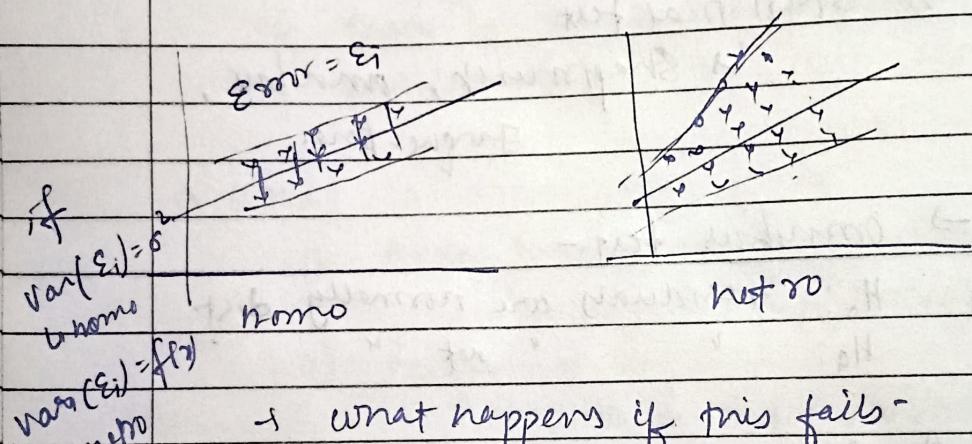
↳ Use Non-parametric methods

↳ Use bootstrapping

{ No need to worry if sample size > 30 }

g) Homoscedasticity :-

↳ The spread of error terms should be constant across all level of indep. variable.



→ what happens if this fails -

- SE, b_0, b_1 will be unreliable

- t-test, f-test fails

- Invalid confidence intervals.

Standard error → Standard dev of the

Sample means \bar{x}

$$SE = \frac{\sigma}{\sqrt{n}}$$

This line can be anywhere
w/ no two lines