

Forward selection.

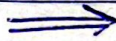
Backward elimination

Subset selection algorithms.

Stepwise regression

* Forward selection.

Starts with **no** predictors.



Adding predictors
one by one.

Which predictor add next?

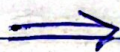
One that increases R^2_{adj} the **most.**

When/Where to stop?

When **R^2_{adj} stops increasing.**

* Backward elimination.

Starts with **all** predictors.



Removing predictors
one by one.

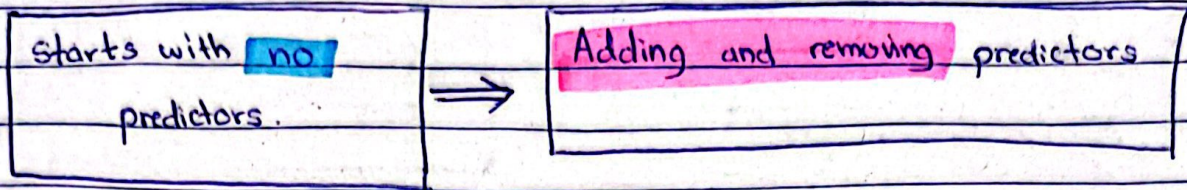
Which predictor remove next?

One that increases R^2_{adj} the **most.**

When/Where to stop?

When **R^2_{adj} stops increasing.**

* Stepwise regression.



Which predictor add/remove next?
Based on AIC / R^2_{adj} .

When/Where to stop?
When dropped R^2_{adj} increases.

- ❌ - Problems:-

- 1) Backward elimination methods are not feasible when $p > n$.
Number of variables/features p is greater than Number of observations n .
- 2) Different methods select different models.
- 3) There is no guarantee to provide optimal solution where the model with the highest R^2_{adj} .

* Accuracy measure

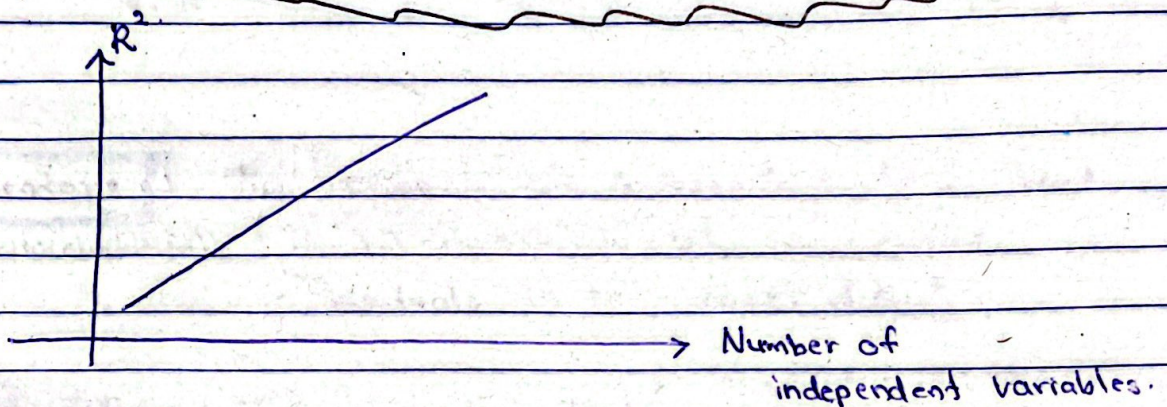
$$* R^2 = 1 - \frac{RSS}{TSS}$$

R^2 = coefficient of determination.

RSS = Residuals sum of squares.

TSS = Total sum of squares.

Interpretation: The percentage of the response variable variation that is explained by a linear model.



$$* R^2_{adj} = 1 - \left[\frac{(n-1)(1-R^2)}{(n-p-1)} \right]$$

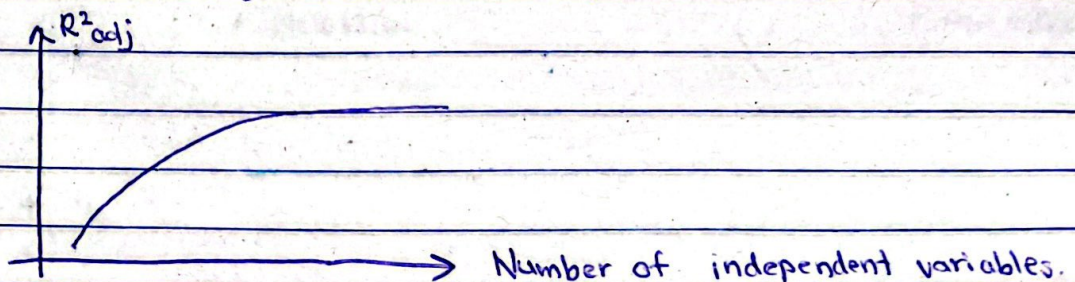
n = Number of observations
 p = Number of variables.

Difference

⊙ R^2_{adj} : adjusted for the number of predictors.

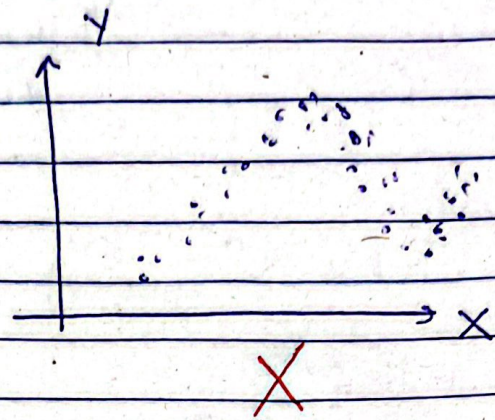
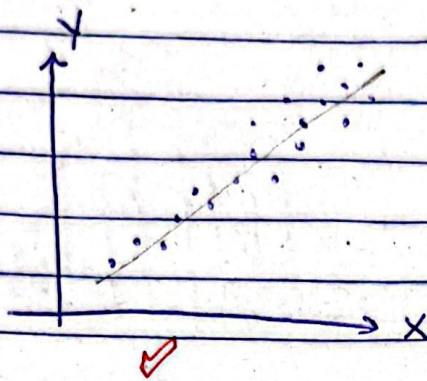
⊙ R^2_{adj} increases only if the new term improves the model.

⊙ Always less than R^2 .



* Assumptions of linear regression

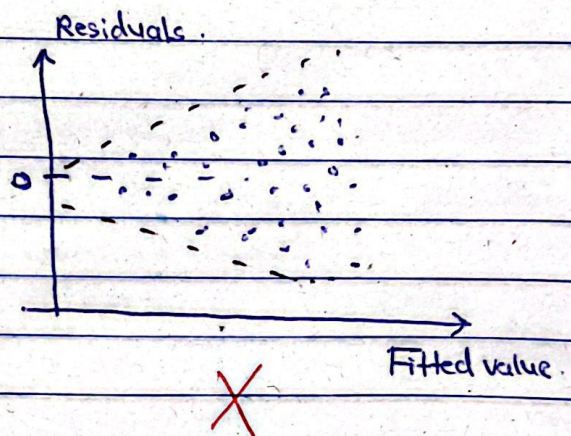
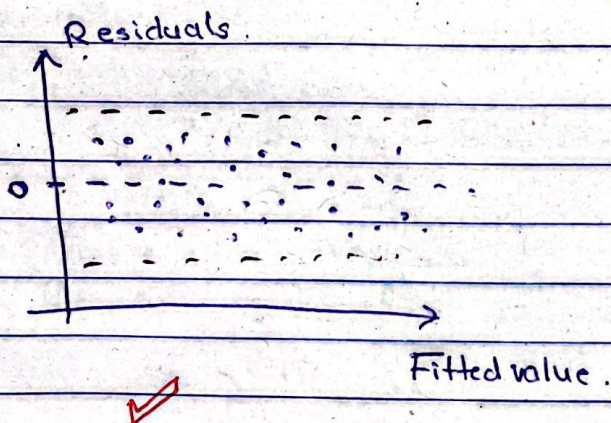
- ① **Linear relationship**: There exists a linear relationship between the independent variable, X , and the dependent variable, Y .



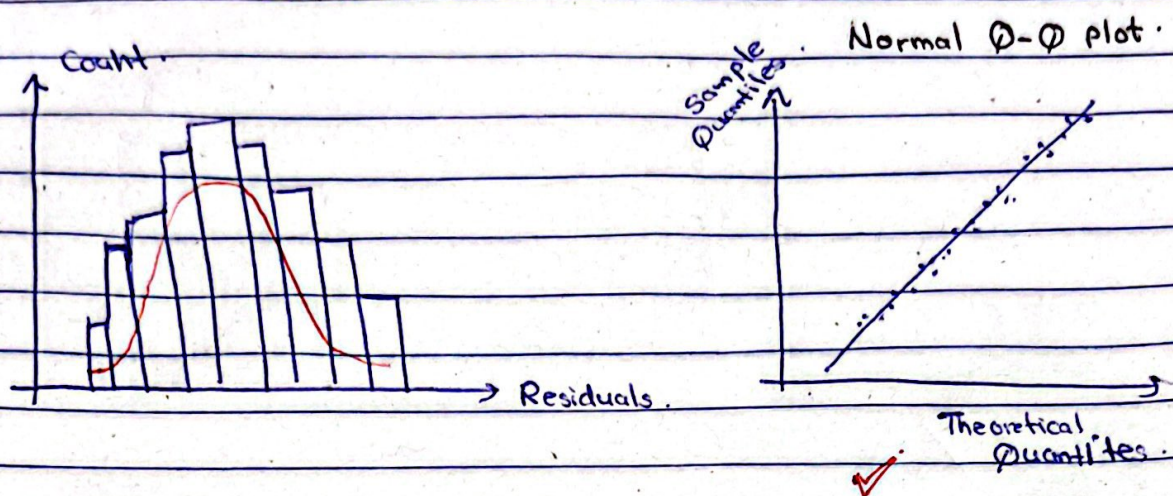
- ② **Independence**: The residuals are independent. (i.e. there (No Autocorrelation) is no correlation b/w consecutive residuals in time series data.)

⇒ Check: Durbin-Watson test. ← Not required in this tutorial.

- ③ **Homoscedasticity**: The residuals have constant variance at every level of X .



- ④ **Normality**: The residuals of the model are normally distributed.



- ⑤ **No multicollinearity**: There is no relationship between predictor variables.

⇒ Check pairwise plots.

