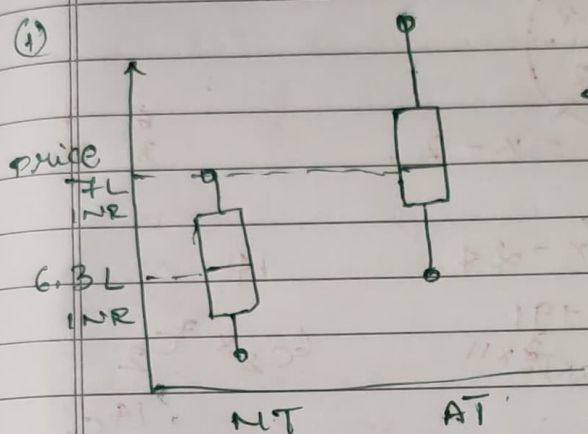


Questions

(1)



$$\hat{\text{Price}} = \beta_0 + \beta_1 (\text{NT} | \text{AT})$$

where β_0 & β_1
(No usage of computers)

$$= 0.7 +$$

(30-35%)
scale

soln:- $6.3 + 0.7 \times$

→ Encode $\text{NT} = 0$ & $\text{AT} = 1$

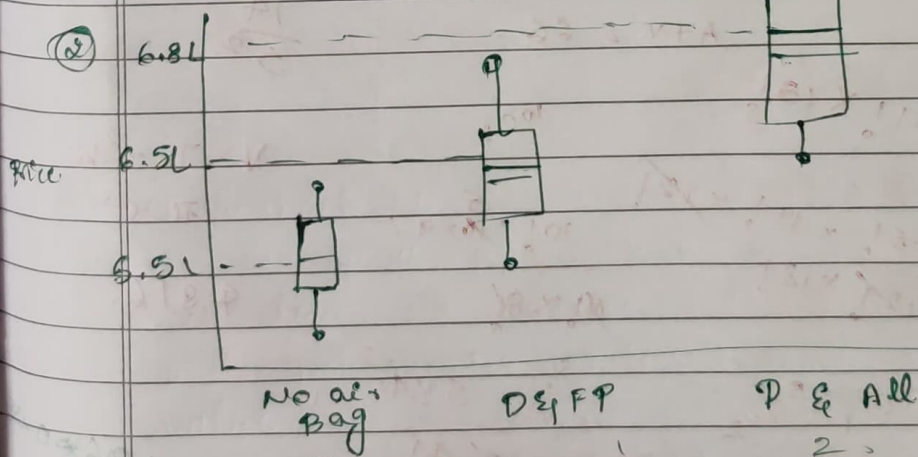
When I need to predict price of NT,
($\text{NT} = 0$) I need to get 6.3. i.e. $\beta_0 = 6.3$.

→ If Encoded as $\text{NT} = 1$ & $\text{AT} = 0$

$$= 6.3 + 0.7 \times 1$$

→ Slope captures the difference in the mean.

(2)



$$\beta_0 + \beta_1 (\text{D&FP}) + \beta_2 (\text{D&FP})$$

$$5.5 + (1 \text{ D&FP} + 1.3 \text{ DA scale})$$

- For ordinal category, there is no consistency in change / difference
- Hence, we don't do ordinal encoding for linear regression

Nominal } categorical
ordinal }

Interval Ratio } Phase

→ NO air bags
→ D F P.

$$\rightarrow D F \Phi.$$

C.S DFP

Now Drop DFP

$$\hat{\beta} = \beta_0 + \beta_1 d_1 + \beta_2 d_2$$

$$= 6.5 - 1d_1 + 0.3d_2$$

New Prop. DALL

$$\hat{q} = 6.8 - 1.3d_1 - 0.3d_2$$

③ Covariance & Correlation coefficient Understanding.
Soln: Unit gets disrupted in Covariance.

Correlation of Coefficient = $\frac{\text{Cov}(x, y)}{\sigma_x \sigma_y}$ Takes $\text{Cov}(x, y)$ in Z score (x, y)

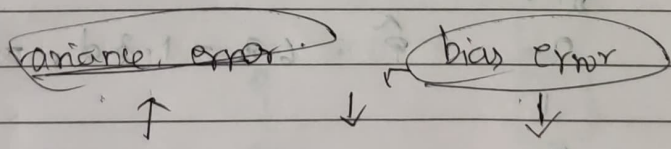
$$\delta_{xy} = \text{cov}(\text{standardized}(x, y))$$

Calculating using formula - Analytical / Parametric Approach

- 4) Strength & Weakness of Linear Regressor Model
- Strength \rightarrow It is very simple. - decoded easily.
 - Weakness \rightarrow Accuracy is low - slightly for benefit of inference high bias error

can reduce variance error at the cost of high bias error - Ridge, Lasso, EN.

			overfit \leftarrow High variance errors
P_1	P_2	P_3	\rightarrow No consistency but low error (RMSE)
89	10	100	High bias error \rightarrow Consistency but low high error (RMSE) underfit
85	18	60	
86	12	12	
83	17	20	
82	13	62	
$\mu = 85$	$\mu = 16$	$\mu = 64$	



- 5) Drawback of Train-Test-split validation.
- Soln: It gives a number & point estimation is done. We cannot tell if new data will be greater or lesser than this estimated value.

Hence, we go with K fold cross validation.

- 6) What is the R^2 of Null model regressor?
- Null model \rightarrow Model having no feature.
- ~~R^2 of null model~~ $\hat{y} = \bar{y}$ - mean of target actual
- R^2 of null model is 0

For Null Model:-

$$\bar{y} = \beta_0 + \text{for Null, } x \geq 0.$$

$$\beta_0 = \bar{y} - b, \bar{x}$$

$$\beta_0 = \bar{y}$$

$$\hat{y} = \beta_0 + \beta_1 x_1 + \beta_2 x_2$$

$$\hat{y} = \beta_0 = \bar{y}$$

7) What is the AUC score of Null model classifier
Soln:- 0.5.

8) Use of Base line score / why do we talk about null model (Base line ref)

9) What do you mean by deviance = 0.
→ Indicates perfectly fit data

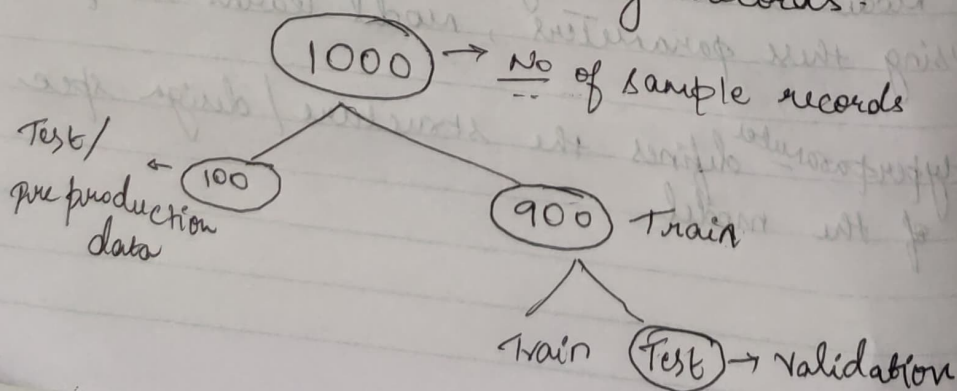
10) classical probability - Sample space is fixed

Empirical probability - Having historical probability

Subjective probability - No historical probability
Need domain expert

11) If something is revealed in conditional probability, the sample space reduced.

12) Training, Validation & Testing records.



- 12) Impact when n neighbour is increased.
Soln:- Bias increases. - Model behaves like a Null model

Eg:- 1's \rightarrow 23

0's \rightarrow 18.

If I give $n=41$, it will consider all the points 41 as its neighbours and picks the one which has maximum or more no of records.

- If n neighbour is decreased - It leads to overfit (high variance error)

- 13) KNN cannot be boosted but can be bagged.
Boost to reduce bias error.
No luxury of reducing bias error later.
Try to design with least bias error.

Cannot boost as there are no coeff kind of thing.

- 14) Difference b/w hyperparameter & model parameter

All models have model parameter except KNN model understanding data in its own way using these parameters, model learns the data.

Hyperparameter defines the structure / design spec of the model

15) Neural n/w without hidden layer acts as
(logistic) Linear model.