

1).

	Coefficient	Std. error	t-statistic	p-value
Intercept	2.939	0.3119	9.42	< 0.0001
TV	0.046	0.0014	32.81	< 0.0001
radio	0.189	0.0086	21.89	< 0.0001
newspaper	-0.001	0.0059	-0.18	0.8599

→ large

TABLE 3.4. For the **Advertising** data, least squares coefficient estimates of the multiple linear regression of number of units sold on TV, radio, and newspaper advertising budgets.

NULL HYPOTHESIS:-

↓
There is no R^n

Simple Regression $\begin{cases} H_0: \beta_1 = 0 \\ H_a: \beta_1 \neq 0 \end{cases}$

Multiple Regression $\begin{cases} H_0: \beta_1 = \beta_2 = \dots = \beta_p = 0 \\ H_a: \text{at least one } \beta_j \text{ is non zero} \end{cases}$

A: Null Hypo:

-) TV has no R^n wrt Sales
-) Rad has no R^n wrt sales
-) News has no R^n wrt sales

A: conclusion:

-) As the p value for TV & radio is very less, we can say that there is R^n between these parameters and Sales.
-) P (Newspaper) ...

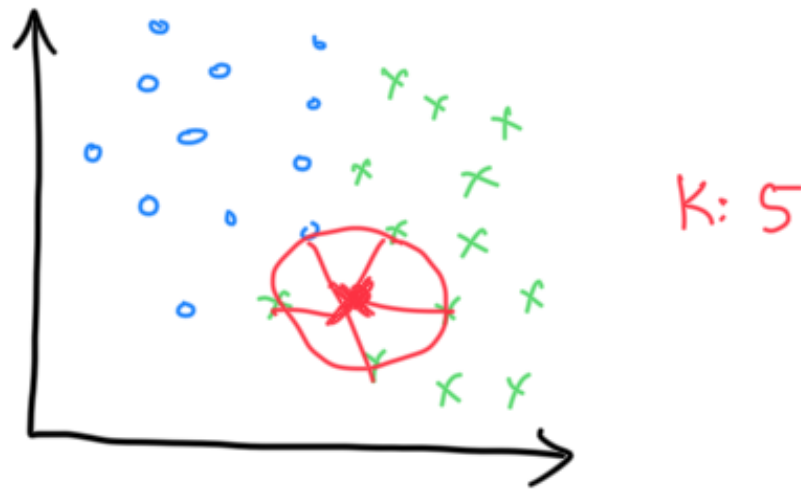
... is right \therefore do not reject H_0

2). KNN classifier:

1. get the data
2. Init the K
3. Find nearest K point to our target point

for a class j and point i :

$$P(Y_j | X_i) = \frac{1}{K} \sum_{i: N_0} \mathbb{I}(y_i = j)$$



$$\therefore P(Y_{\text{blue}} | X) = \frac{1}{5} (1) //$$

$$P(Y_{\text{green}} | X) = \frac{1}{5} (4) //$$

KNN regression:-

Same:

$$\hat{f}(x_0) = \frac{1}{K} \sum_{x_i \in N_0} y_i$$

Find dist btw points

mean - L1 loss

median - L2 loss

3)- $X_1 = \text{GPA}$ $X_2 = \text{IQ}$ $X_3 = \begin{cases} \text{college} : 1 \\ \text{School} : 0 \end{cases}$ $X_4 = X_1 \oplus X_2$ $X_5 = X_1 \oplus X_3$

$\beta_0 : 50$ $\beta_1 : 20$ $\beta_2 : 0.07$ $\beta_3 : 35$ $\beta_4 : 0.01$ $\beta_5 : -10$

Y (starting salary) in 1000 \$

A:

$$\text{Salary}_{\text{coll}} : \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3 \dots \rightarrow \textcircled{1}$$

$$= 50 + 20 X_1 + 0.07 X_2 + 35 X_3 + 0.01 X_4 + (-10 X_5)$$

$$\text{Salary}_{\text{HS}} : 50 + 20 X_1 + 0.07 X_2 + 0 + 0.01 X_4 + 0$$

$$\text{Salary}_{\text{coll}} = \text{Salary}_{\text{HS}}$$

$$\cancel{50} + \cancel{20 X_1} + \cancel{0.07 X_2} + 35 X_3 + \cancel{0.01 X_4} - 10 X_5 = \cancel{50} + \cancel{20 X_1} + \cancel{0.07 X_2} + \cancel{0.01 X_4}$$

$$35 X_3 - 10 X_5 = 0$$

$$35 X_3 > 10 X_5$$

$$35 \cancel{X_3} > 10 X_1 \cancel{X_3}$$

$$\frac{35}{10} < X_1$$

$$\text{GPA} > 3.5 //$$

Only (iii) is correct.

(b) IQ = 110 GPA = 4.0

Plug these into (1)

$$\underline{\underline{\$137100.0}}$$

(c) The coeff tells that how much effect it has on unit increase.
Every unit \uparrow of β_4 , $Y \uparrow 10 \$$

(1) Do S-E to check how good our estimate is

S-E $\lll \rightarrow$ reliable params

$\beta_p \lll \& S-E \lll =$ affects \hat{y}

$\beta_p \lll \& S-E >>> =$ not affecting \hat{y}

(4) $n = 100$

Linear: $Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \epsilon$

Cubic: $Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2^2 \dots + \epsilon$

(a) X & Y is linear

$RSS_{\text{train (lin)}}$

$RSS_{\text{train (cubi)}}$

what would be it?

From my learning:

RSS_{cubi} will be less than linear because

$$RSS = e_1^2 + e_2^2 + e_3^2$$

Bias: error of Train data

Var: error of test data

While training linear would have High B & High Var.
training cubi would have low bias & High Var.

$$RSS_{\text{train linear}} > RSS_{\text{train cubi}}$$

(b) For testing:

cubi - overfit & linear - underfit

$$RSS(\text{linear}) < RSS(\text{cubi})$$

It explains more variance

(c) RSS training \downarrow as the model is of high variance
RSS cubic \ll New model

(d) For test we cannot say because if the new model is little more poly - the variance \uparrow it