

2. (a). Data Description:

The data are on the homicide rate in Detroit for the years 1961-1973.

FTP - Full-time police per 100,000 population

UEMP - % unemployed in the population.

MAN - number of manufacturing workers in thousands.

LIC - number of handgun licenses per 10,000 population.

GR - number of handgun registrations per 100,000 population.

~~CLAR - % homicides cleared~~

MAN - number of non-manufacturing workers in thousands

GOV - number of government workers in thousands.

HE - Average hourly earnings.

WE - Average weekly earnings

HOM - number of homicides per 100,000 of population.

From the problem, we know that we use three of the variables can predict HOM.

Using the basic linear regression function model: $y(x, w) = w_0 + w_1x_1 + \dots + w_nx_n$. $X = (x_1, \dots, x_n)^T$

s.t. $y(x, w) = w_0 + w_1x_1 + w_2x_2 + w_3x_3$. $x_1 = (FTP(1), FTP(2), \dots, FTP(13))^T$

$$= WX$$

$$x_2 = (WE(1), WE(2), \dots, WE(13))^T$$

If we want to find the best third variable; and we need to find x_3 .

We can try all the other variables and find the minimize sum-of-squares error = 0

$$E(w) = \frac{1}{2} \sum_{n=1}^N \{t_n - w^T \phi(x_n)\}^2 \quad w = (\Phi^T \Phi)^{-1} \Phi^T t \quad \Phi \text{ is } X \quad t \text{ is } (HOM(1), HOM(2), \dots, HOM(13))^T$$

then for different third variable we will get different w .

And we can use RMSE, by finding the minimize RMSE. we can find the right third variable:

$$RMSE = \sqrt{\frac{\sum_{i=1}^N (HOM(i) - y_i(x, w))^2}{N}} = \sqrt{\frac{\sum_{i=1}^N (y_i(x, w) - HOM(i))^2}{n}}$$

I wrote 2 piece of code. one is just use the original data, the other is normalized data by ^{divide} its ~~max value~~ upper bound in column. s.t. every value will $\in [0, 1]$.

Both of them show that LIC is the best third variable to choose.

without normalization: $y = -58.1244 + 0.1847 FTP + 0.1068 WE + 0.0165 LIC$. \rightarrow result in Linear R.m

with normalization: $y = -0.5812 + 0.7388 FTP + 0.3205 WE + 0.1482 LIC$. \rightarrow result in Linear R.m

2(b). i: using b to replace ~~figure~~ feature / s '?'

for real-value just replace missing values with the label-conditioned mean.

for letter-value just replace missing values with mode value.

st. replace all '?' in feature2, 3.15 with mean others with ~~mod~~ value