

Statistics and Data Analysis Formula sheet

Descriptive: One variable

Mean: $\bar{x} = \frac{1}{N} \sum_{U} x_i$

Mode: Most frequently occurring value.

Variance: $S_x'^2 = \frac{1}{N} \sum_U (x_i - \bar{x})^2$ $(x_i -$ $S_x^2 = \frac{1}{N} \sum_U (x_i - \bar{x})^2$ $S_x^2 = \frac{1}{N-1} \sum_U (x_i - \bar{x})^2$ $S_y^2 = \frac{1}{N-1} \sum_U (x_i - \bar{x})^2$ $S_y^2 = b S_x$ $S_y^2 = b^2 S_x^2$

If x is a dummy variable with mean $\bar{x} =$ P_x , then $S_x'^2 = P_x(1-P_x)$ and $S_x^2 = \frac{N}{N-1}P_x(1-P_x)$ P_x).

$$S_x' = \sqrt{S_x'^2}$$
 and $S_x = \sqrt{S_{x,U}^2}$

Result: Let a and b be constants, x be a variable and $y_i = b(x+a)$ then

$$\bar{y} = b(\bar{x} + a) \qquad \qquad \dot{y} = b(\dot{x} + a)$$

$$\dot{y} = b(\dot{x} + a)$$

$$\breve{y}_p = b(\breve{x}_p + a)$$

$$Sk_y = Sk_x$$

$$S_{n}^{2} = b^{2} S_{n}^{2}$$

Standardization:

Let $z = \frac{x - \bar{x}}{S_x}$ then $\bar{z} = 0$ and $S_z = 0$.

Let $x_{(1)}, x_{(2)}, \dots, x_{(N)}$ be the ordered pop- (<o): 为体系统

ulation.

a be the integer part of c and b be the decimal part of c. The pth percentile is $\check{x}_p =$ $(1-b)x_{(a)} + bx_{(a+1)}$.

First quartile: $\breve{x}_{0.25}$

Median (second quartile): $\breve{x}_{0.50}$

Third quartile: $\breve{x}_{0.75}$

Range: range_r = $x_{(N)} - x_{(1)}$

Interquartile range: $IQR_x = \ddot{x}_{0.75} - \ddot{x}_{0.25}$

$$r_{xy,U}^{s} \equiv \frac{\sum_{U}(R(x_i) - \bar{R}_U)(R(y_i) - \bar{R}_U)}{\left(\sum_{U}(R(x_i) - \bar{R}_U)^2 \sum_{U}(R(y_i) - \bar{R}_U)^2\right)^{1/2}} \in [-1, 1]$$

with $\bar{R}_U = (N+1)/2$ and R(x) and R(y)

the ranks of x and y

Kendall's correlation: - 扱知+1. 不- 扱ー1

 $r_{xy,U}^k \equiv \frac{2}{n(n-1)} \sum_{i < j} \operatorname{sgn}(x_j - x_i) \operatorname{sgn}(y_j - y_i)$ $\bar{R}^2 = 1 - \frac{SSE/(n-K-1)}{SST/(n-1)}$.

Simple Linear regression

Fitted line: $\hat{y} = b_0 + b_1 x$ with

$$b_1 = r_{xy} \frac{S_y}{S_x} = \frac{\sum_U (x_i - \bar{x})(y_i - \bar{y})}{\sum_U (x_i - \bar{x})^2}$$

$$b_0 = \bar{y} - b_1 \bar{x}.$$

Fitted values: $\hat{y}_i = b_0 + b_1 x_i$.

Residuals: $e_i = y_i - \hat{y}_i$.

Sum of squares: SST = SSE + SSR

$$SST = \sum_{s} (y_i - \bar{y}_s)^2$$
 数据以干方和、固定

$$SSE = \sum_{s} (y_i - \hat{y}_i)^2$$
 误充平方式 越小越好

$$SSR = \sum_s (\hat{y}_i - \bar{y}_s)^2$$
 內耳方如,越大越好。

Coefficient of determination:

$$R^2 = \frac{SSR}{SST} = 1 - \frac{SSE}{SST}.$$

Prediction for y given x_0 : $\hat{y}_0 = b_0 + b_1 x_0$

Multiple Linear regression

Fitted line: $\hat{y} = b_0 + b_1 x_1 + \cdots + b_K x_K$.

Fitted values: $\hat{y}_i = b_0 + b_1 x_{1i} + \cdots + b_K x_{Ki}$.

Residuals: $e_i = y_i - \hat{y}_i$.

Sum of squares: Same as above.

Coefficient of determination: Same.

Adjusted coef. of determination:

$$\bar{R}^2 = 1 - \frac{SSE/(n-K-1)}{SST/(n-1)}.$$