1 Statistische Maßzahlen

2 Wahrscheinlichkeit

empirische Kovarianz

$$\tilde{s}_{xy} = \frac{1}{n-1} \sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})$$

es gilt:
$$P(A \cap B) = P(A) + P(B) - P(A \cup B)$$

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

$$P(A \cap \overline{B}) = P(A) - P(A \cap B)$$

$$P(\overline{A} \cap \overline{B}) = P(\overline{A} \cup \overline{B}) = 1 - P(A \cup B)$$

empirische Varianz:

$$\sigma^2 = \tilde{s}^2 = \tilde{s}_x^2 := \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 = \frac{1}{n-1} (\sum_{i=1}^n x_i^2 - n\bar{x}^2)$$

Standardabweichung

$$\sigma = \tilde{s} = \tilde{s}_x = \sqrt{\tilde{s}^2}$$

empirischer Korrelationskoeffizient (nach Bravais Pearson)

$$r_{xy} = \frac{\tilde{s}_{xy}}{\tilde{s}_x \cdot \tilde{s}_y}$$

Spearman'sche Rangkorrelationskoeffizient von X und Y

$$r_{SP}(x,y) = \frac{\sum_{i=1}^{n} \left(rg(x_i) - \overline{rg_x} \right) \left(rg(y_i) - \overline{rg_y} \right)}{\sqrt{\sum_{i=1}^{n} \left(rg(x_i) - r\overline{g}_x \right)^2 \cdot \sum_{i=1}^{n} \left(rg(y_i) - r\overline{g}_y \right)^2}}$$

$$\tag{1}$$

1.1 Lineare Regression

optimale Ausgleichgerade:

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

$$\hat{a} = \frac{\tilde{s}_{xy}}{\tilde{s}_x^2}$$

$$\hat{b} = \bar{y} - \hat{a}\bar{x}$$

Bestimmtheitsmaß, Modellgüte:

$$R^2 = r_{xy}^2$$