

Tæthedsfunktioner

Gyldighed

$$\int_{-\infty}^{\infty} f_X(x) dx = 1$$

CDF til PDF

$$f_X(x) = \frac{d}{dx} F_x(x)$$

PMF til CDF

$$F_X(x_i) = \sum_{x=-1}^{x_i} f_X(x)$$

PDF til CDF

$$F_X(x_i) = \int_{-\infty}^{x_i} f_X(u) du$$

Marginale tæthedsfunktioner

$$f_X(x) = P(X=x) = \sum_y P(X=x, Y=y) = \sum_y f_{XY}(x, y)$$

$$f_Y(y) = P(Y=y) = \sum_x P(X=x, Y=y) = \sum_x f_{XY}(x, y)$$

Sandsynlighed (pdf)

$$Pr(X > x) = 1 - F_X(x)$$

$$N = N_{1k} + N_{10k} + N_{100k}$$

Betinget sandsynlighed

$$f_{X|Y}(x|y) = \frac{f_{XY}(x, y)}{f_Y(y)} = Pr(X=x | Y=y)$$

Middelværdi

Forventningsværdi (pdf)

$$E[X] = \bar{X} = \mu_X = \int_{-\infty}^{\infty} x_i \cdot f_X(x) dx$$

Forventningsværdi (pmf)

$$E[X] = \sum_{i=1}^n x_i \cdot E(X_i) = \sum_{i=1}^n x_i \cdot f_X(x=x_i)$$

Uniform fordeling

$$E[X] = \frac{a+b}{2}$$

Ensemble middelværdi

$$\mu_X(t) = E[X(t)] = \int_{-\infty}^{\infty} x(t) \cdot f_X(x(t)) dx(t) \quad \mu_X(n) = E[X(n)] = \int_{-\infty}^{\infty} x(n) \cdot f_X(x(n)) dx(n)$$

Tidslig middelværdi

$$\mu_{xi} = (X_i)_T = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} x_i(t) dt \quad \mu_X(t) = E[X(t)] = E[a \cdot t]$$

Varians

Varians (pdf)

$$var(x) = \sigma_x^2 = E[x^2] - E[x]^2 = \int_{-\infty}^{\infty} x^2 \cdot f_X(x) dx - E[x]^2$$

Varians (pmf)

$$E[X^2] = \sum_{i=1}^n x_i^2 \cdot E(X_i) = \sum_{i=1}^n x_i^2 \cdot f_X(x=x_i)$$

Normal fordeling

$$\mu_X(t) = E[X(t)] = E[a \cdot t]$$

$$\sigma_x^2 = E[x^2] - E[x]^2$$

$$E[X_n] = E[2 \cdot w(n) - 1] = 2 \cdot E[w(n)] - 1$$

Uniform fordeling

$$var(w(n)) = \frac{(b-a)^2}{12}$$

$$var(2 \cdot w(n) - 1) = 2^2 \frac{(b-a+1)^2 - 1}{12}$$

$$E[w(n)] = \frac{a+b}{2}$$

$$var(X(t)) = var(a \cdot t) = t^2 var(a)$$

Ensemble varians

$$E[3 \cdot x^2] = 3 \cdot E[X^2]$$

$$var(X(t)) = \sigma_x^2(t) = E[X(t) - \mu_X(t)^2]$$

$$var(X(n)) = \sigma_x^2(n) = E[X(n) - \mu_X(n)^2]$$

Varians af en konstant

$$Var(k) = E[k^2] - E[k]^2 = k^2 - k^2 = 0$$

Standardafvigelse

Standardafvigelse

$$\sigma_x = \sqrt{E[x^2] - E[x]^2}$$

$$\sigma_x = \sqrt{\sigma_x^2}$$

Hypoteser

Estimator (middelværdi)

$$\delta = \bar{x}_1 - \bar{x}_2 \sim N\left(0, \frac{\sigma^2}{n_1} + \frac{\sigma^2}{n_2}\right)$$

Estimator (varians)

$$s^2 = \frac{1}{n_1 + n_2 - 2} ((n_1 - 1) s_1^2 + (n_2 - 1) s_2^2) \quad \sigma = \sqrt{s^2}$$

Test size

$$t = \frac{\delta}{s \cdot \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

P value

$$p_{val} = 2 \cdot (1 - t_{cdf}(|t|, n_1 + n_2 - 2))$$

95% konfidens interval

$$\delta_{lower} = (x_1 - x_2) - t_0 \cdot s \cdot \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$$

$$\delta_{upper} = (x_1 - x_2) + t_0 \cdot s \cdot \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$$

Korrelation

$$\rho = E \left[\frac{X - \bar{X}}{\sigma_X} \cdot \frac{Y - \bar{Y}}{\sigma_Y} \right] = \frac{E[XY] - E[X]E[Y]}{\sigma_X \cdot \sigma_Y} = \frac{\text{cov}(X, Y)}{\sigma_X \cdot \sigma_Y}$$

$$\sigma_Y = \sqrt{E[Y^2] - E[Y]^2}$$

Autokorrelation

$$R_{XX}(t_1, t_2) = E[X(t_1) \overline{X(t_2)}]$$

$$X(n) := \underline{w}(n)$$

$$R_{XX}(\tau) := E(X(n) \cdot X(n-\tau)) \rightarrow E(w(n) \cdot w(n-\tau))$$

$$\tau := 0, 1..3 \quad n := 1, 2..10$$

$$R_{XX}(\tau) \rightarrow \begin{bmatrix} E(w(n)^2) \\ E(w(n) \cdot w(n-1)) \\ E(w(n) \cdot w(n-2)) \\ E(w(n) \cdot w(n-3)) \end{bmatrix}$$

Sandsynlighed

Prior

$$Pr(A) = 0.001$$

$$Pr(\bar{A}) = 1 - Pr(A)$$

$$Pr(B)$$

Likelihood

$$Pr(B|A) = 0.92$$

$$Pr(B|A') = 1 - Pr(Pr(B|A))$$

$$Pr(B'|A) = 0.98$$

$$Pr(B'|A') = 1 - Pr(B'|A)$$

Posterior

$$Pr(B'|A) = \frac{Pr(A|B') \cdot Pr(B')}{Pr(A)}$$

Total Probabilty (Sum Rule)

$$Pr(B) = Pr(B \cap A) + Pr(B \cap \bar{A}) = Pr(B|A) \cdot Pr(A) + Pr(B|\bar{A}) \cdot Pr(\bar{A})$$

$$Pr(A) = Pr(A \cap B) + Pr(A \cap \bar{B}) = Pr(A|B) \cdot Pr(B) + Pr(A|\bar{B}) \cdot Pr(\bar{B})$$

Total Probabilty (Bayes Rule)

$$Pr(A|C) = \frac{Pr(C|A) Pr(A)}{Pr(C)}$$

Joint Events

$$Pr(B \cap A) = Pr(B|A) Pr(A)$$

Relative Frequency Approach

$$N_A = 28131$$

$$N_B = 29785$$

$$N = N_A + N_B = 57916$$

$$28131 + 29785 = 57916$$

$$Pr(A) = \frac{N_A}{N}$$

Binomial

$$Pr_n(k) = \frac{n!}{k!(n-k)!} p^k q^{n-k} = \binom{n}{k} p^k q^{n-k}$$

Ensemble

Ensemble mean

$$\mu_x(x) = E[X(n)] = \frac{a+b}{2}$$

Ensemble variance

$$var(x) = E[X(n)^2] - E[X(n)]^2 = \frac{(b-a)^2}{12}$$

Statistik

Known variance

$$\mu_{est} = \bar{x} = \frac{1}{n} \sum_{i=1}^n X_i$$

$$z = \frac{\bar{x} - \mu_0}{\sigma \cdot \sqrt{n}}$$

$$p = 2 \cdot |1 - \Phi(|z|)|$$

$$\mu_{lower} = \bar{x} - 1.96 \cdot \frac{\sigma}{\sqrt{n}}$$

$$\mu_{upper} = \bar{x} + 1.96 \cdot \frac{\sigma}{\sqrt{n}}$$

$$s = \sqrt{s^2}$$

$$z = \frac{(\bar{x}_1 - \bar{x}_2)}{\sigma \cdot \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

Unknown variance

$$\mu_{est} = \bar{x} = \frac{1}{n} \sum_{i=1}^n X_i$$

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

$$t = \frac{\bar{x} - \mu_0}{s \cdot \sqrt{n}}$$

$$p = 2 \cdot |1 - t_{cdf}(|t|)|$$

$$t_0 = tinv\left(\frac{1-0.05}{2}, n-1\right)$$

$$\mu_{lower} = \bar{x} - t_0 \cdot \frac{s}{\sqrt{n}}$$

$$\mu_{upper} = \bar{x} + t_0 \cdot \frac{s}{\sqrt{n}}$$

Poisson

$$H_0 \quad \lambda = \lambda_0$$

$$H_1 \quad \lambda \neq \lambda_0$$

$$\lambda = \frac{x}{t}$$

$$z = \frac{x - t \cdot \lambda}{\sqrt{t \cdot \lambda}}$$

$$z = \frac{X - E[X]}{\sqrt{var(X)}} = \frac{X - t \cdot \lambda}{\sqrt{t \cdot \lambda}} \quad (\text{normal approximation})$$

$$p = 2 \cdot |1 - \Phi(|z|)|$$

$$\lambda_{lower} = \frac{1}{t} \left[x + \frac{1.96^2}{2} - 1.96 \cdot \sqrt{x + \frac{1.96^2}{4}} \right]$$

$$\lambda_{upper} = \frac{1}{t} \left[x + \frac{1.96^2}{2} + 1.96 \cdot \sqrt{x + \frac{1.96^2}{4}} \right]$$

Binomial

$$p_{est} = \frac{x}{n}$$

$$z = \frac{x - n \cdot p_0}{\sqrt{n \cdot p_0 (1 - p_0)}}$$

$$p = 2 \cdot |1 - \Phi(|z|)|$$

$$p_{lower} = \frac{1}{n + 1.96^2} \left[x + \frac{1.96^2}{2} - 1.96 \cdot \sqrt{\frac{x(n-x)}{n} + \frac{1.96^2}{4}} \right]$$

$$p_{upper} = \frac{1}{n + 1.96^2} \left[x + \frac{1.96^2}{2} + 1.96 \cdot \sqrt{\frac{x(n-x)}{n} + \frac{1.96^2}{4}} \right]$$

Regression

$$\alpha = \bar{y} - \beta \bar{x} \quad (\text{skæring})$$

$$\beta = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2} \quad (\text{hældning})$$

$$\varepsilon_i = y_i - (\alpha + \beta x_i) \quad (\text{residual})$$

$$s_r^2 = \frac{1}{n-2} \sum_{i=1}^n (y_i - (\alpha + \beta x_i))^2 = \frac{1}{n-2} \sum_{i=1}^n \varepsilon_i^2$$

$$\delta_{est} = \bar{x}_1 - \bar{x}_2$$

$$\delta_{lower} = (\bar{x}_1 - \bar{x}_2) - 1.96 \cdot \sigma \cdot \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$$

Konfidensinterval for hældningen β

$$\beta_{lower} = \beta - t_o \cdot \sqrt{\frac{s_r^2}{\sum_{i=1}^n (t_i - \bar{t})^2}}$$

$$\beta_{upper} = \beta + t_o \cdot \sqrt{\frac{s_r^2}{\sum_{i=1}^n (t_i - \bar{t})^2}}$$