

Exam CompStat: 26.7.17 16:00

July 18, 2017

- y_i, \hat{y} RV
- $\langle y \rangle = \mu_y$ not random \rightarrow Prior

0.1 Basic Properties of random Variables

- Transitional, cond.prob., joint prob. margin

$$\text{int}(p(x, y), y) = f(x)$$

- law of error propagation

$$\begin{aligned} \text{sig}_y^2 &= \text{sig}_x^2 * (dy/dx)^2 \\ P(x|y) &= P(x, y)/P(y) \end{aligned}$$

0.2 Distributions

- uniform, binomial, poissonian, (multivariate-) gaussian, χ^2 , exponential

$$\begin{aligned} f(x) &> 0 \\ \int f(x)dx &= 1 \end{aligned}$$

- PDF

$$f(x)dx$$

- CDF

$$\text{int}(f(x), x, -\text{Inf}, y) = g(y)$$

- MGF

$$\langle e^{tx} \rangle = \int e^{tx} f(x)dx$$

- moments

$$\begin{aligned} \int x^m f(x)dx \\ \int (x - \alpha)^m f(x)dx \end{aligned}$$

0.3 Bayes Theorem

- $\langle \rangle$
 $P(d, t)$
 $G(x, \mu) = \exp(-.5*(x-\mu)^2/\text{sig}^2)$
 $P(D|T) = P(D,T)/P(T)$
 $P(T|D) = P(D,T)/P(D)$
- frequentist
 $P(D,T) = \dots$
 $d_i, d_{\text{hat}} = \text{sum}(d_i, i)/N = f \text{ of data}$
 $\langle d_{\text{hat}} \rangle = \mu$
 $\rightarrow \text{PDF}(\theta_{\text{hat}})$ (PDF of estimator)
- Bayes
 $P(T|D) = P(D|T) * P(T)/P(D)$
 $E(D) = \int L(D|T) * \pi(T), T$
 $B_{\text{A}}(AB) = E_{\text{A}}(D)/E_{\text{B}}(D)$

0.4 Fisher Estimation

$$F_{\text{L}}(\theta, \theta_0) = (\text{dell}^2 \log(L))/(\text{dell}_{\theta_0} * \text{dell}_{\theta_0}) @ \text{maxL} = (..)$$

- we assume gaussian distr. params. around the peak for a found maximum likelihood estimator
 $\exp(-.5(x_{i-} - \mu_i) C_{(i,j)}^{-1} (x_{j-} - \mu_j))$

0.5 fitting

$$t = f(x) = A_0 + A_1 * f_1(x) + A_2 * f_2(x)$$

$$\exp(-.5 * \text{sum}((d_i - t_i)^2 / \sigma_i^2))$$

$$\rightarrow C_{(i,j)}$$

$$A |_{\text{maxL}} = G^{-1} D, F_{\text{L}}(\alpha, \beta) = G$$