

Data Science Hw 1

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Problem 1 Random Number Transformation

$$(1) F_w(x) = P[W \leq x] = P[\max(X, Y) \leq x] = P[X \leq x]P[Y \leq x] = F_X(x)F_Y(x)$$

$$f_w(x) = dF_w(x)/dx = d(F_X(x)F_Y(x))/dx = f_X(x)F_Y(x) + f_Y(x)F_X(x) \\ = 2 * e^{-x} * (1 - e^{-x})$$

$$(2) P[\min(X, Y) < x] = P[\{X < x\} \text{ or } \{Y < x\}] = P[X < x] + P[Y < x] - P[\{X < x\} \text{ and } \{Y < x\}] \\ = F_X(x) + F_Y(x) - F_X(x)F_Y(x)$$

$$f_w(x) = f_X(x) + f_Y(x) - 2 * e^{-x} * (1 - e^{-x}) = 2e^{-2x}$$

Problem 2 Statistical Distances

$$(1) f(x, y) \geq 0 \text{ (non-negativity)}$$

$$f(x, y) = 0 \text{ if and only if } x = y \text{ (identity of indiscernibles)}$$

$$f(x, y) = d(y, x) \text{ (symmetry)}$$

$$f(x, z) \leq d(x, y) + d(y, z) \text{ (subadditivity / triangle inequality)}$$

$$(2) d^2(x, y) = \|x - y\|^2 = \langle x - y, x - y \rangle = \|x\|^2 - \|y\|^2 - \langle 2y, x - y \rangle$$

$$\text{The Bregman distance is defined as: } D_F(p, q) = F(p) - F(q) - \langle \nabla F(q), p - q \rangle_{f(x)}$$

$$\text{Let } F(x) = \|x\|^2$$

$$d^2(x, y) = F(x) - F(y) - \langle \nabla f(y), x - y \rangle$$

$$(3) H(x, y) = H(x|y) + H(y) = H(y|x) + H(x)$$

$$y = f(x), \text{ so } f(y|x) = 0$$

$$H(x, y) = H(y) + H(x|y) = H(x)$$

$$H(y) \leq H(x)$$

Problem 3 Point Estimation

$$(1) E(X) = \int_0^\theta x \frac{1}{\theta} dx = (1/2) * \theta = \bar{X}$$

$$\hat{\theta} = 2\bar{X}$$

$$(2) f(x|\theta) = 1/\theta, \text{ for } 0 < x < \theta$$

$$g(\theta) = 1$$

$$\hat{\theta}_{\text{MAP}}(x) = \arg\max f(x|\theta)g(\theta) = 0$$

$$(3) f(x|\theta) = 1/\theta, \text{ for } 0 < x < \theta$$

$$h(\theta) = 1, \text{ for } 0 < \theta < 1$$

$$u(x, \theta) = h(\theta)f(x|\theta) = 1/\theta, \text{ for } 0 < x < \theta < 1$$

$$g(x) = \int_x^1 u(x, \theta) d\theta = \int_x^1 \frac{1}{\theta} d\theta = -\ln(x), \text{ for } 0 < x < 1$$

$$k(\theta|x) = u(x, \theta)/g(x) = -1/(\theta \ln(x)), \text{ for } 0 < x < \theta < 1$$

$$\hat{\theta} = E[\theta|x] = \int_x^1 \theta k(\theta|x) d\theta = -1/\ln(x) * \int_x^1 d\theta = (x-1)/\ln(x)$$

$$\hat{\theta} = (X-1)/\ln(X)$$

Problem 4 Goodness of Estimation

(1) set $c_1 + c_2 = 1$, $\text{Var}(c_1\theta_1 + c_2\theta_2) = c_1^2 + 2c_2^2 + 1/2 * c_1c_2 = c_1^2 + 2(1-c_1)^2 + 1/2 * c_1(1-c_1) = 5/2 * c_1^2 - 7/2 * c_1 + 2$, to minimize the expression, $c_1 = 7/10$, $c_2 = 3/10$,

unbiased estimator: $7/10 * \theta_1 + 3/10 * \theta_2$

variance: 0.775

(2) set $c_3 + c_4 = 1$, $\text{Var}(c_3\theta_3 + c_4\theta_4) = c_3^2 + 2c_4^2 + 3/2 * c_3c_4 = c_3^2 + 2(1-c_3)^2 + 3/2 * c_3(1-c_3) = 3/2 * c_3^2 - 5/2 * c_3 + 2$, to minimize the expression, $c_3 = 5/6$, $c_4 = 1/6$,

unbiased estimator: $5/6 * \theta_3 + 1/6 * \theta_4$

variance: 0.9583

(3) θ_1, θ_2 , whose covariance is lower.

Problem 5 Interval Estimation

(1) $P[Q \leq x] = P[X(1) - \theta \leq x] = P[\min[X_i] - \theta \leq x] = F_{\min(X) - \theta}(x)$

$$F_{\min(X) - \theta}(x) = \int_{\theta}^{x+\theta} e^{-(x-\theta)} dx = -e^{-(x-\theta)} = -e^{-x} + 1$$

Since it is independent of θ , Q is a pivotal quantity.

(2) $P(a \leq Q \leq b) = 1 - \alpha$

$$\alpha = P(a \leq X(1) - \theta \leq b) = P(a - X(1) \leq -\theta \leq b - X(1)) = P(b - X(1) \leq \theta \leq a - X(1))$$

$$= P(X_{1-\alpha/2}^2 - X(1) \leq \theta \leq X_{\alpha/2}^2 - X(1))$$

100(1- α)% confidence interval: $[X_{1-\alpha/2}^2 - X(1), X_{\alpha/2}^2 - X(1)]$