

Megoldások: kismintás tulajdonságok

Elek Péter

1. feladat. (Megoldás)

- $E(\hat{p}_1) = p$ and $E(\hat{p}_2) = 1/4 + p/2$, so only \hat{p}_1 is unbiased.
- $Var(\hat{p}_1) = p(1-p)/4$ and $Var(\hat{p}_2) = p(1-p)/16$.
- $MSE(\hat{p}_1) = p(1-p)/4$ because of unbiasedness, while

$$MSE(\hat{p}_2) = Var(\hat{p}_2) + Bias^2(\hat{p}_2) = (3p^2 - 3p + 1)/16.$$

By solving the inequality we obtain that $MSE(\hat{p}_2) < MSE(\hat{p}_1)$ as long as $0.173 < p < 0.827$, i.e. when p is in the "middle" of the unit interval.

2. feladat. (Megoldás)

- $E(\bar{X}) = E(X) = \theta$, so the sample mean is unbiased for θ . Its variance is $Var(\bar{X}) = Var(X)/n = \theta^2/n$.
- The "survival function" of the exponential distribution with parameter θ is $\Pr(X_i > x) = \exp(-x/\theta)$ for all $x > 0$ (because of the formula for its cdf). Since H_n is the sample minimum and our sample is i.i.d., this implies $\Pr(H_n > x) = \prod_{i=1}^n \Pr(X_i > x) = \exp(-nx/\theta)$ for all $x > 0$. Hence $\Pr(nH_n > x) = \exp(-x/\theta)$ for all $x > 0$, which is just the survival function of an exponential distribution with parameter θ . Hence, $E(R_n) = \theta$ and thus R_n is also unbiased. Its variance is given by $Var(R_n) = \theta^2$.
- Both estimators are unbiased but $Var(\bar{X})/Var(R_n) = 1/n < 1$ if $n > 1$. (In fact \bar{X} is much more efficient than R_n if n is large.) Hence we clearly prefer \bar{X} over R_n .

3. feladat. (Megoldás)

See Amemiya Example 7.2.5.

Note that $Var(\bar{X}) \sim c_1/n$ as $n \rightarrow \infty$ (as usual in i.i.d. samples), while $Var(T_n) \sim c_2/n^2$, so the rate of decay is faster in the second case. This is a "non-regular" example. (For those who are interested: this "strange" behavior is the result of the fact that the support of the pdf (i.e. the set of x -s for which the pdf is positive) depends on the estimated parameter θ .)