

In this lab., we will study the behavior of different estimators for a specific problem. In particular, we will have in mind that the performance of an unbiased estimator can be accessed by comparing its variance with the Cramér-Rao bound (CRB). This bound, which depends on the problem at hand, *i.e.*, on the model of the data, represents the smallest possible variance of an unbiased estimator.

Goal scoring in a soccer game. Recently, researchers have proposed¹ that the number of goals g a team scores against a particular opponent can be modeled as a Poisson random variable, *i.e.*, following

$$P(g) = \frac{\lambda^g e^{-\lambda}}{g!}, \quad g = 0, 1, 2, \dots$$

where λ is a free parameter. Note that $E\{g\} = \lambda$ and $\text{Var}\{g\} = E\{(g - \lambda)^2\} = \lambda$ (from the definition and using the expression above, you can show that these equalities hold).

Three students debate on what should be the score of team i on this week's match against team j . They are aware of the study above and try to find the expected number of goals team i scores in such a game, looking at the results of the past N matches between the same opponents. Let g_n , $n = 1, 2, \dots, N$ be the number of goals scored by team i against team j in match n between them. Student A notes that the last games are more relevant than the old ones for estimating the forthcoming result, thus should count more; student B says he knows nothing about soccer, proposing a sample mean; student C reminds that the decreasing strength of team i in the current season should be taken into account. Accordingly, they propose the following estimators for λ :

$$\hat{\lambda}_A = \frac{2}{N(N+1)} \sum_{n=1}^N n g_n, \quad \hat{\lambda}_B = \frac{1}{N} \sum_{n=1}^N g_n, \quad \hat{\lambda}_C = \frac{1}{2N} \sum_{n=1}^N g_n.$$

You want to check which estimator is the most reliable. Since the matches are distant in time, we consider their results as independent events, *i.e.*, we consider the problem of estimating the parameter λ of a Poisson random variable from a set of N independent observations, $\{g_1, g_2, \dots, g_N\}$.

Analytical performance evaluation.

1. Find the mean and variance of each of the estimators $\hat{\lambda}_A$, $\hat{\lambda}_B$, and $\hat{\lambda}_C$.
2. Write the log-likelihood function for the estimation problem and find the CRB for the estimate of λ .
3. For $\lambda = 1$, plot the CRB as a function of $N = 1, 2, \dots, 20$. Superimpose to this plot the estimator variances. Which is the best estimator? Why?

Simulation.

1. Using $\lambda = 1$ as the expected number of goals, simulate the past scores of $N = 20$ matches, by generating a 20×1 vector \mathbf{g} with independent samples of a Poisson random variable. `{poissrnd}`
2. Visualize the histogram of \mathbf{g} . Specify the bin centers with all the values from 0 to the maximum of g . Is this data model credible, according to your knowledge of the game?
3. Compute the estimates $\hat{\lambda}_A$, $\hat{\lambda}_B$, and $\hat{\lambda}_C$ from the data in \mathbf{g} and compare them with the histogram. Which seems more adequate?
4. To evaluate the performances of the estimators, use a Monte Carlo method: run 1000 times **1.** and **3.**, storing the values of the estimates; at the end, compute the mean square error (MSE) of each estimator with respect to the true value of λ . Comment and draw conclusions about which is the best estimator.

¹A. Heuer, C. Muller, and O. Rubner, "Soccer: Is scoring goals a predictable Poissonian process?", A Letters Journal Exploring the Limits of Physics, EPL 89 (2010).