

Project

You **must** return a single Jupyter notebook in Python on [ecampus](#). No other format will be accepted. Mathematical formulas must be written in Markdown / LaTeX. Python code must run without errors.

The deadline is October 30, 2021.

We focus on the cost of nuclear accidents before the accident of [Three Mile Island](#) that occurred on March 28, 1979. The dataset is available on [ecampus](#).

Exercise 1 (Statistical model):

1. Load data as a vector $x = (x_1, \dots, x_n)$ of nuclear accident costs before the Three Mile Island accident and remove all missing values. You must get $n = 55$ observations.
2. Let F_θ be the cumulative distribution function of a Gaussian distribution with mean μ and variance σ^2 , with $\theta = (\mu, \sigma^2)$.

(a) Show that the quantile function F_θ^{-1} satisfies:

$$\forall p \in (0, 1), \quad F_\theta^{-1}(p) = \sigma F_{(0,1)}^{-1}(p) + \mu.$$

This suggests that, if observations have a normal distribution, the corresponding [Q-Q plot](#) is well approximated by a line.

(b) Show the Q-Q plot of data for the Gaussian model using the [probplot](#) function of SciPy.

3. Let F_θ be the cumulative distribution function of an exponential distribution with parameter $\theta > 0$.

(a) Show that the quantile function F_θ^{-1} satisfies:

$$\forall p \in (0, 1), \quad F_\theta^{-1}(p) = \frac{1}{\theta} F_1^{-1}(p).$$

Again, if observations have an exponential distribution, the [Q-Q plot](#) is well approximated by a line.

(b) Show the Q-Q plot of data for the exponential model.

4. Discuss the results.

In the following, we use the **exponential model** for the cost of accidents before the Three Mile Island accident. Specifically, we assume that accident costs are i.i.d. samples of an exponential distribution with parameter $\theta > 0$. Let $\Theta = (0, +\infty)$ be the set of parameters. We denote by $X = (X_1, \dots, X_n)$ a random vector of n i.i.d. samples.

Exercise 2 (Point estimation):

We first focus on the estimation of θ or some functions of θ .

1. Give the maximum likelihood estimator $\hat{\theta}$ of θ .
2. Show on the same plot the histogram of data in density and the probability density function of the exponential distribution with parameter $\hat{\theta}(x)$.
3. We seek to estimate the expected cost $g(\theta) = \frac{1}{\theta}$. Let $\hat{g}(x) = \frac{1}{n} \sum_{i=1}^n x_i$. Show that \hat{g} is an efficient estimator of $g(\theta)$.
4. Compute $\hat{g}(x)$ from the available observations.
5. For any $\eta > 0$, define the estimator

$$\hat{g}_\eta = \eta \hat{g}.$$

Show that for some values of η (to be specified), the quadratic risks associated with the estimation of $g(\theta)$ satisfy $R(\theta, \hat{g}_\eta) < R(\theta, \hat{g})$ for all $\theta > 0$. Discuss this result.

6. Give η so that \hat{g}_η is an unbiased estimator of the median cost. Compute $\hat{g}_\eta(x)$ from the available observations. Compare with the empirical median.
7. For the previous value of η , compare the quadratic risks of \hat{g} and \hat{g}_η , both viewed as estimators of $g(\theta)$, depending on the number of samples n .

Exercise 3 (Hypothesis testing):

We wish to show that the expected cost of an accident is less than one billion dollars. This amounts to reject the null hypothesis H_0 that the expected cost of an accident is at least one billion dollars.

1. Give the null hypothesis H_0 and the alternative hypothesis H_1 as subsets of Θ .
2. Using Neyman-Pearson's approach, give a uniformly most powerful test at level α .
3. Apply the test at level $\alpha = 5\%$. Give the p -value and conclude.
4. For $n = 55$ samples and the parameter θ associated with an expected cost of one billion dollars, plot the probability density function of $\hat{g}(X)$ and show the rejection region of H_0 at level $\alpha = 5\%$.
5. Plot the power of the test at level α with respect to θ for $n = 10, 50, 100, 1000$ samples. Explain the results.
6. Propose a new test using the approximation of $\hat{g}(X)$ suggested by the Central Limit Theorem and give the result of this test.