

Probabilistic tools

Central Limit Theorem

Let $(X_n)_{n \geq 1}$ be a real and independent sequence with same law such that $\mu = \mathbb{E}[X_1]$ and $\mathbb{V}[X_1] = \sigma^2$ are defined ($\mathbb{V}[X_1] \leq +\infty$). Noting $\bar{X}_n = \frac{1}{n}(X_1 + \dots + X_n)$, we have:

$$\sqrt{n} \frac{(\bar{X}_n - \mu)}{\sigma} \underset{n \rightarrow \infty}{\sim} \mathcal{N}(0, 1)$$

Spectral Theorem

Let M be a symmetric matrix with real coefficients. Then it exists U orthogonal and D diagonal with real coefficients such that $M = UDU^T$.

Inferential statistics

Likelihood method

This method consists on finding the parameter that maximizes the likelihood:

$L(x_1, \dots, x_n; \theta) = f(X|\theta) = \prod_{i=1}^n f_{\theta}(x_i; \theta)$ which is the product of densities across all samples.

Intuitively, we want to find the θ that maximizes a certain event, that is, obtaining some data X (which is why we have $X|\theta$).

We often use the log in order to get rid of power coefficients appearing with the product.

likelihood equation: $\frac{d}{d\theta} \ln(L(x_1, \dots, x_n; \theta)) = 0$

Exploratory statistics