



Normal Maximum Likelihood

1 Why

We want to select a normal density which summarizes well a dataset.

2 Formulation

Let $D = (x^1, \dots, x^n)$ be a dataset in \mathbf{R} . We want to select a density from among normal densities, which require specifying a mean and covariance.

Following the principle of maximum likelihood, we want to solve

$$\begin{aligned} & \textbf{find} \quad \mu, \sigma \in \mathbf{R} \\ & \textbf{to maximize} \quad \prod_{k=1}^n \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{1}{2} \left(\frac{x^k - \mu}{\sigma}\right)^2\right) \\ & \textbf{subject to} \quad \sigma > 0 \end{aligned}$$

We call a solution to the above problem a *maximum likelihood normal density* with respect to the dataset.

3 Solution

Proposition 1. *Let (x^1, \dots, x^n) be a dataset in \mathbf{R} . Let f be a gaussian density with mean*

$$\frac{1}{n} \sum_{k=1}^n x^k$$

and covariance

$$\frac{1}{n} \sum_{k=1}^n \left(x^k - \frac{1}{n} \sum_{k=1}^n x^k \right)^2.$$

Then f is a maximum likelihood gaussian density.

Proof. Every gaussian density has two parameters: the mean and the covariance. If the likelihood of one gaussian is less than or equal to the likelihood of another, then so are their log likelihoods. Let f be a gaussian density with parameter μ and σ^2 . We express the log likelihood of f by

$$\sum_{k=1}^n \left(\frac{1}{2\sigma^2} (x^k - \mu)^2 - \frac{1}{2} \log 2\pi\sigma^2 \right)$$

The partial derivative of the log likelihood with respect to the mean $(\partial_\mu \ell) : \mathbf{R}^2 \rightarrow \mathbf{R}$ is

$$(\partial_\mu \ell)(\mu, \sigma^2) = - \sum_{k=1}^n \frac{1}{\sigma^2} (x^k - \mu)$$

and with respect to the covariance $(\partial_{\sigma^2} \ell) : \mathbf{R}^2 \rightarrow \mathbf{R}$ is

$$(\partial_{\sigma^2} \ell)(\mu, \sigma^2) = \left(\frac{-1}{2(\sigma^2)^2} \sum_{k=1}^n (x^k - \mu)^2 \right) - \frac{1}{2\sigma^2}$$

We are interested in finding $\mu_0 \in \mathbf{R}$ and $\sigma_0^2 > 0$, at which $\partial_\mu \ell(\mu_0, \sigma_0^2) = 0$ and $\partial_{\sigma^2} \ell(\mu_0, \sigma_0^2) = 0$. So we have two equations. First, notice that $\partial_\mu \ell$ is zero if and only if its first argument (the mean) is $\frac{1}{n} \sum_{k=1}^n x^k$. Second, notice that for all μ, σ^2 , $\partial_{\sigma^2} \ell$ is zero if and only if

$$\sigma^2 = \sum_{k=1}^n (x^k - \mu)^2.$$

So the pair

$$\left(\frac{1}{n} \sum_{k=1}^n x^k, \frac{1}{n} \sum_{k=1}^n (x_k - \frac{1}{n} \sum_{k=1}^n x^k)^2 \right)$$

is a stationary point of ℓ . □