

3.4.6 Parameter estimation (Density estimation) with Laplace prior.

X is a dataset of observables \vec{x}_n : $X = (\vec{x}_1, \vec{x}_2, \dots, \vec{x}_N)$.

We model them as Gaussians (i.i.d): $x_i \sim \mathcal{N}(\mu, \sigma) \perp x_j$,

where μ and σ have to be estimated. ($\forall i \neq j$)

From knowing about the problem, we have a prior on μ :

$$P(\mu) \propto \text{Laplace}(0, b) = \frac{1}{2b} e^{-|\mu|/b}$$

The MAP estimate is:

$$\begin{aligned} \mu_{\text{MAP}} &= \underset{\mu}{\text{argmax}} P(\mu, \sigma | X = \{\dots\}) \\ &= \underset{\mu}{\text{argmax}} \log \left(\frac{P(X = \{\dots\} | \mu) P(\mu)}{P(X = \{\dots\})} \right) \end{aligned}$$

(We can insert log because it's a monotonously increasing function, and is well defined since probas are ≥ 0).

(Also $P(X = \{\dots\})$ does not depend on μ , it is a constant w.r.t. the argmax)

$$\begin{aligned} \mu_{\text{MAP}} &= \underset{\mu}{\text{argmax}} \left(\sum_{n=1}^N \log(P(x_n = x_n | \mu)) + \log P(\mu) \right) \\ &= \underset{\mu}{\text{argmax}} \left(\sum_{n=1}^N \log \left(\frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x_n - \mu)^2}{2\sigma^2}} \right) + \log \left(\frac{1}{2b} e^{-|\mu|/b} \right) \right) \\ &= \underset{\mu}{\text{argmax}} \left(\sum_{n=1}^N \left(\underbrace{-\frac{(x_n - \mu)^2}{2\sigma^2}}_{\left(\frac{1}{2\sigma^2}\right) \text{ from } \mu} \right) - \underbrace{\frac{|\mu|}{2b}}_{\perp \text{ from } \mu} \right) \\ &= \underset{\mu}{\text{argmin}} \left(+ \sum_{n=1}^{N-1} (x_n - \mu)^2 + \frac{\sigma^2}{b} |\mu| \right) \end{aligned}$$

we switch the sign and multiply by $2\sigma^2$

We can search for minima by asking the (pseudo) gradient of this to be 0.

$$\mathcal{L} = \sum_{n=1}^N (x_n - \mu)^2 + \frac{\sigma^2}{b} |\mu|$$

There are 2 cases:

$$\mu \neq 0 \Rightarrow \nabla_{\mu} |\mu| = \pm 1 = \text{sign}(\mu)$$

$$\mu = 0 \Rightarrow \nabla_{\mu} |\mu| \in [-1, 1]$$

(By continuity. One can think of a function like $\text{abs}(\cdot)$, but rounded at zero, over a range of ε , with $\varepsilon \rightarrow 0$)
 \hookrightarrow (To be discussed another time).

Case 1: $\mu \neq 0$.

$$\nabla_{\mu} \mathcal{L} = 0 \Leftrightarrow \nabla_{\mu} \sum_{n=1}^N (x_n - \mu)^2 + \frac{\sigma^2}{b} \text{sign} \mu = 0$$

$$\Leftrightarrow -2 \sum_{n=1}^N (x_n - \mu) + \frac{\sigma^2}{b} \text{sign} \mu = 0$$

$$2N\mu - 2 \sum_{n=1}^N x_n + \frac{\sigma^2}{b} \text{sign}(\mu) = 0$$

$$\mu - \bar{x} + \frac{\sigma^2}{2bN} \text{sign}(\mu) = 0, \quad \bar{x} = \frac{1}{N} \sum_{n=1}^N x_n$$

2 sub-cases: $\mu_{\text{MAP}} > 0$ and $\mu_{\text{MAP}} < 0$

$$* \mu > 0: \mu - \bar{x} + \frac{\sigma^2}{2bN} \times (+1) = 0$$

$$\Rightarrow \mu_{\text{MAP}} = \bar{x} - \frac{\sigma^2}{2bN} (> 0) \Leftrightarrow (\bar{x} > \frac{\sigma^2}{2bN})$$

$$* \mu < 0: \mu - \bar{x} + \frac{\sigma^2}{2bN} \times (-1) = 0$$

$$\Rightarrow \mu_{\text{MAP}} = \bar{x} + \frac{\sigma^2}{2bN} (< 0) \Leftrightarrow (\bar{x} < -\frac{\sigma^2}{2bN})$$

Remark: in both cases, we notice that $|\bar{x}| > \frac{\sigma^2}{2bN}$.

$$\text{---}, \text{sign}(\mu_{\text{MAP}}) = \text{sign} \bar{x}$$

Case 2: $\mu = 0$

$$\nabla_{\mu} \mathcal{L} = 0 \Leftrightarrow -2 \sum_{n=1}^N (x_n - \mu) + \frac{\sigma^2}{b} \nabla_{\mu} |\mu| \stackrel{\mu=0}{=} 0 \quad \downarrow \frac{1}{2bN}$$

$$\bar{x} - \bar{x} + \frac{\sigma^2}{2bN} \cdot \nabla_{\mu} |\mu| = 0$$

$$\nabla_{\mu} |\mu| = \frac{1}{\left(\frac{\sigma^2}{2bN}\right)} \cdot \bar{x}$$

Is this $(\nabla_{\mu} |\mu|)$ in the range $[-1, 1]$?

It is, iff: $-1 < \nabla_{\mu} |\mu| < 1$

$$\Leftrightarrow -1 < \frac{1}{\left(\frac{\sigma^2}{2bN}\right)} \cdot \bar{x} < 1$$

$$\Leftrightarrow -\frac{\sigma^2}{2bN} < \bar{x} < \frac{\sigma^2}{2bN}$$

This is complementary to the range found in case 1 ($\mu \neq 0$)

Summary: $\mathcal{L} = \sum_n (x_n - \mu)^2 + \frac{\sigma^2}{b} |\mu|$ is minimized by $\mu = \mu_{\text{MAP}}$:

$$\mu_{\text{MAP}} = \begin{cases} \text{if } |\bar{x}| > \frac{\sigma^2}{2bN}, & \mu_{\text{MAP}} = \bar{x} - \text{sign}(\bar{x}) \cdot \frac{\sigma^2}{2bN} \\ \text{if } |\bar{x}| < \frac{\sigma^2}{2bN}, & \mu_{\text{MAP}} = 0. \end{cases}$$

Is this sol^o continuous in \bar{x} ?

When $\bar{x} \rightarrow \left(\frac{\sigma^2}{2bN}\right)^+ (= \frac{\sigma^2}{2bN} + \varepsilon, \varepsilon \rightarrow 0^+)$,

$$\mu_{\text{MAP}} = \bar{x} - \frac{\sigma^2}{2bN} = \varepsilon \rightarrow 0^+$$

Yes, the sol^o is C⁰:

