

TDA231 - Homework 1

Theoretical Problems

Problem 1.1

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Problem 1.1

$$N(\mu, \sigma^2 I)$$

$$L = p(x) = \frac{1}{(2\pi)^{\frac{n}{2}} |\sigma^2 I|^{\frac{1}{2}}} \cdot \exp\left(-\frac{1}{2} (x-\mu)^T (\sigma^2 I)^{-1} (x-\mu)\right)$$

$$= \frac{1}{(2\pi)^{\frac{n}{2}} \cdot \sigma^n} \cdot \exp\left(-\frac{1}{2} (x-\mu)^T \left(\frac{1}{\sigma^2} I\right) (x-\mu)\right)$$

$$\ln L = \ln(1) - \left(\ln(2\pi)^{\frac{n}{2}} + \ln(\sigma^n)\right) + \left(-\frac{1}{2} (x-\mu)^T \left(\frac{1}{\sigma^2} I\right) (x-\mu)\right)$$

$$= -\frac{n}{2} \ln(2\pi) - n \ln(\sigma) - \frac{1}{2\sigma^2} (x-\mu)^T I (x-\mu)$$

$$\frac{d \ln L}{d \sigma} = 0 - \frac{n}{\sigma} + \frac{1}{\sigma^3} (x-\mu)^T I (x-\mu) = 0$$

$$\Leftrightarrow \frac{1}{\sigma^2} (x-\mu)^T I (x-\mu) = n$$

$$\Leftrightarrow \frac{1}{n} \sum_{i=1}^n (x_i - \mu_i)^2 = \sigma^2$$

$$\Rightarrow \hat{\sigma} = \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \mu_i)^2}$$

Problem 1.2

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1.2 a) $\mathcal{D} = \{x_1, \dots, x_n\}$

$$p(\sigma^2 = s | \mathcal{D}) \propto \prod_{i=1}^n P(X=x_i | \sigma^2) P(\sigma^2 = s | \alpha, \beta)$$

$$= \frac{\beta^\alpha s^{-\alpha-1}}{2\pi s \Gamma(\alpha)} \exp\left(\frac{\sum_{i=1}^n (x_i - \mu_i)^T (x_i - \mu_i) - 2\beta}{s}\right)$$

b)

$$BF = \frac{P(\mathcal{D} | M_A)}{P(\mathcal{D} | M_B)} = \frac{\int P(\mathcal{D} | \sigma^2 = s) P(\sigma^2 = s | \alpha_A, \beta_A) ds}{\int P(\mathcal{D} | \sigma^2 = s) P(\sigma^2 = s | \alpha_B, \beta_B) ds}$$

c) $P(M_A) = P(M_B) = \frac{1}{2}$

$$\hat{\sigma}_{MAP}^2 = \frac{d}{d\sigma} \left(\argmax \log(P(\mathcal{D} | \sigma^2, \mu = \bar{x}) P(\sigma^2 = s | \alpha, \beta)) \right) = 0$$

$$\Leftrightarrow \sigma^2 = \frac{1}{2\alpha} \cdot \left(\sum_{i=1}^n [(x_i - \bar{x})^T (x_i - \bar{x})] + 2\beta \right)$$

$$P(M | \mathcal{D}) = \frac{P(\mathcal{D} | M) P(M)}{P(\mathcal{D})} \Leftrightarrow P(\mathcal{D} | M) = \frac{P(M | \mathcal{D}) P(\mathcal{D})}{P(M)}$$

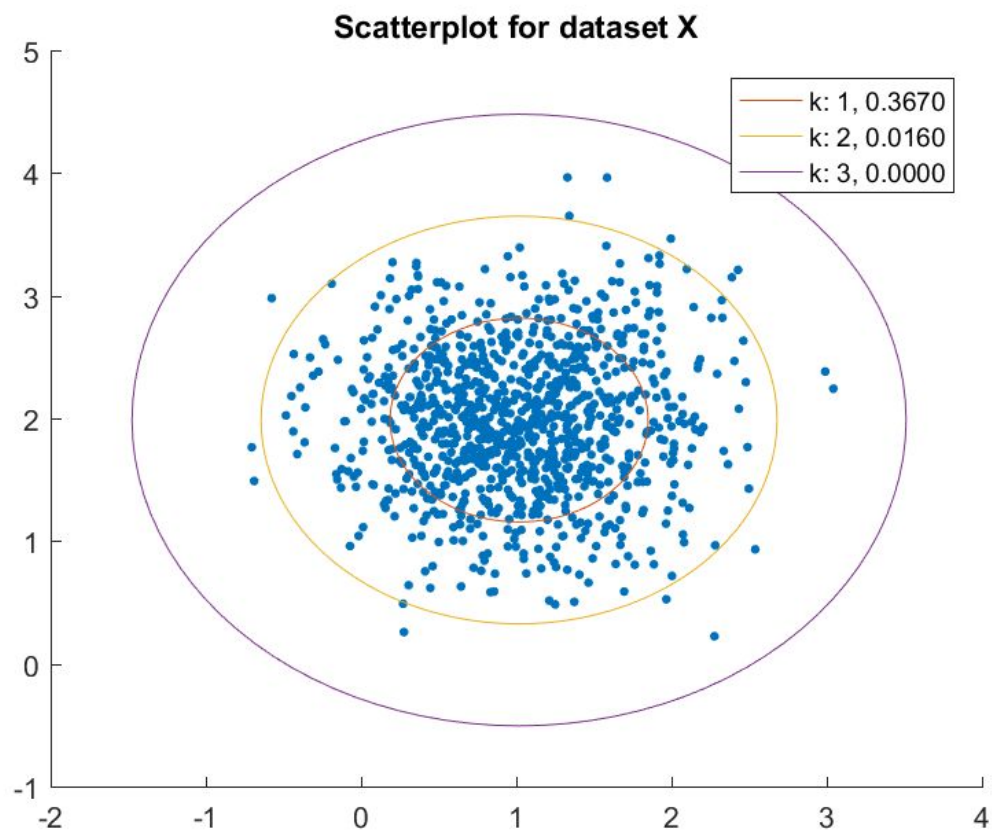
$$BF = \frac{P(\mathcal{D} | M_A)}{P(\mathcal{D} | M_B)} = \frac{P(M_A | \mathcal{D}) P(M_B)}{P(M_B | \mathcal{D}) P(M_A)} = \frac{P(M_A | \mathcal{D})}{P(M_B | \mathcal{D})}$$

Matlab problems

Problem 2.1

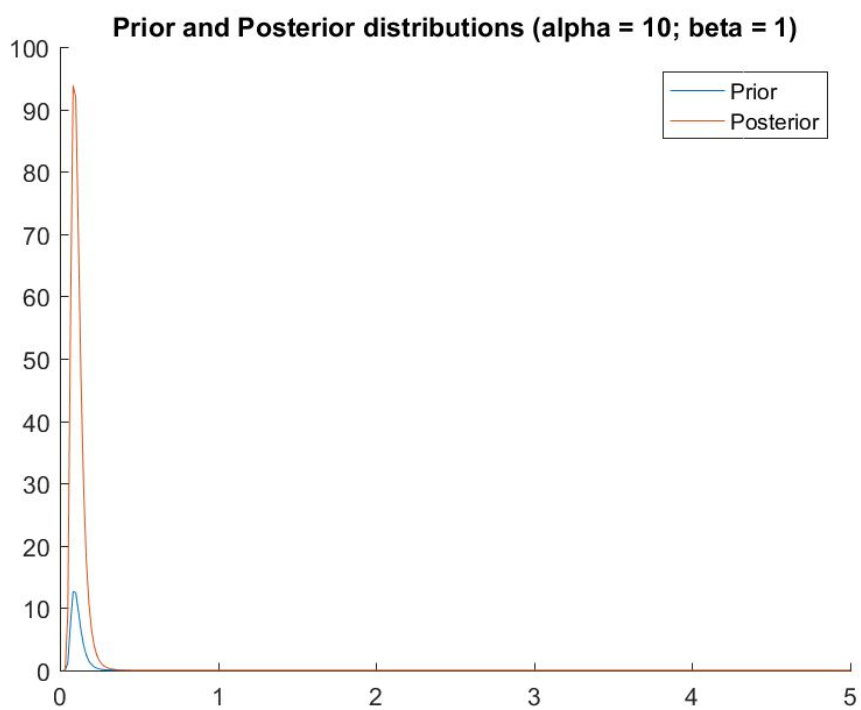
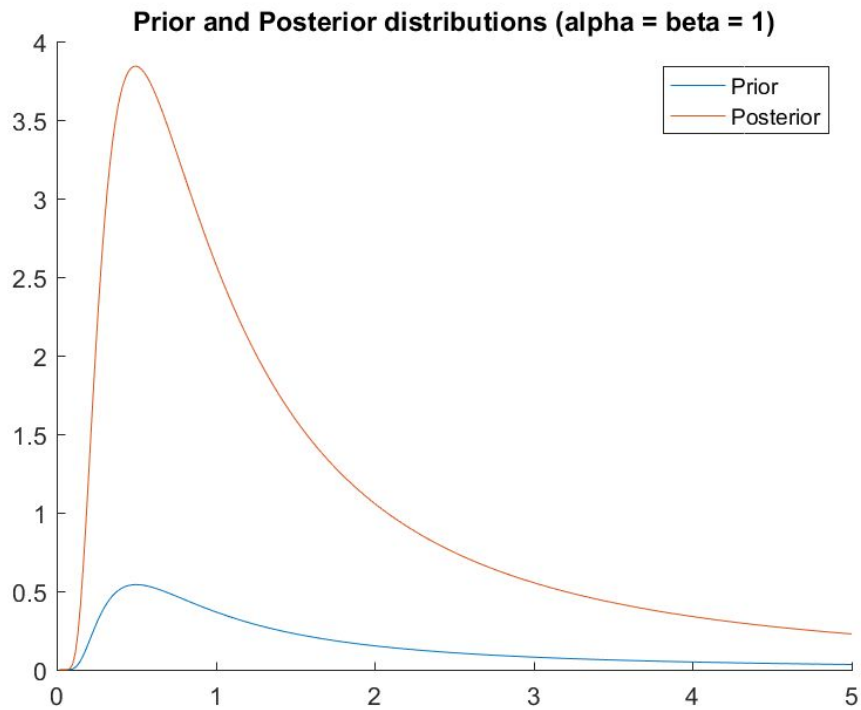
$$\mu = \frac{\sum x}{n}$$

$$\sigma = \frac{1}{n} \sqrt{\sum_{i=1}^n (x_i - \mu_i) \cdot (x_i - \mu_i)^T}$$



Problem 2.2

a)



b) Map estimate for $\sigma^2 = s$

The analytical model has been derived in Task 1.2 c).

Model A (alpha=beta=1)

$$s_{MAP} = 346.3813$$

Model B (alpha=10, beta=1)

$$s_{MAP} = 34.63813$$

c) Bayes Factor

$$BF = \frac{P(M_A|D)}{P(M_B|D)} = 2.61717 * 10^{17} > 1$$

Since the bayes factor BF is greater than 1 Model A is the better model.