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Machine Learning with Python-From Linear Models to Deep Learning

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A Course / Unit 4. Unsupervised Learning (2 weeks) / Lecture 15. Generative Mo



10. MLE for Gaussian Distribution

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Exercises due Apr 19, 2023 08:59 -03 Completed

MLEs for Gaussian Distribution



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MLE for the Gaussian Distribution

1/1 point (graded)

In this problem, we will derive the maximum likelihood estimates for a Gaussian model.

Let X be a Gaussian random variable in d-dimensional real space (\mathbb{R}^d) with mean μ and

Note that μ, σ are the parameters of a Gaussian generative model.

Recall from the lecture that, the probability density function for a Gaussian random variation

$$f_X(x \mid \mu, \sigma^2) = \frac{1}{(2\pi\sigma^2)^{d/2}} e^{-\|x-\mu\|^2/2\sigma^2}$$

-1 - (-1) - (2) - (n) + (n)

$$= -\frac{nd}{2}\log(2\pi\sigma^2) - \frac{1}{2\sigma^2} \sum_{t=1}^{n} \|\chi^{(t)}\|_{X}$$

Compute the partial derivative $\frac{\partial \log P(S_n | \mu, \sigma^2)}{\partial \mu}$ using the above derived expression for lo

Choose the correct expression from options below.

$$\frac{\partial \log P(S_n|\mu,\sigma^2)}{\partial \mu} = -\frac{1}{\sigma^2} \sum_{t=1}^n (x^{(t)} - \mu)$$

$$\frac{\partial \log P(S_n | \mu, \sigma^2)}{\partial \mu} = \frac{1}{\sigma^2} \sum_{t=1}^n (x^{(t)} - \mu)$$

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$$\frac{\partial \log P(S_n|\mu,\sigma^2)}{\partial \mu} = -\frac{1}{\mu^2} \sum_{t=1}^n (x^{(t)} - \mu)$$



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You have used 1 of 2 attempts

MLE for the Mean

1/1 point (graded)

Use the answer from the previous problem in order to solve the following equation

$$\frac{\partial \log P(S_n | \mu, \sigma^2)}{\partial \mu} = 0$$

Compute expression for $\hat{\mu}$ that is a solution for the above equation.

Choose the correct expression from options below

$$\hat{\mu} = \prod_{t=1}^n x^{(t)}$$

1/1 point (graded)

Compute the partial derivative $\frac{\partial \log P(S_n|\mu,\sigma^2)}{\partial \sigma^2}$ using the above derived expression for $\log P(S_n|\mu,\sigma^2)$ restated below as well:

$$\log P(S_n | \mu, \sigma^2) = -\frac{nd}{2} \log(2\pi\sigma^2) - \frac{1}{2\sigma^2} \sum_{t=1}^n \|x^{(t)} - \mu\|^2$$

Choose the correct expression from options below.

$$\frac{\partial \log P(S_n | \mu, \sigma^2)}{\partial \sigma^2} = \frac{nd}{2\sigma^2} + \frac{\sum_{t=1}^n \|x^{(t)} - \mu\|^2}{2(\sigma^2)^2}$$

$$\frac{\partial \log P(S_n | \mu, \sigma^2)}{\partial \sigma^2} = -\frac{nd}{2\sigma^2} + \frac{\sum_{t=1}^n \|x^{(t)} - \mu\|^2}{2(\sigma^2)^2}$$

$$\frac{\partial \log P(S_n | \mu, \sigma^2)}{\partial \sigma^2} = \frac{nd}{2\sigma^2} - \frac{\sum_{t=1}^n \|x^{(t)} - \mu\|^2}{2(\sigma^2)^2}$$

$$\frac{\partial \log P(S_n | \mu, \sigma^2)}{\partial \sigma^2} = \frac{\sum_{t=1}^n \|x^{(t)} - \mu\|^2}{2(\sigma^2)^2}$$



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MLE for the Variance II

1/1 point (graded)

Using the answer from the previous problem in order to solve the equation

$$\frac{\partial \mathrm{log}P(S_n | \mu, \sigma^2)}{\partial \sigma^2} = 0$$

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Compute expression for σ^- that is a solution for the above equation.

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Discussion

Topic: Unit 4. Unsupervised Learning (2 weeks) :Lecture 15. Generative Models / 10. MLE for Gaussian Distribution

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- Partial derivative or gradient?
- ? What is the sequence of the steps we are doing for the generative models?
- ? What is MLE?
 Sorry for the stupid question but I'm bad with abbreviations

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