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9. Gaussian Generative models

Gaussian Generative models

And then here, we have exponent minus 1 divided by 2

sigma squared is x minus μ squared.

OK?

So this is the likelihood of a particular point being

generated by the Gaussian.

And you can see that if we select different μ s and different sigmas,

the same point may get different likelihoods.



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Gaussian distribution

1/1 point (graded)

Recall that the likelihood of x being generated from a gaussian with mean μ and std σ is:

$$P(x|\mu, \sigma^2) = \frac{1}{(2\pi\sigma^2)^{d/2}} \exp\left(-\frac{1}{2\sigma^2} \|x - \mu\|^2\right)$$

Let $x = \begin{bmatrix} \frac{1}{\sqrt{\pi}} \\ 2 \end{bmatrix}$ be a vector in the two dimensional space.

Let G be a two-dimensional gaussian distribution with mean μ and standard deviation σ taking values as follows

$$\mu = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \sigma = \sqrt{\frac{1}{2\pi}}$$

Calculate the probability $p(x|\mu, \sigma^2)$ of x being sampled from the gaussian distribution G with mean μ and variance σ^2 taking values as given above.

Enter the value of $\log p(x|\mu, \sigma^2)$ below (note that we use \log for the natural logarithm, i.e. \log_e)

✓ Answer: -1

Solution:

Note that the probability of vector x being sampled from a gaussian distribution G with mean μ and variance σ^2 is given as follows

$$P(x|\mu, \sigma^2) = \frac{1}{2\pi\sigma^2} \exp\left(-\frac{1}{2\sigma^2} \|x - \mu\|^2\right)$$

Substituting the value of $\sigma = \sqrt{\frac{1}{2\pi}}$ from above, we have

$$P(x|\mu, \sigma^2) = \frac{1}{2\pi \frac{1}{2\pi}} \exp\left(-\frac{1}{2 \frac{1}{2\pi}} \|x - \mu\|^2\right)$$

$$P(x|\mu, \sigma^2) = \exp(-\pi \|x - \mu\|^2)$$

Substituting the value of $x = \begin{bmatrix} \frac{1}{\sqrt{\pi}} \\ 2 \end{bmatrix}$ and $\mu = \begin{bmatrix} 0 \\ 2 \end{bmatrix}$, we have

$$P(x|\mu, \sigma^2) = \exp\left(-\pi \left(\left(\frac{1}{\sqrt{\pi}} - 0\right)^2 + (2 - 2)^2\right)\right)$$

$$P(x|\mu, \sigma^2) = \exp\left(-\pi \frac{1}{\pi}\right)$$

$$P(x|\mu, \sigma^2) = \exp(-1)$$

$$\ln(P(x|\mu, \sigma^2)) = \ln(\exp(-1)) = -1$$

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You have used 1 of 2 attempts

i Answers are displayed within the problem

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? What is d in the formula of the Gaussian distribution?
Is it the dimension of the vector x? And how does it affect the formula?

2 ▼

