## **Machine Learning**

## Quiz 2

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- 1. Consider a set of observations  $\mathbf{X} = \{\mathbf{x}_1, \dots, \mathbf{x}_N\}^T$ ,  $\mathbf{x}_n \in \mathbb{R}^D$ , drawn independently from a multivariate Gaussian distribution with known covariance matrix  $\Sigma$ ,  $\mathcal{N}(\mathbf{x}_n \mid \mu, \Sigma)$ .
  - (a) (2 points) State the *Maximum Likelihood* problem of finding the best estimate for the Gaussian mean,  $\mu_{ML}$ , given the observed data  $\mathbf{X}$  (*Hint:* Your answer should include  $\mathcal{N}(\mathbf{x}_n \mid \mu, \Sigma)$ ):

$$\mu_{\mathsf{ML}} = \underset{\mu}{\mathsf{arg\,max}}$$

(b) (1 point) What is the Maximum Likelihood estimate of the mean (actual solution to (a))?

$$\mu_{\mathsf{ML}} =$$

2. (1 point) In *Bayesian inference*, we maximize the *posterior probability*  $p(\mu \mid \mathbf{X}) \propto p(\mathbf{X} \mid \mu, \Sigma) p(\mu)$ , instead. Considering the prior  $p(\mu) = \mathcal{N}(\mu \mid 0, \Sigma_0)$ , how will the inferred  $\mu_{\text{MAP}}$  differ from  $\mu_{\text{ML}}$ ? (math expression or word answer is ok)