

Simple clustering example

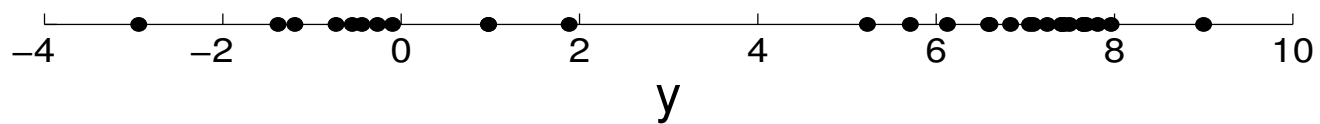
Paper 8, Easter term 2013

Richard E. Turner (ret26@cam.ac.uk)

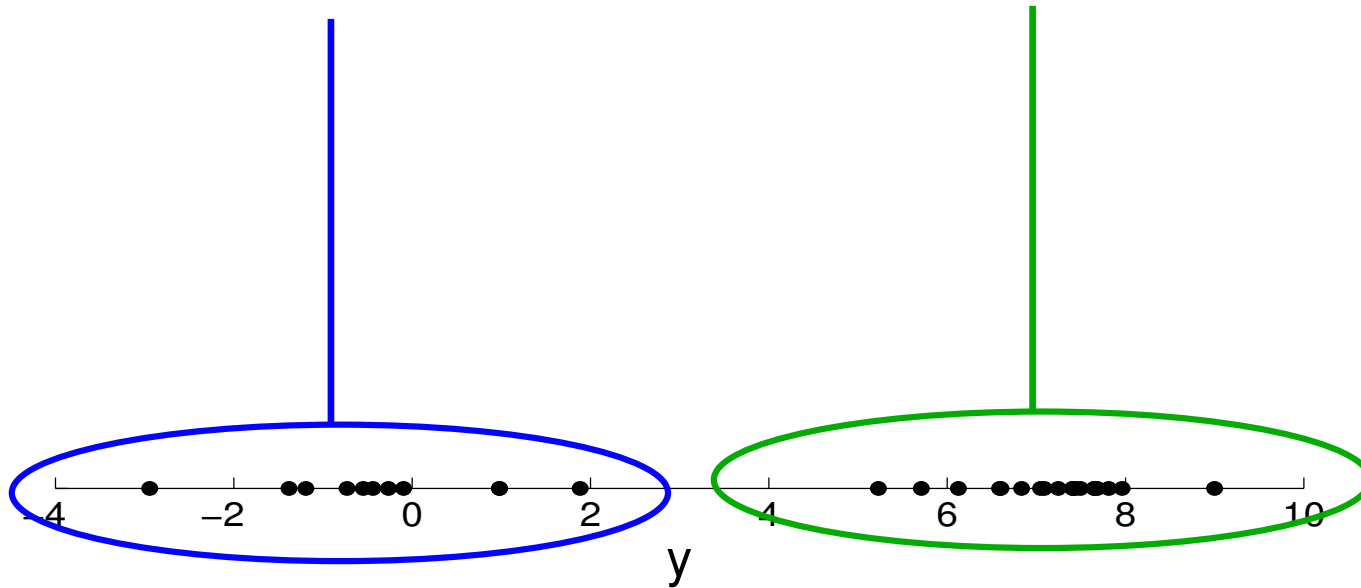
Computational and Biological Learning Lab

University of Cambridge

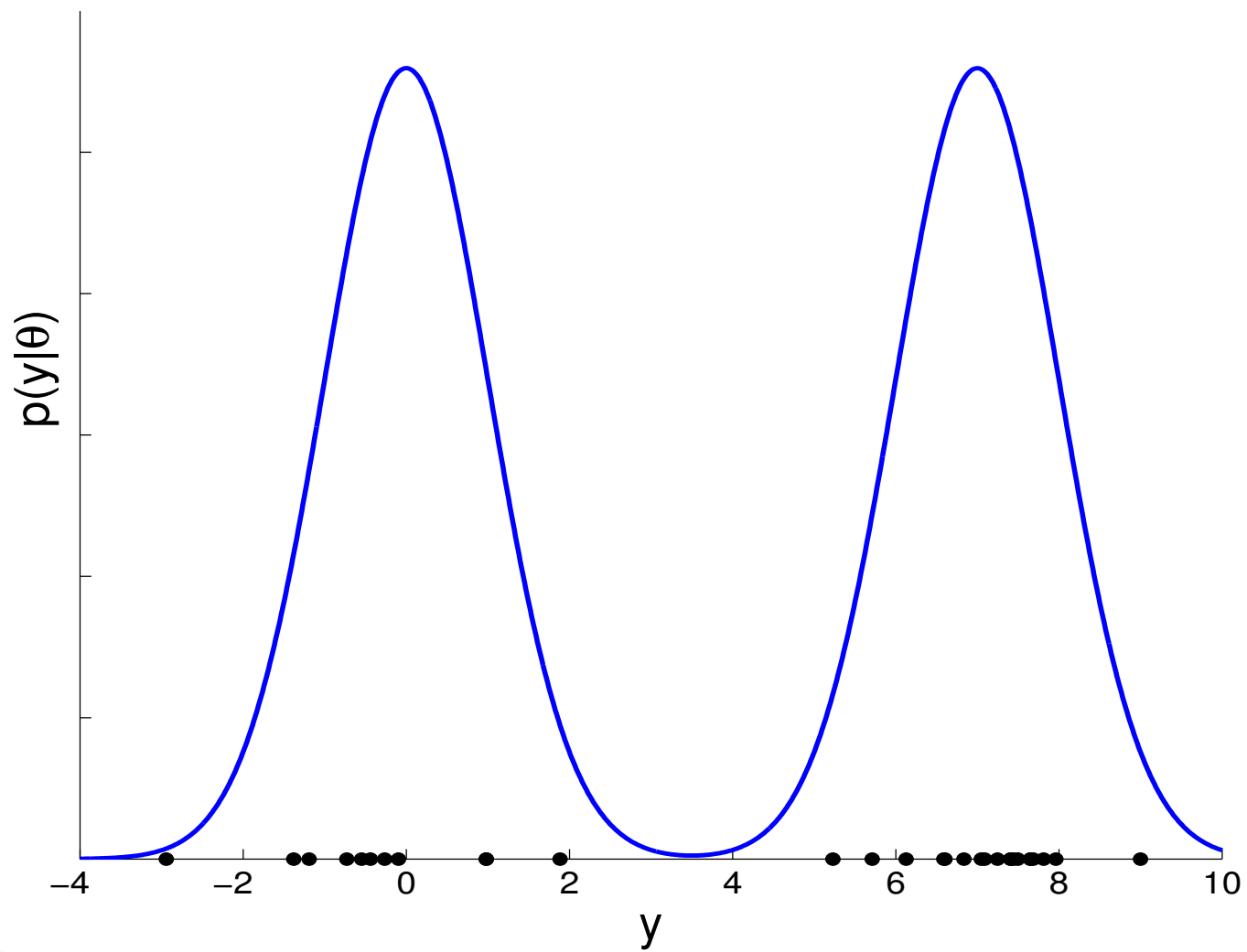
Clustering



Clustering

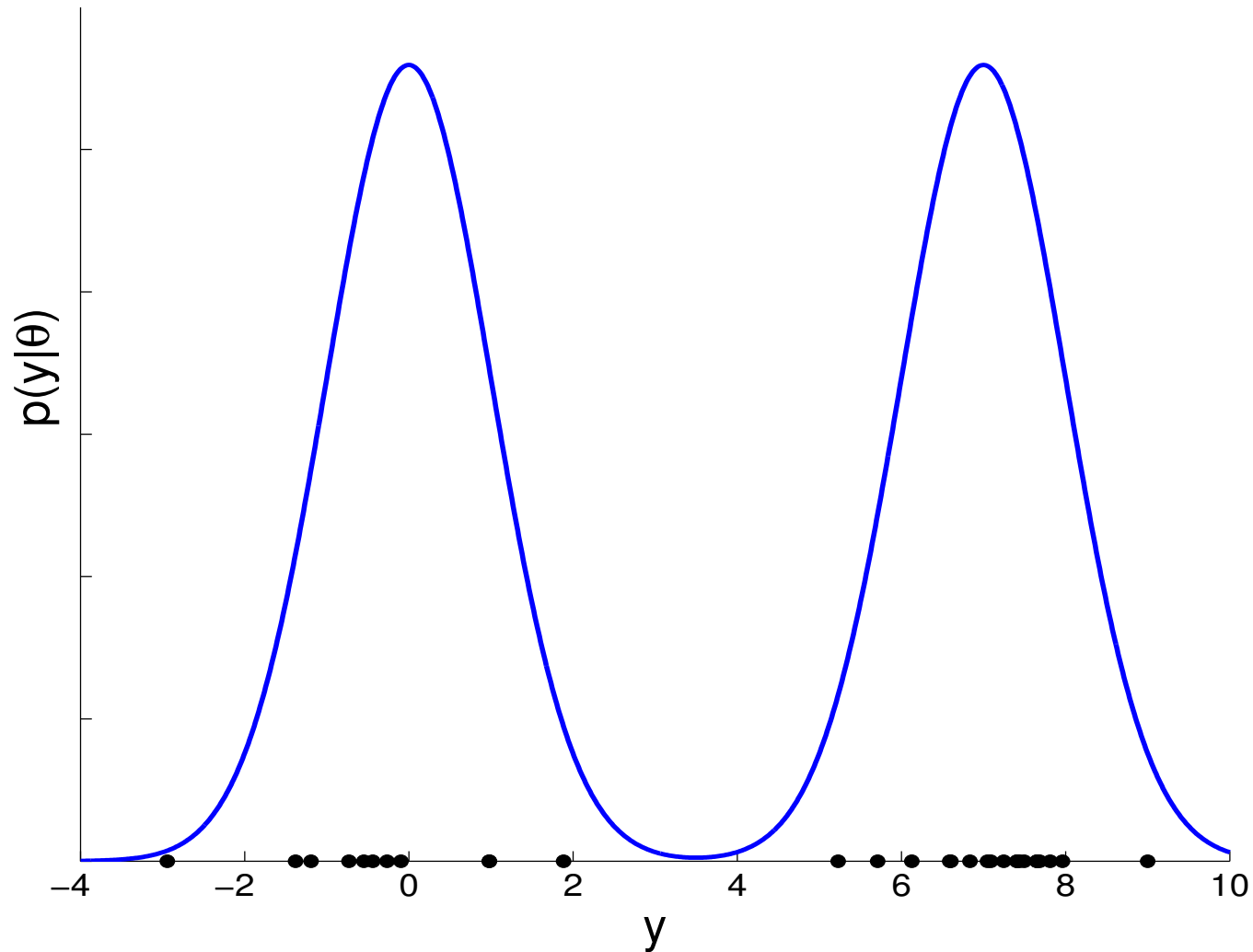


Mixture of Gaussians



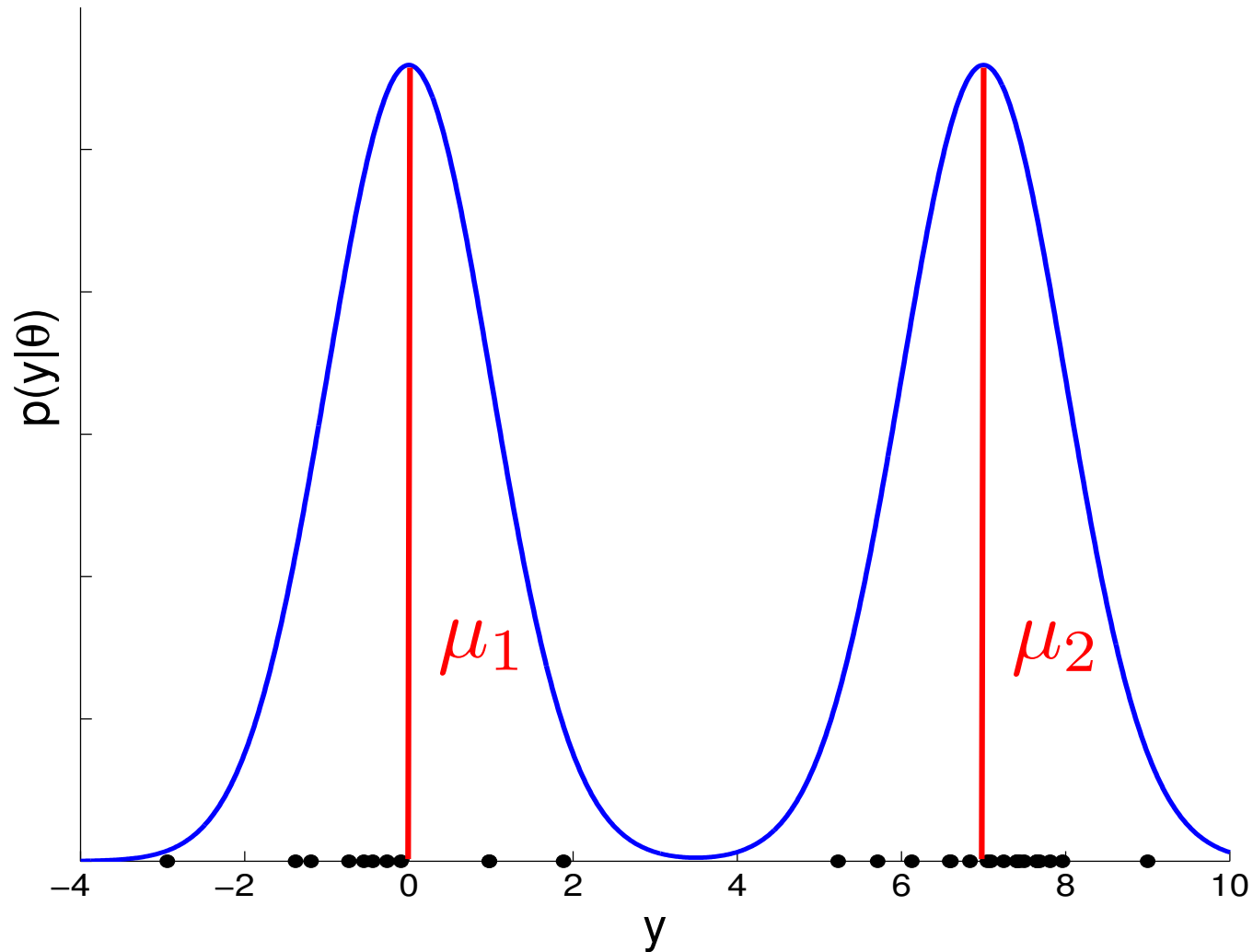
Mixture of Gaussians

$$p(y|\theta) = \frac{1}{2\sqrt{2\pi}} \exp\left(-\frac{1}{2}(y - \mu_1)^2\right) + \frac{1}{2\sqrt{2\pi}} \exp\left(-\frac{1}{2}(y - \mu_2)^2\right)$$



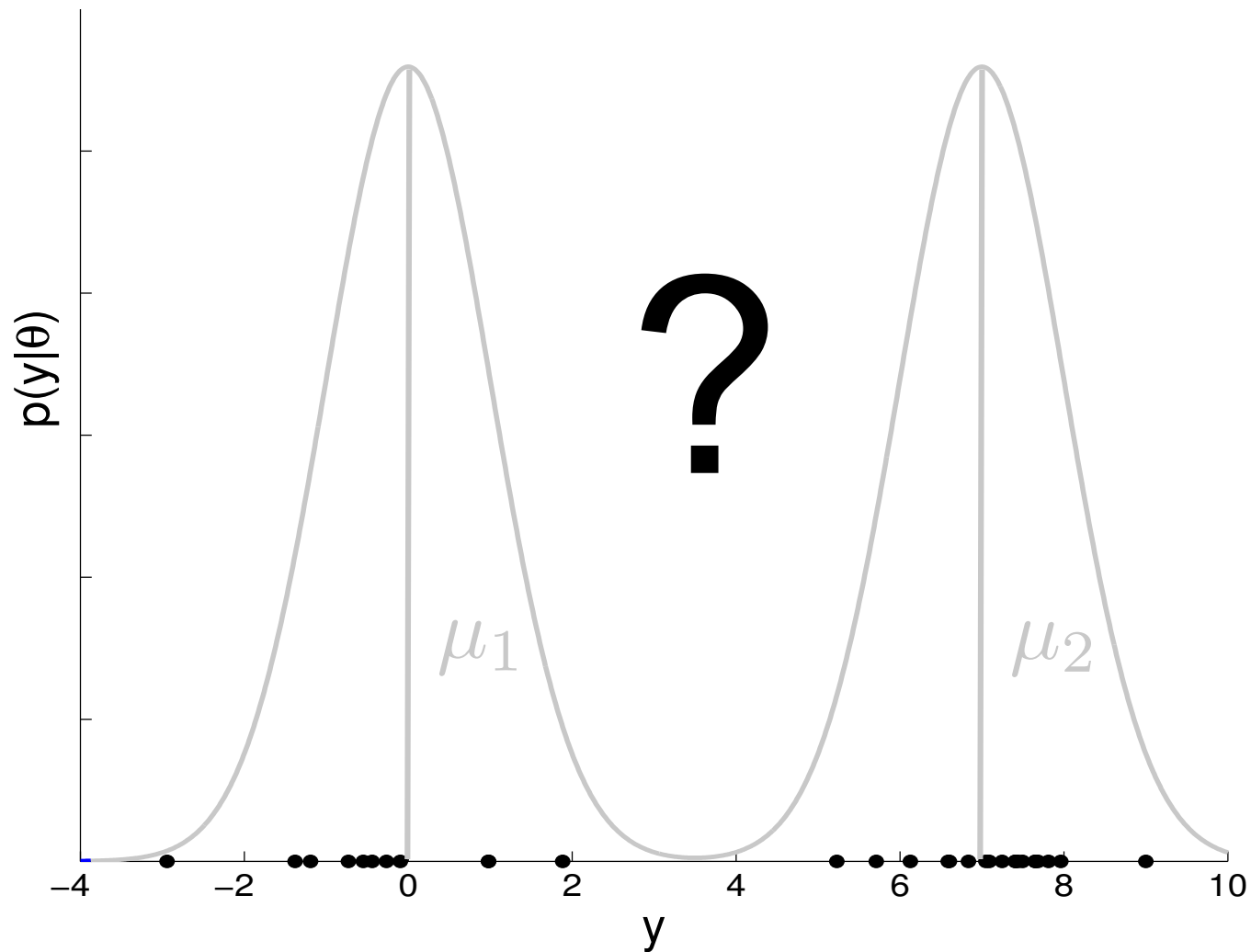
Mixture of Gaussians

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How do we estimate μ_1 and μ_2 from the data?

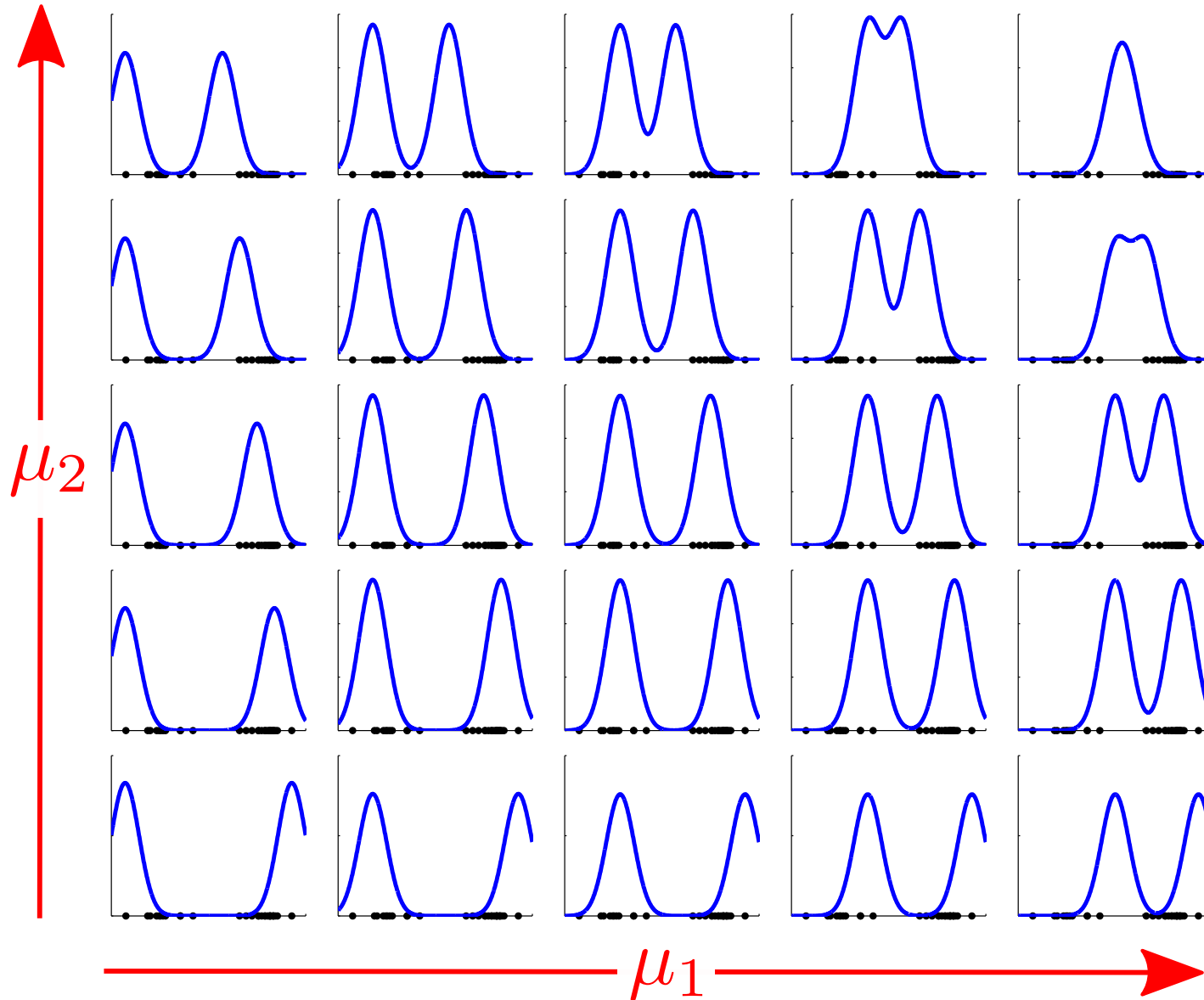
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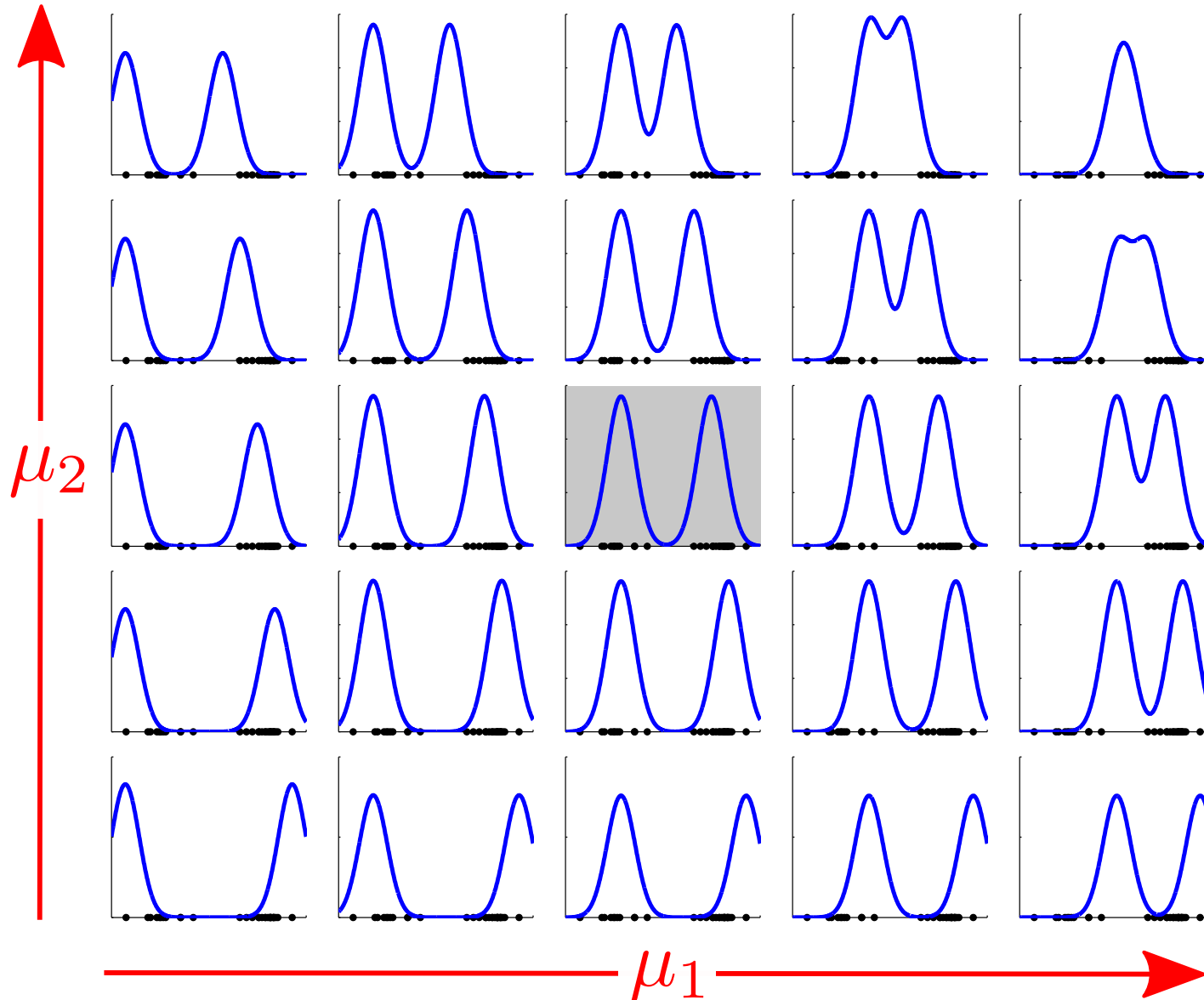
Maximum likelihood learning

$$\begin{aligned} p(y_1, y_2, \dots, y_N | \theta) &= \prod_{n=1}^N p(y_n | \theta) \\ &= (8\pi)^{-N/2} \prod_{n=1}^N \left[\exp\left(-\frac{1}{2}(y_n - \mu_1)^2\right) + \exp\left(-\frac{1}{2}(y_n - \mu_2)^2\right) \right] \end{aligned}$$

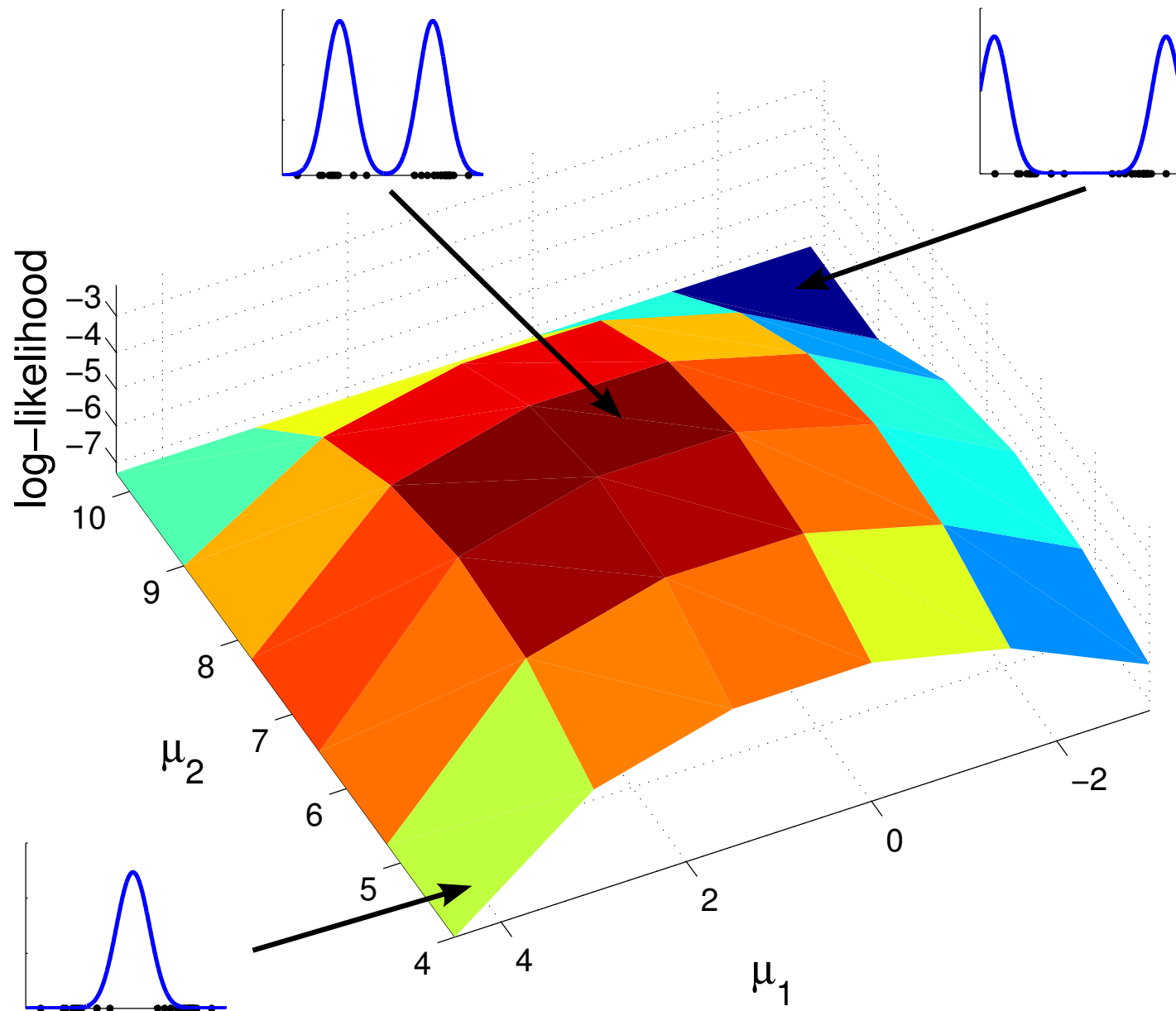
Which parameters have highest likelihood?



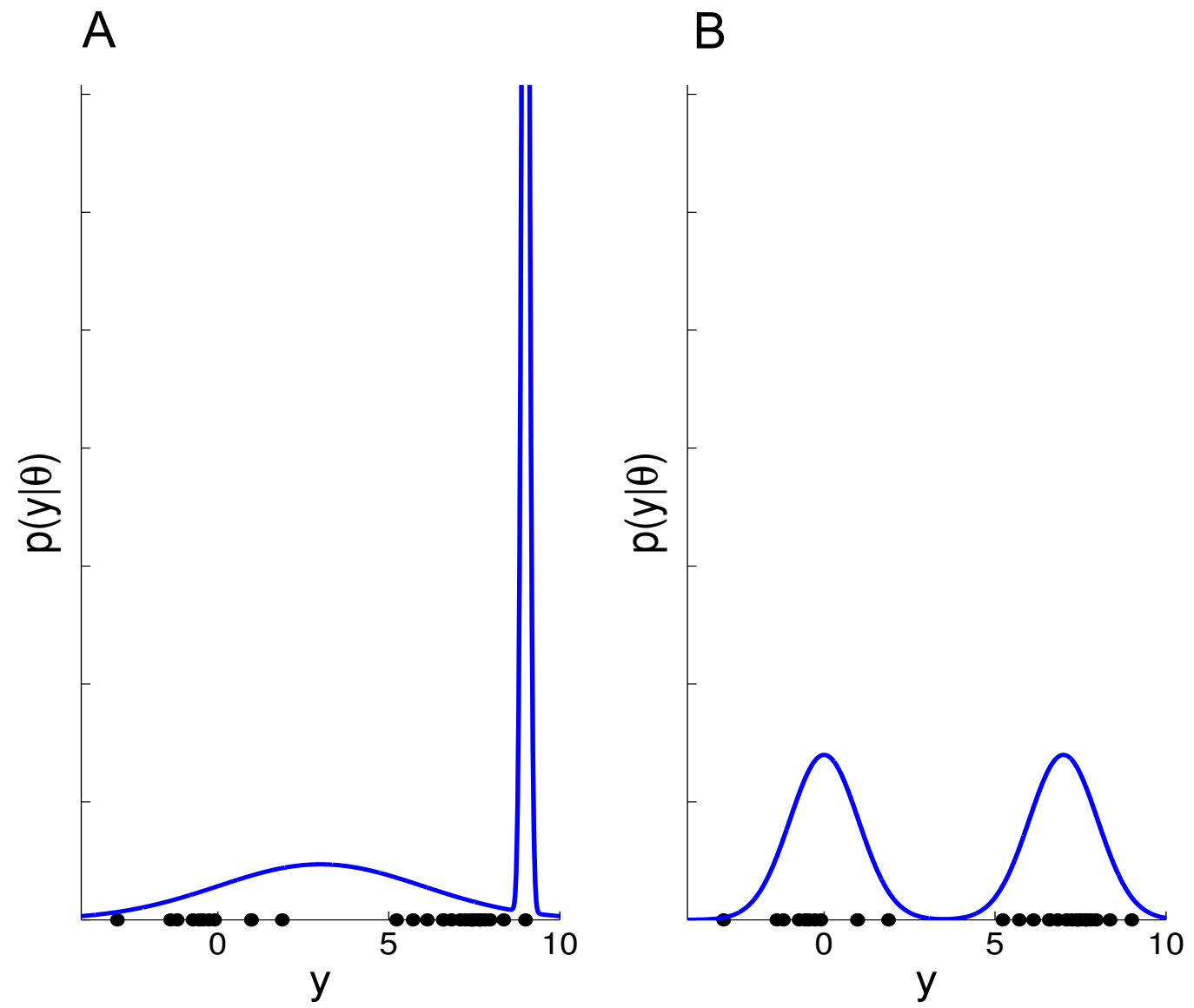
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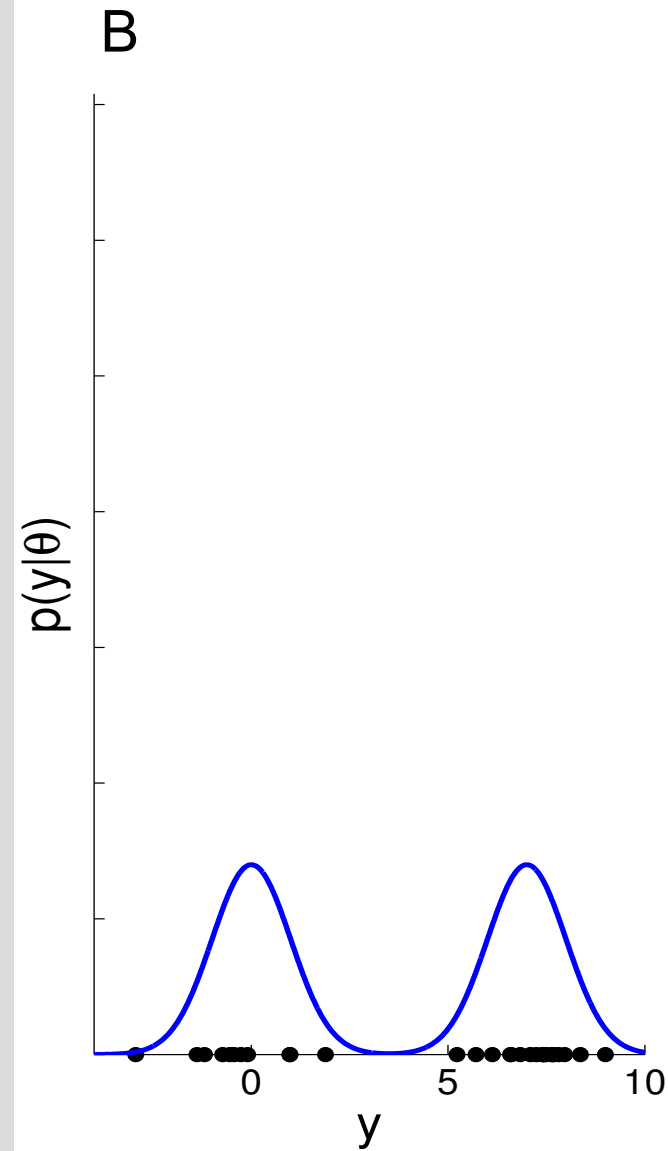
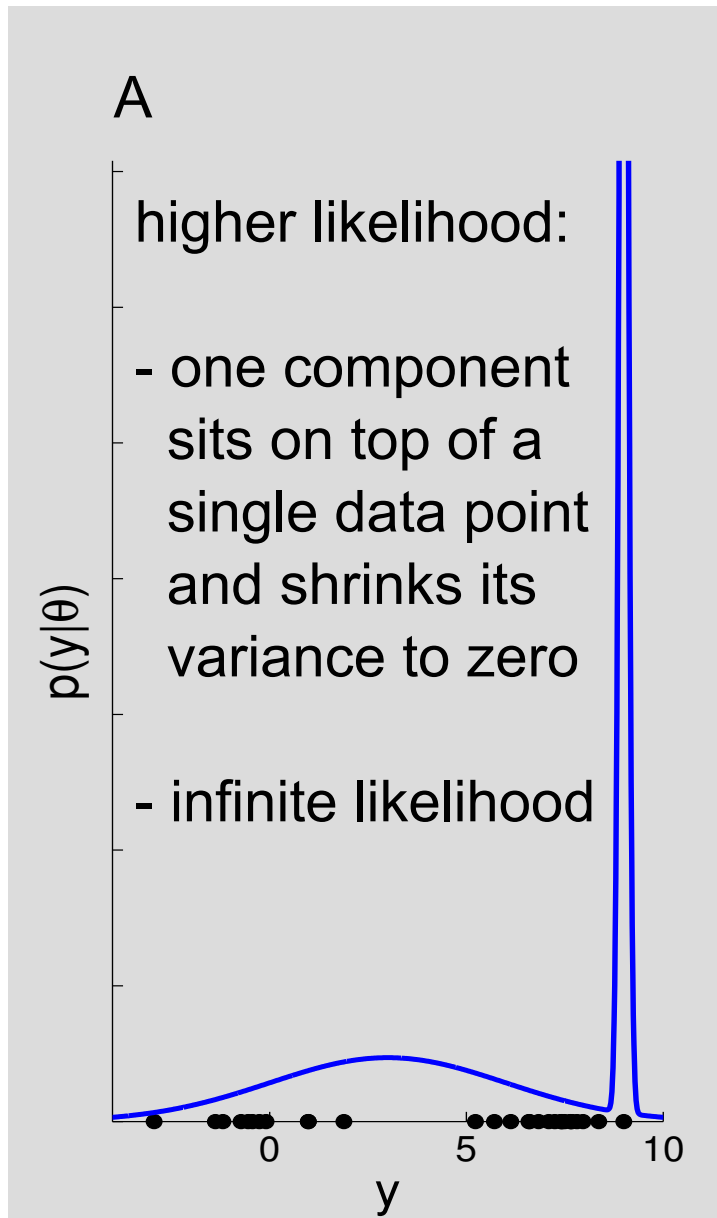
Log-likelihood for the different parameter settings



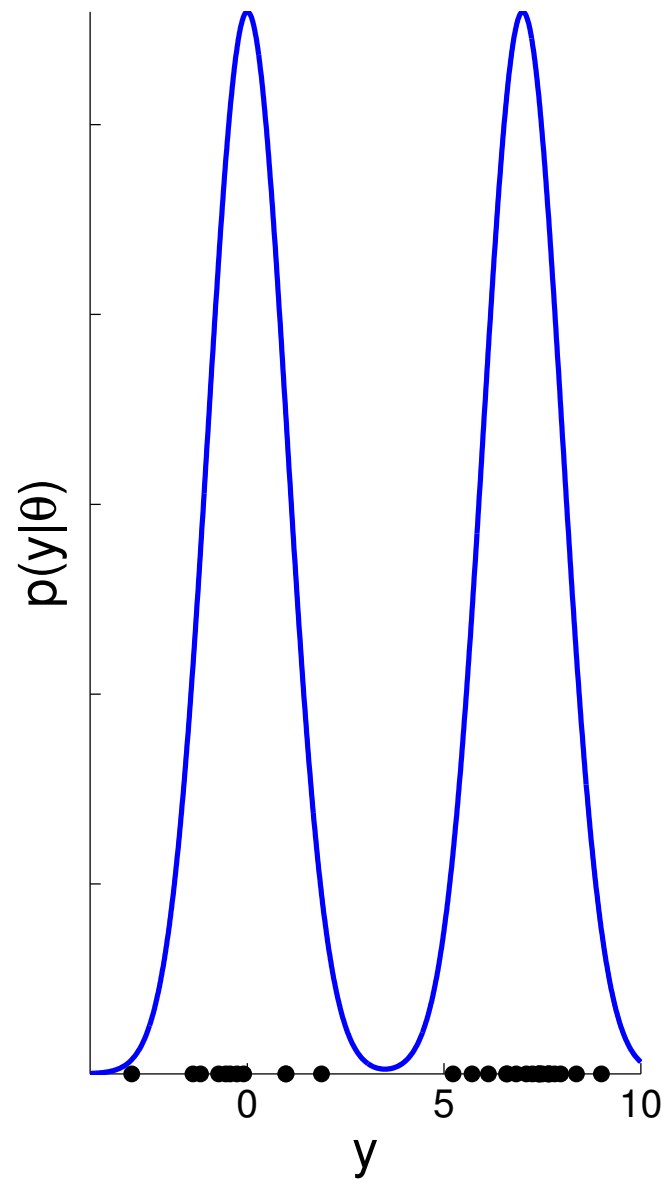
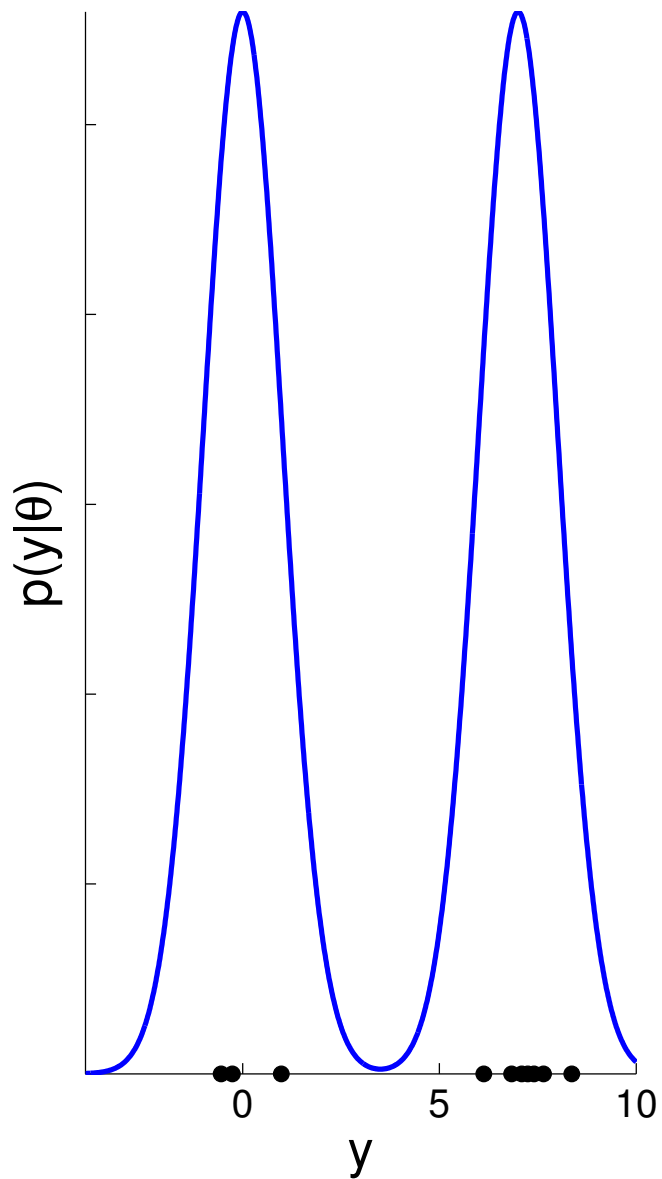
Which parameters have the highest likelihood?



Which parameters have the highest likelihood?



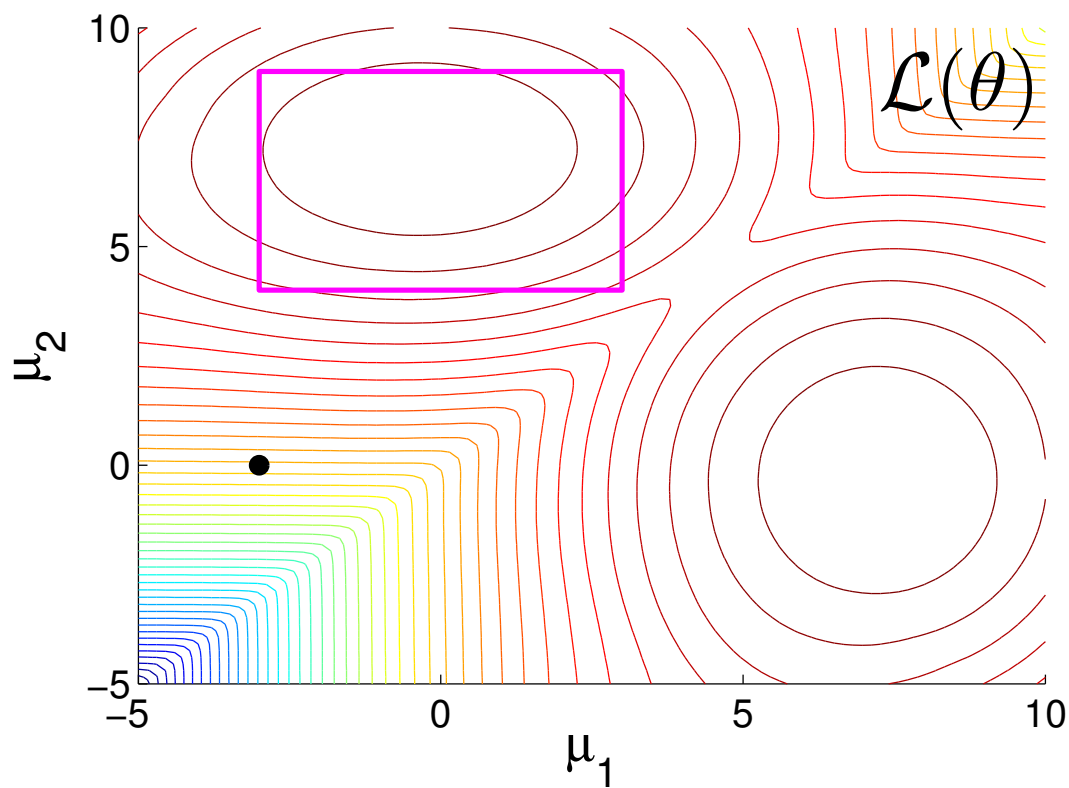
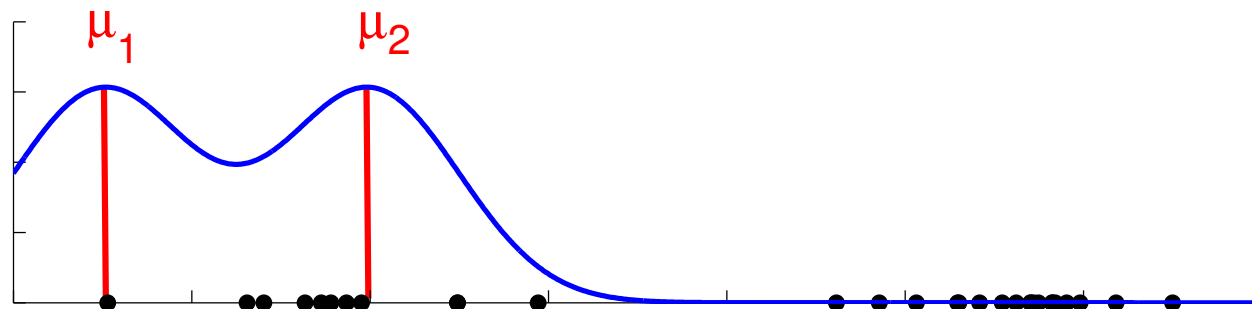
Which dataset leads to estimates with highest confidence?



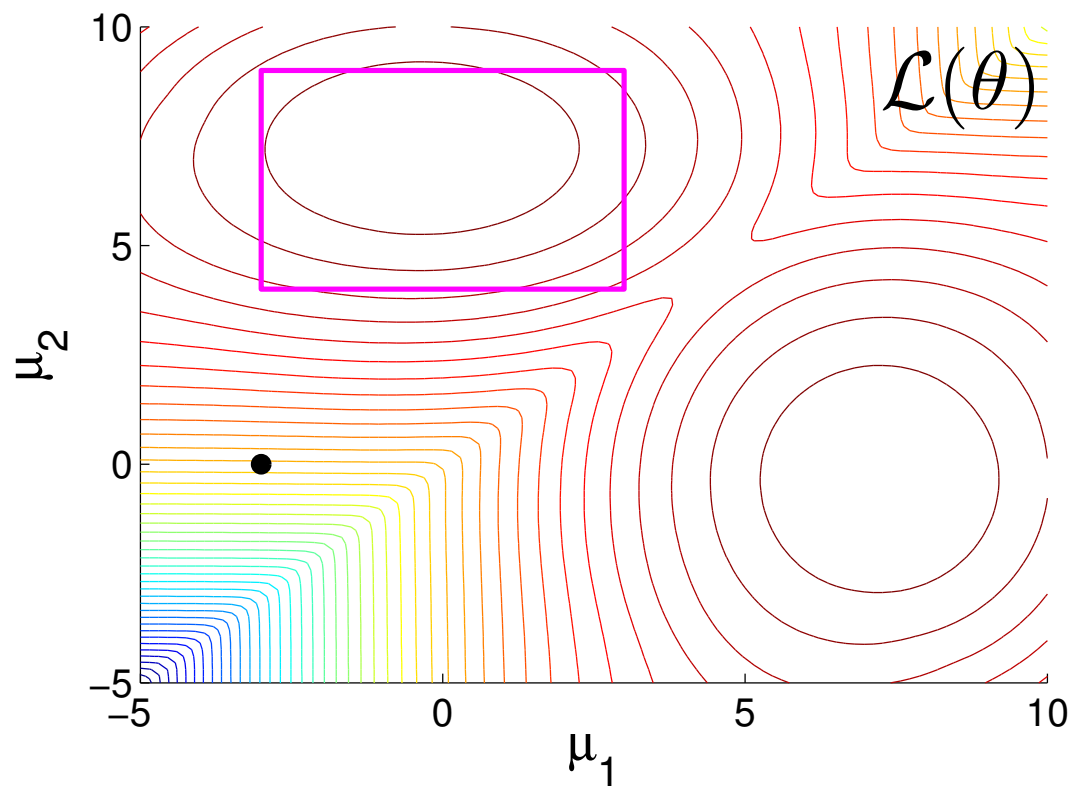
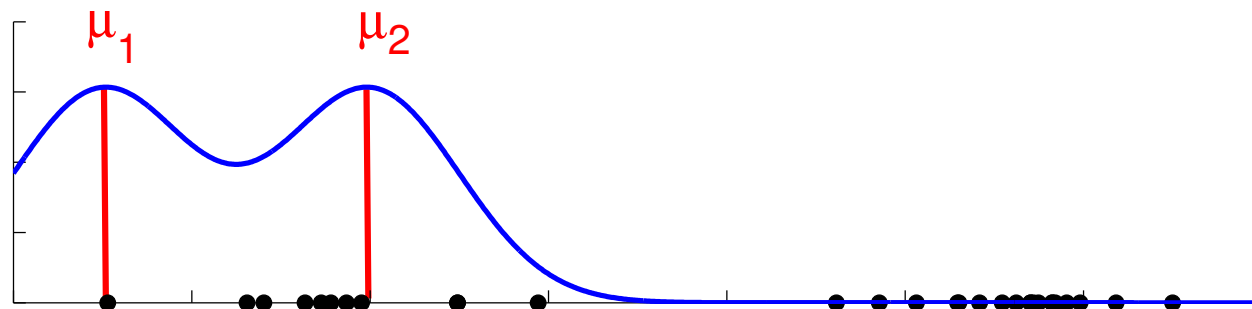
Summary

- Maximum likelihood can be used to estimate parameters from data
- But...
 - costly to grid up parameter space and evaluate likelihood for each parameter setting
 - maximum likelihood can “over fit”
 - unclear how to get back uncertainty estimates
 - unclear how to use it to solve questions like “how many clusters are in the dataset?”

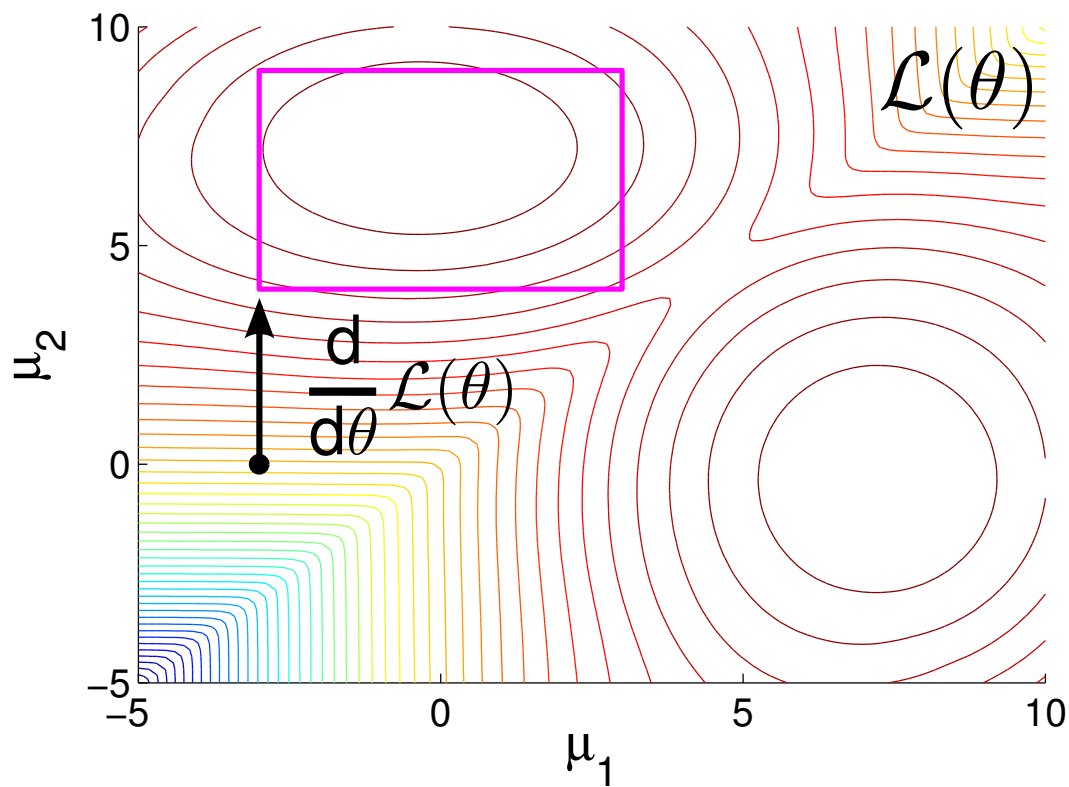
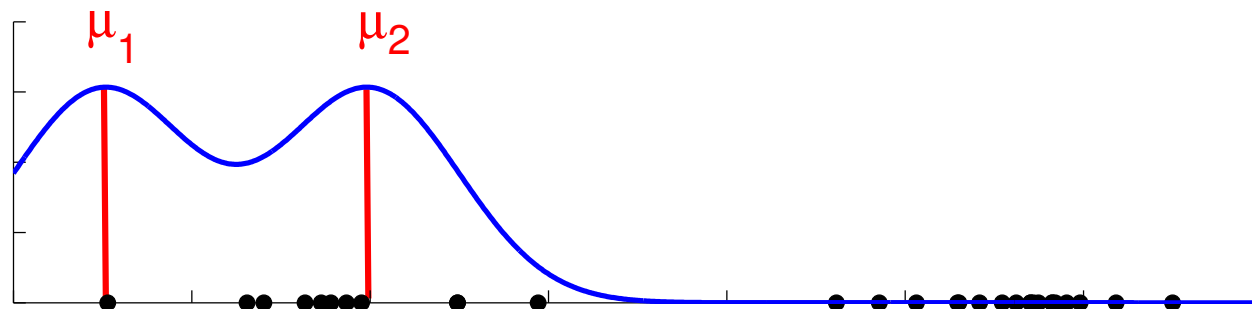
Maximum-likelihood and gradient ascent



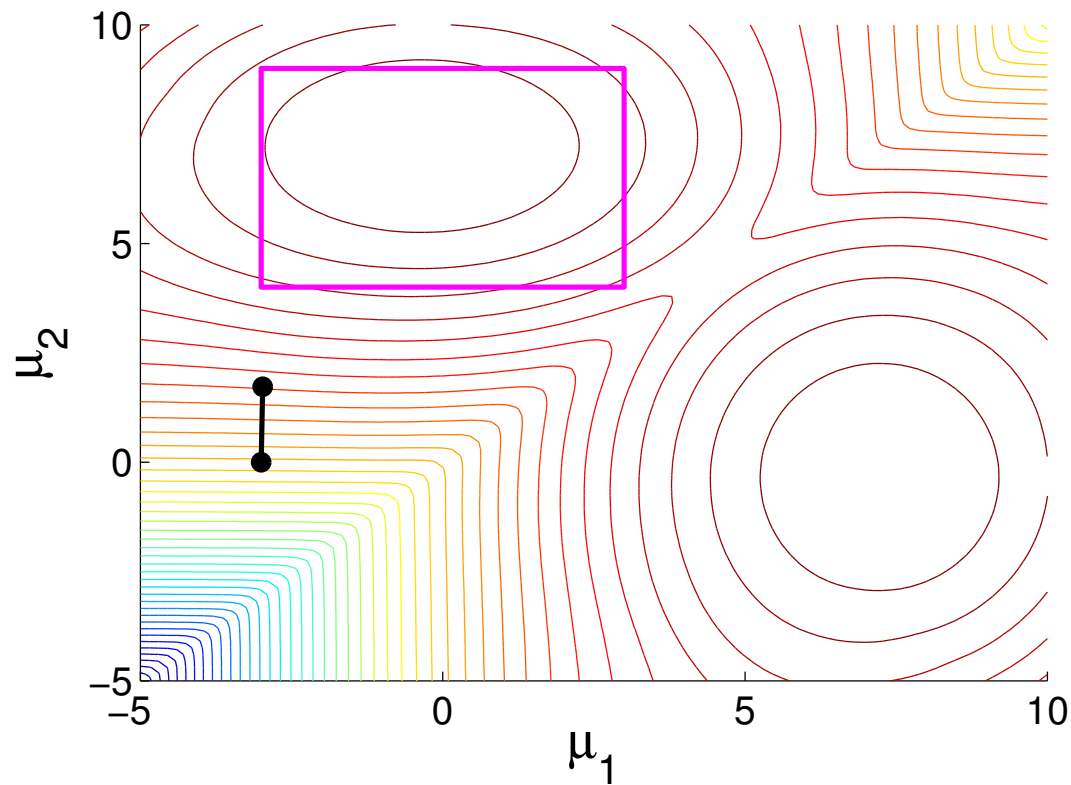
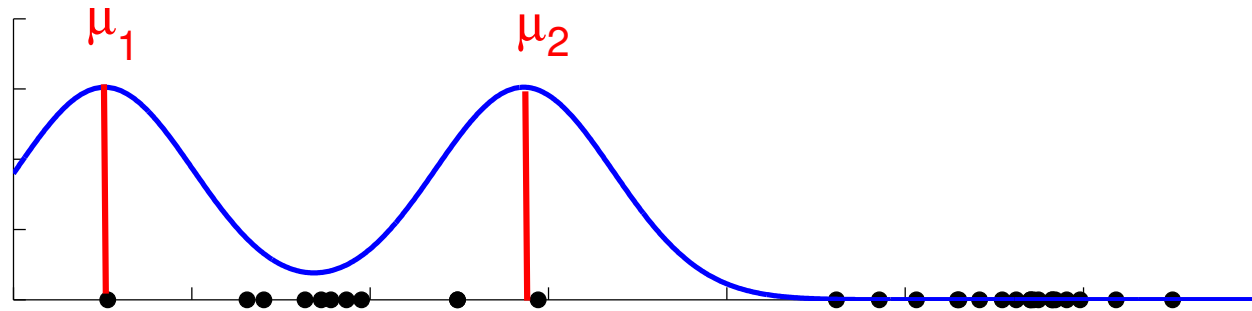
What will happen on the first iteration?



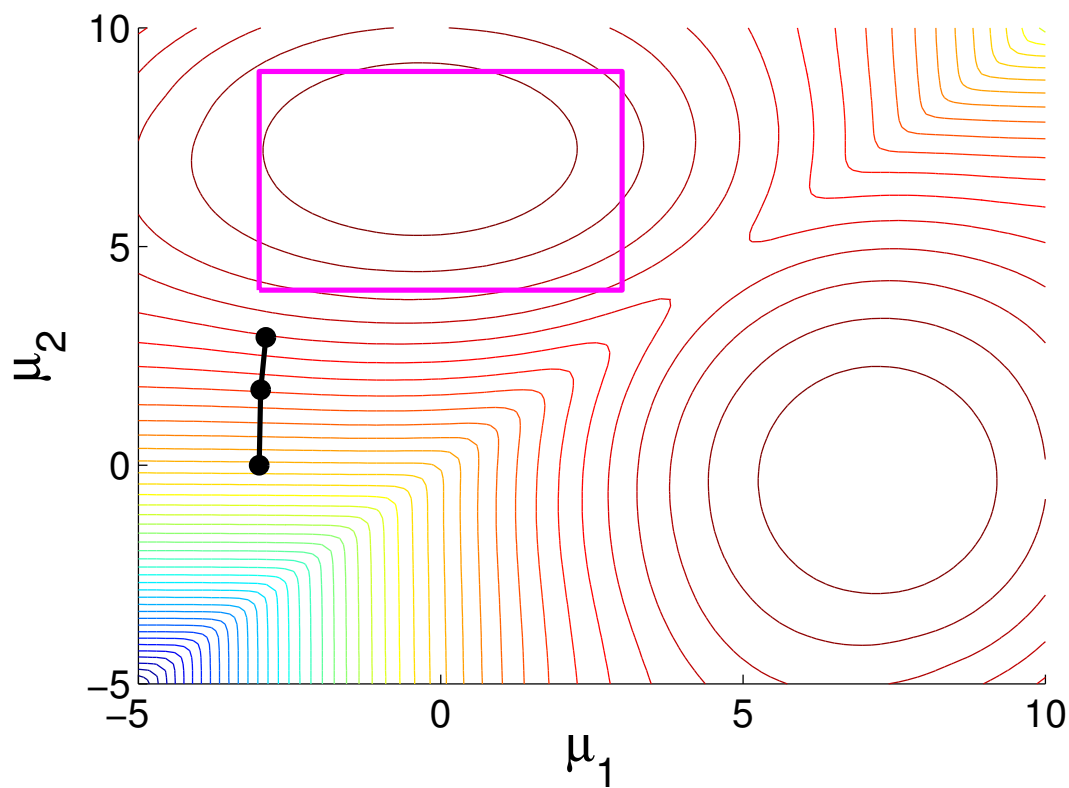
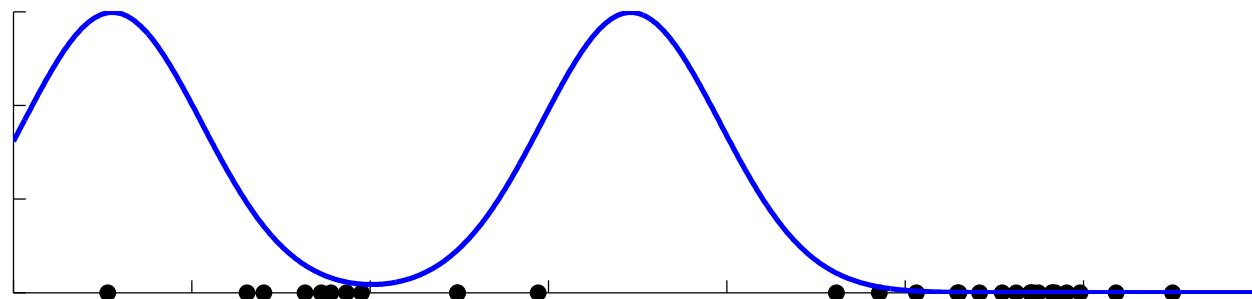
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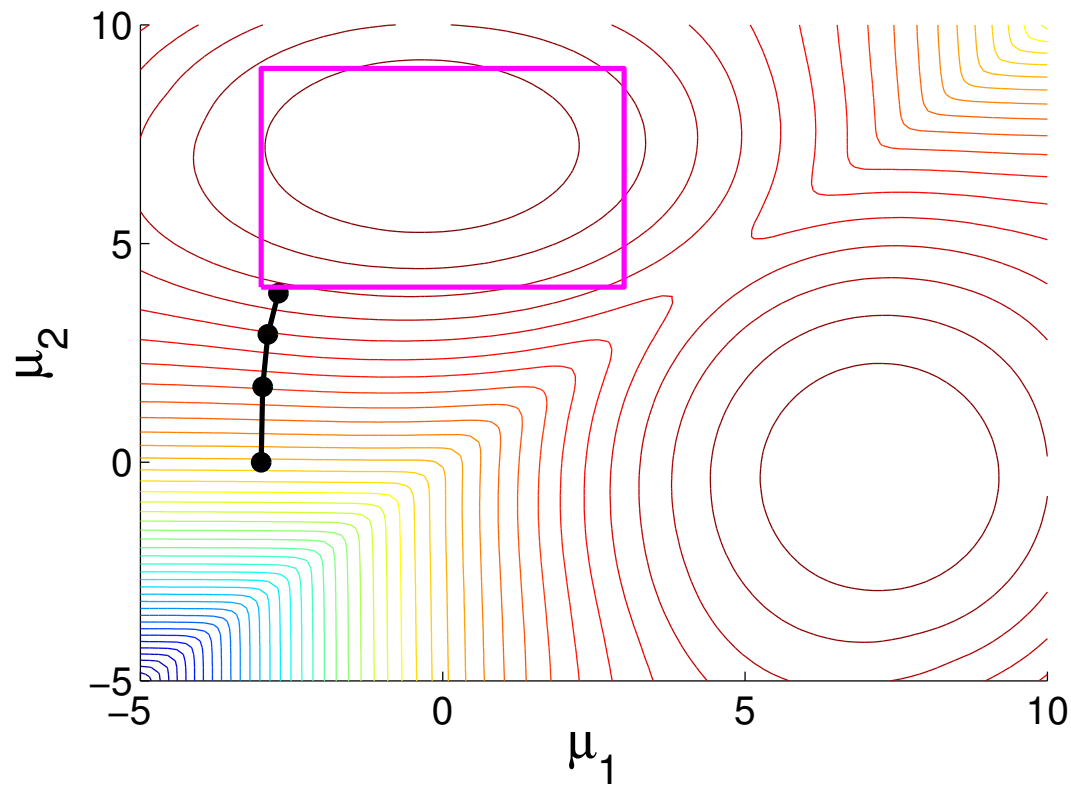
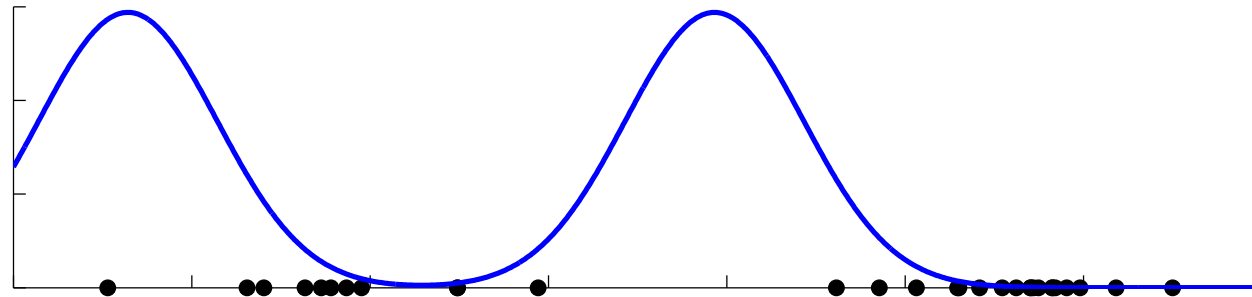
Maximum-likelihood and gradient ascent



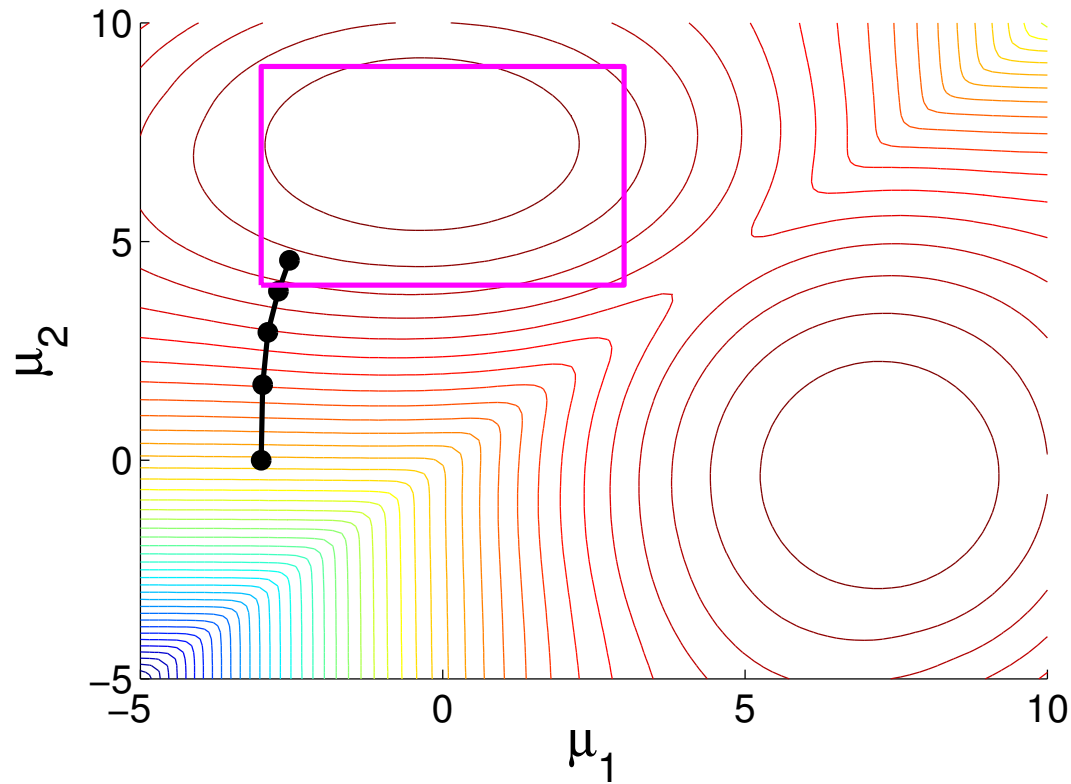
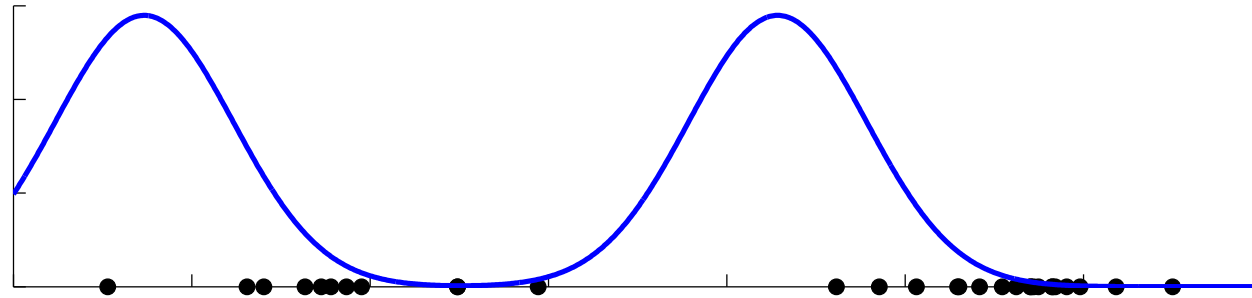
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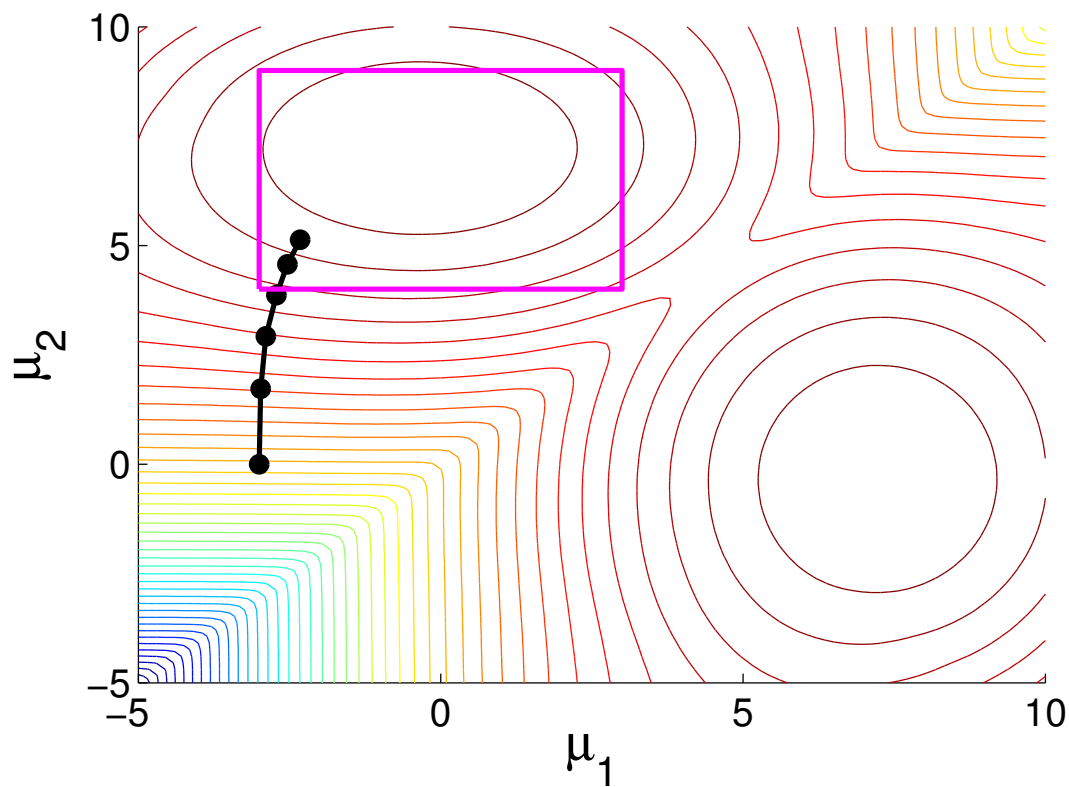
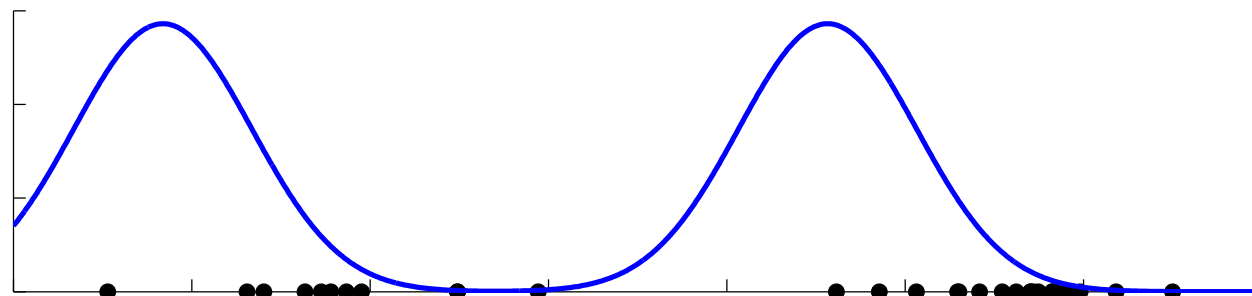
Maximum-likelihood and gradient ascent



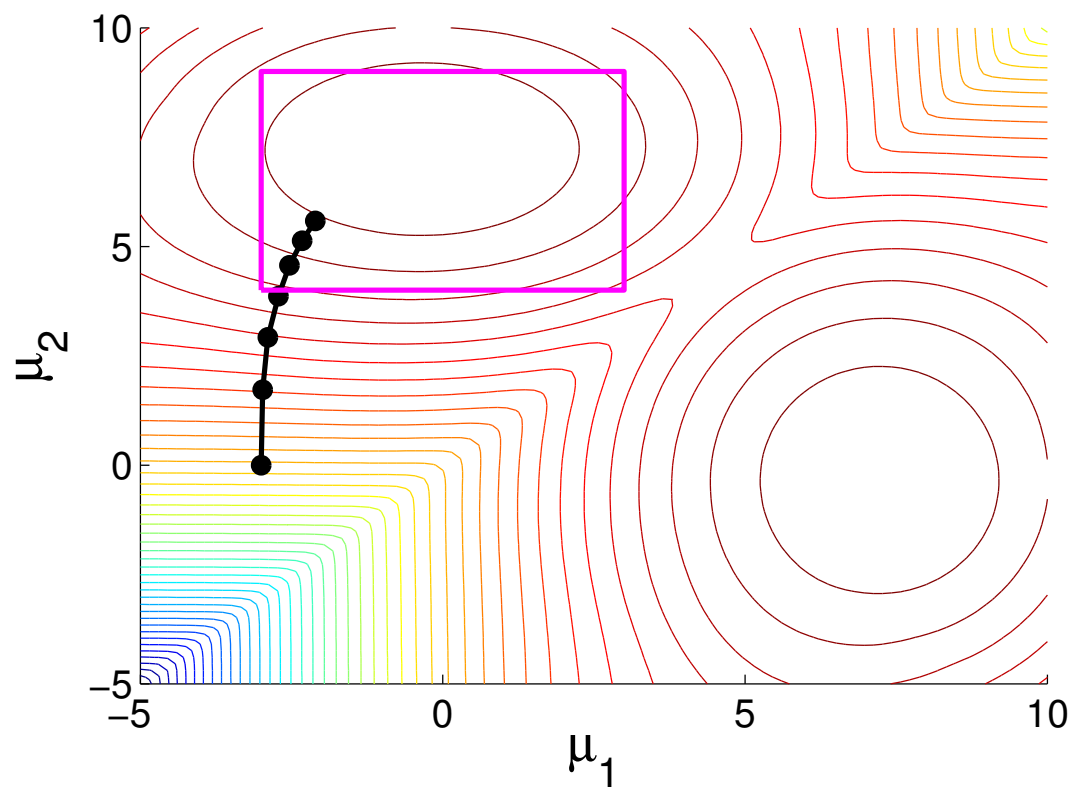
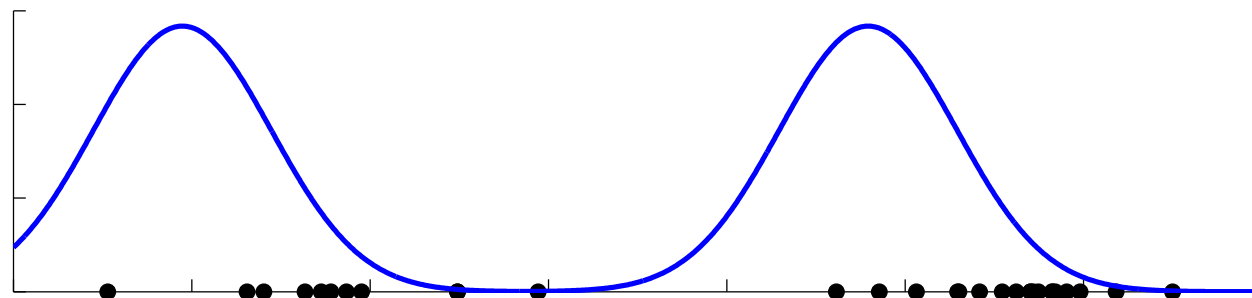
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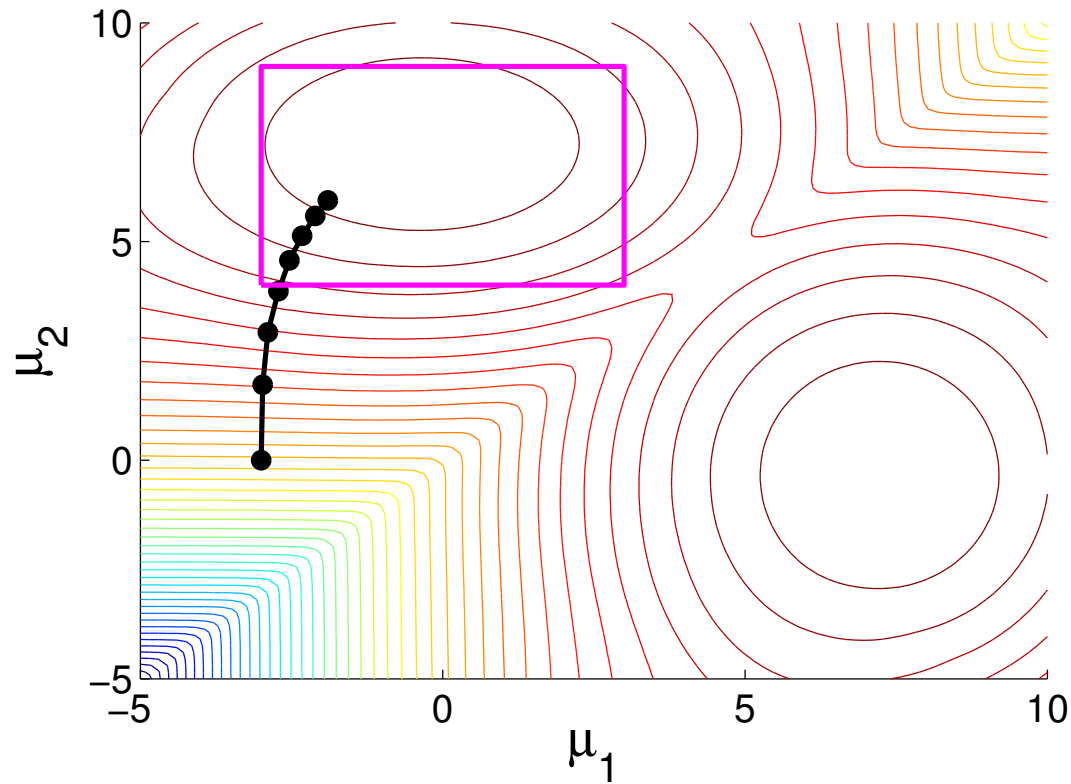
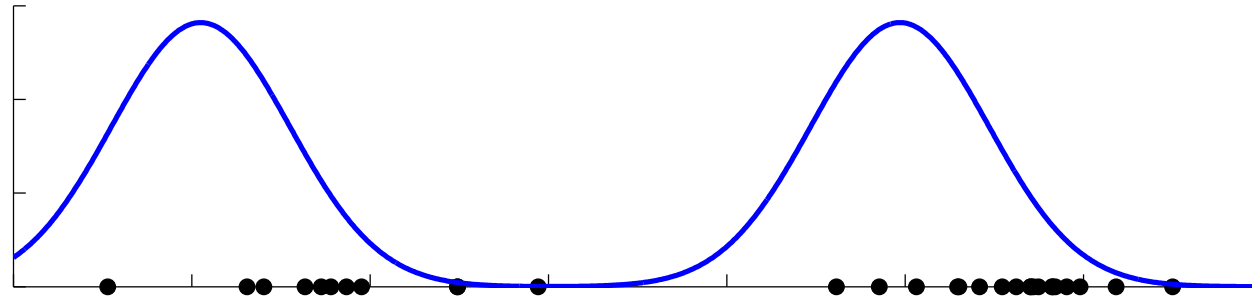
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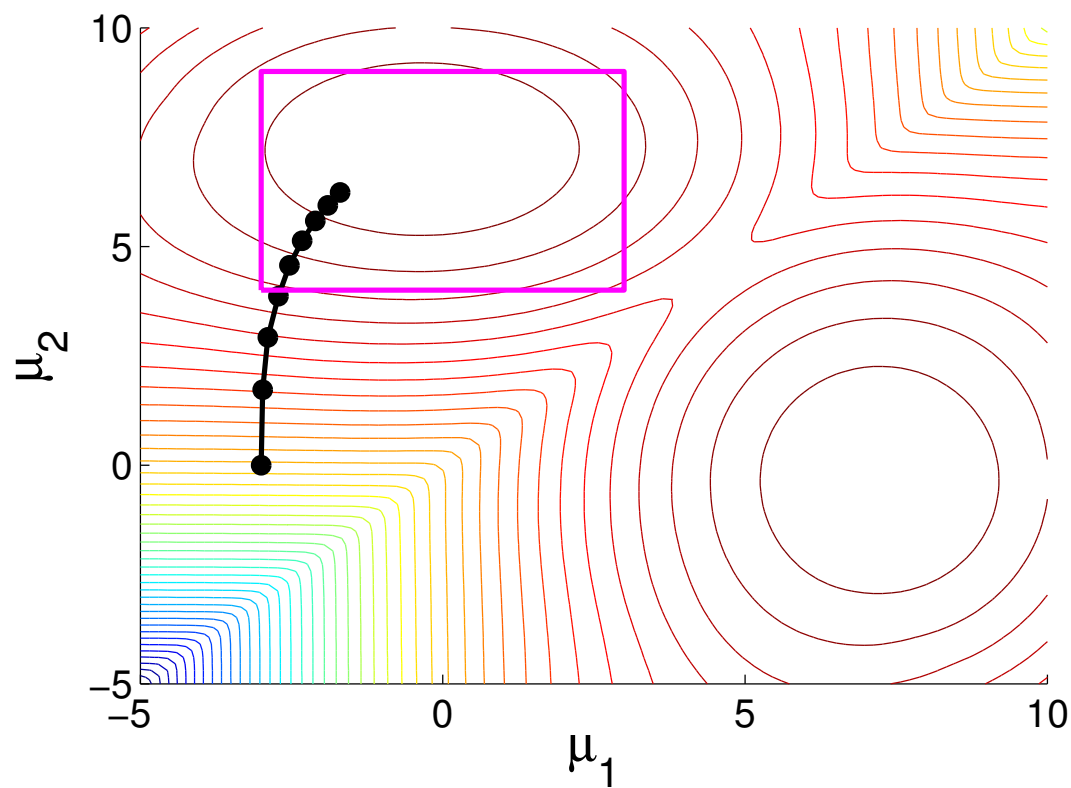
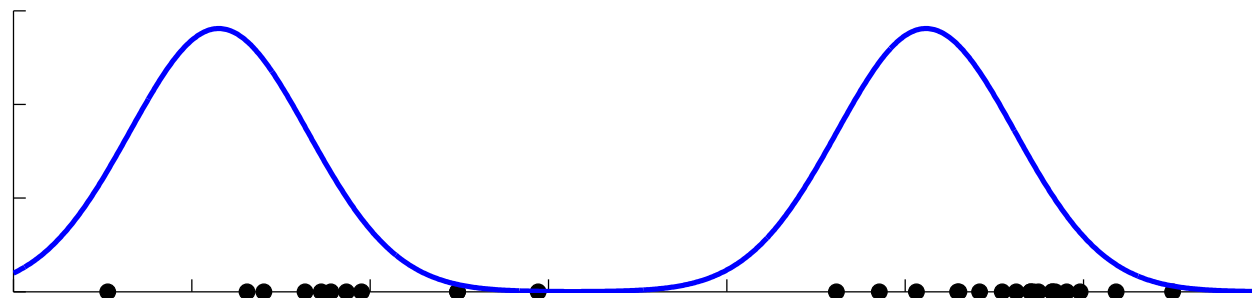
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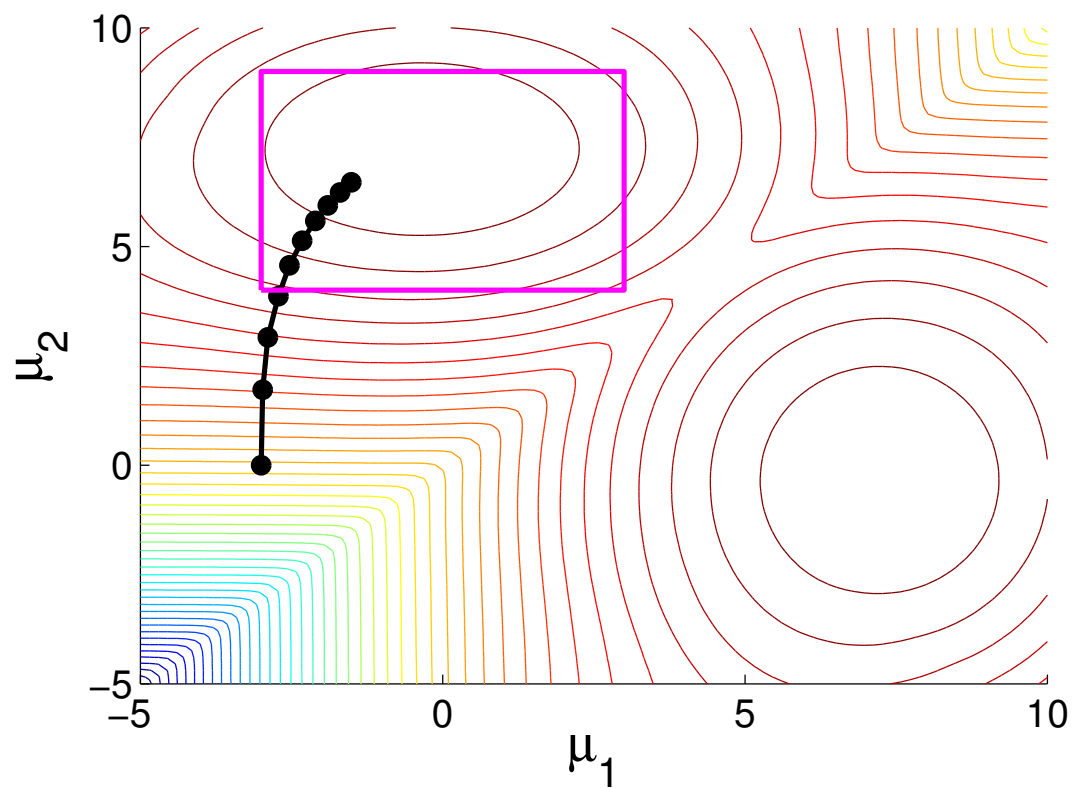
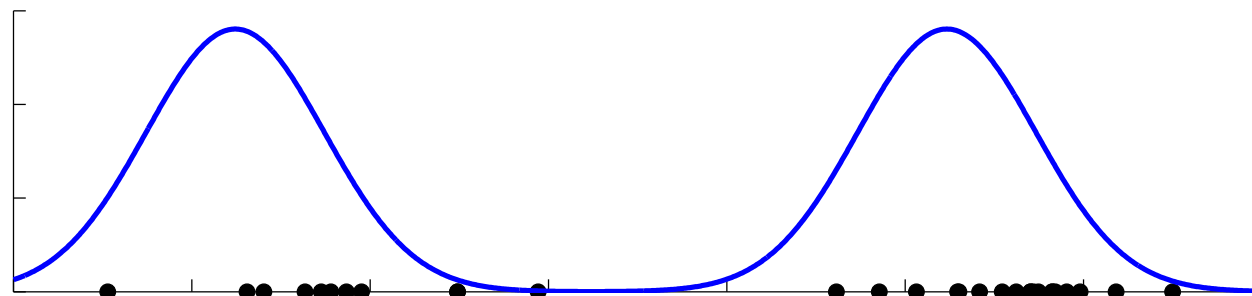
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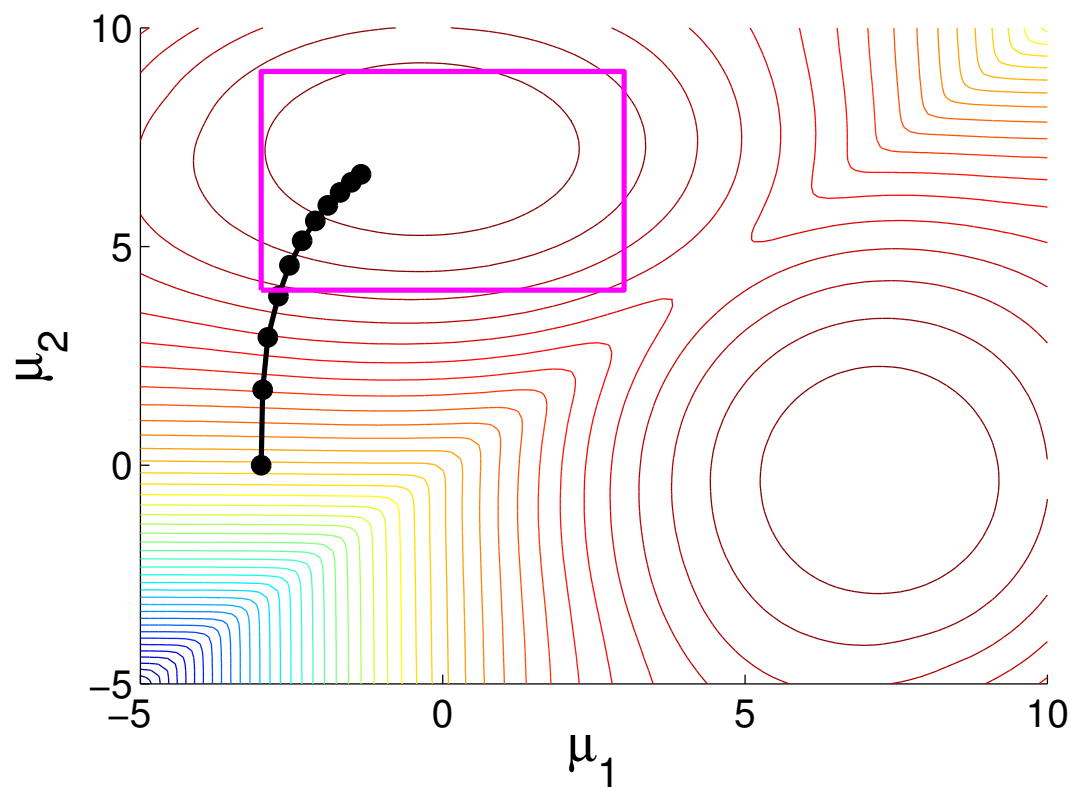
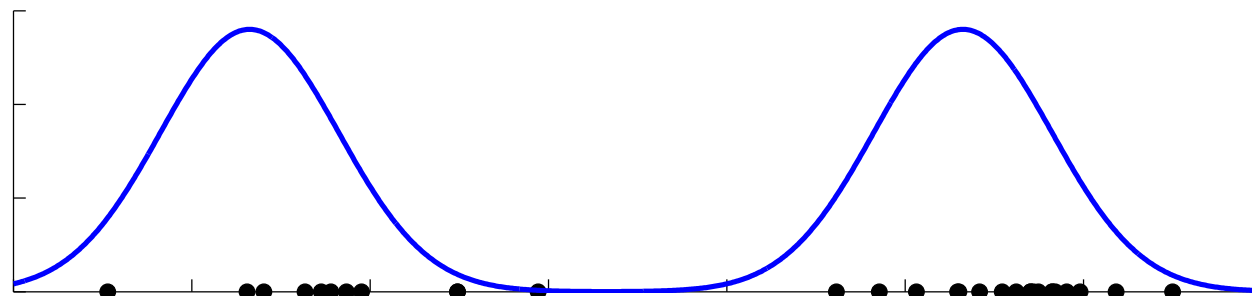
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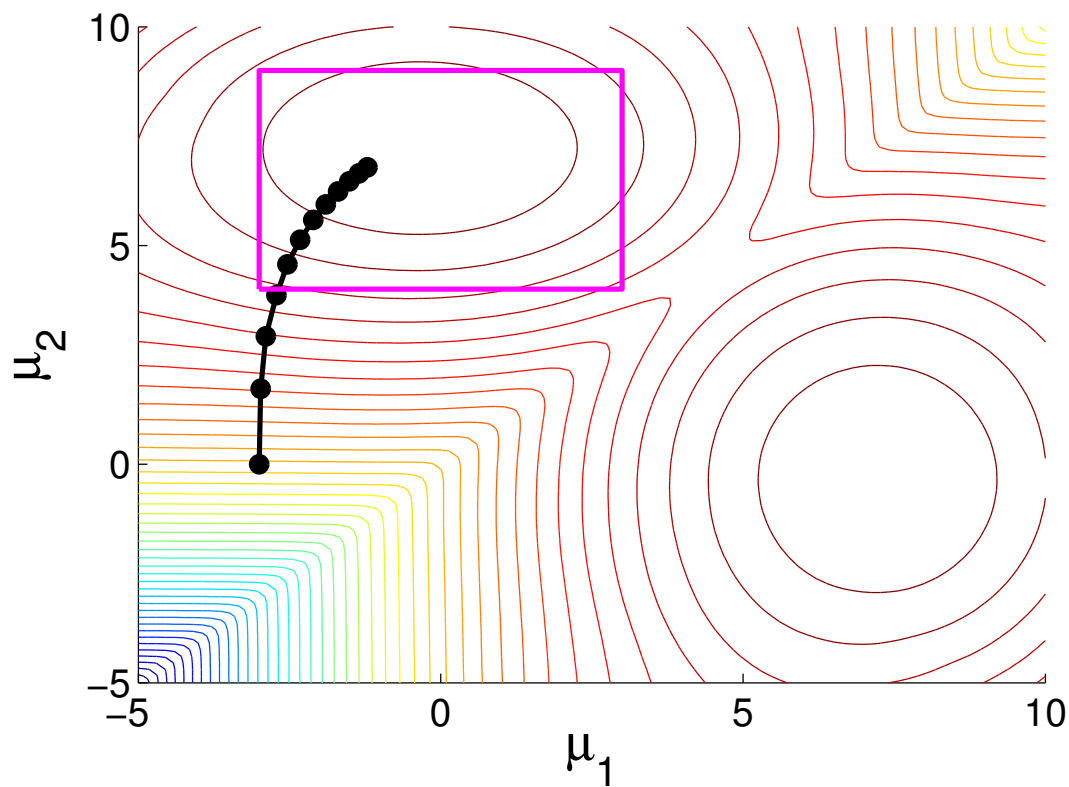
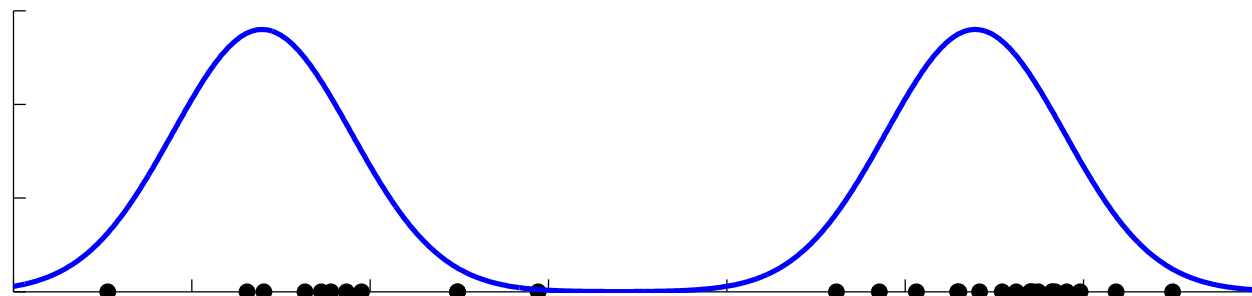
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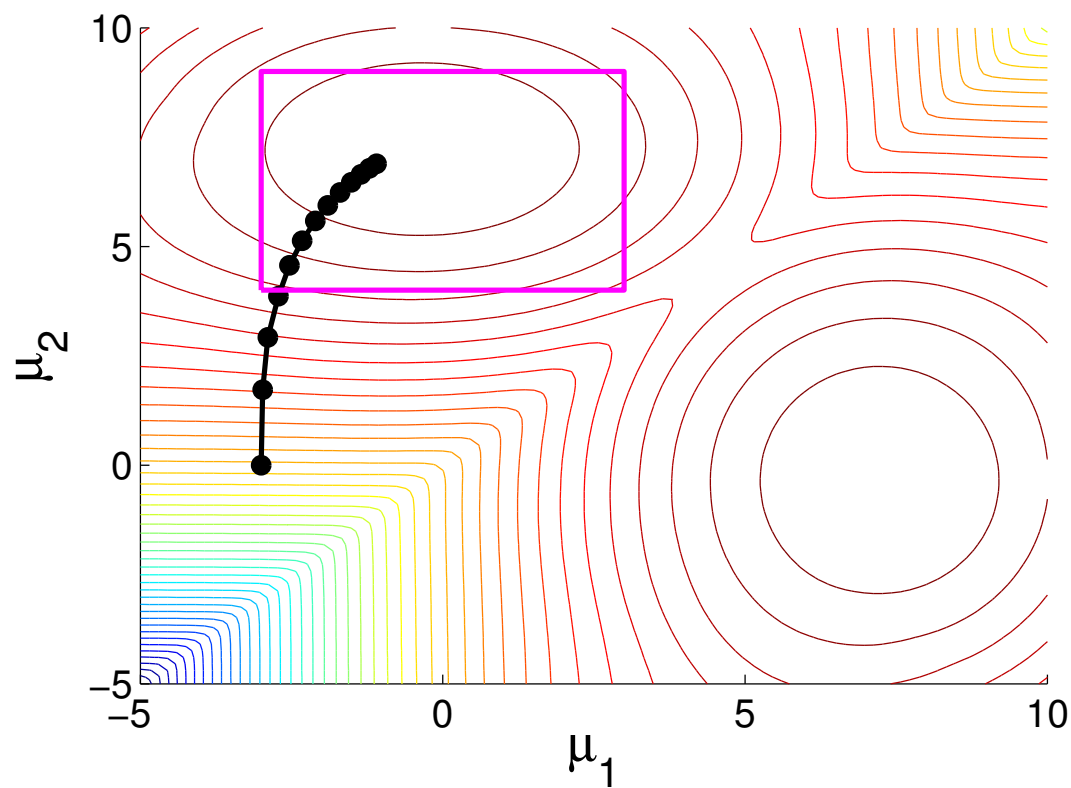
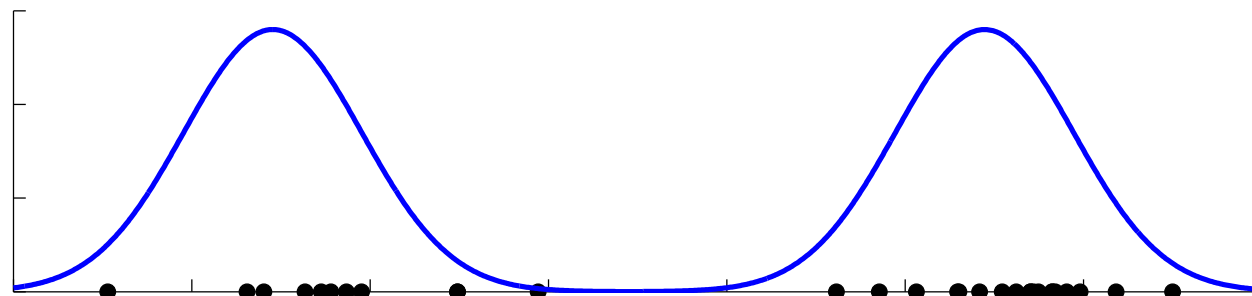
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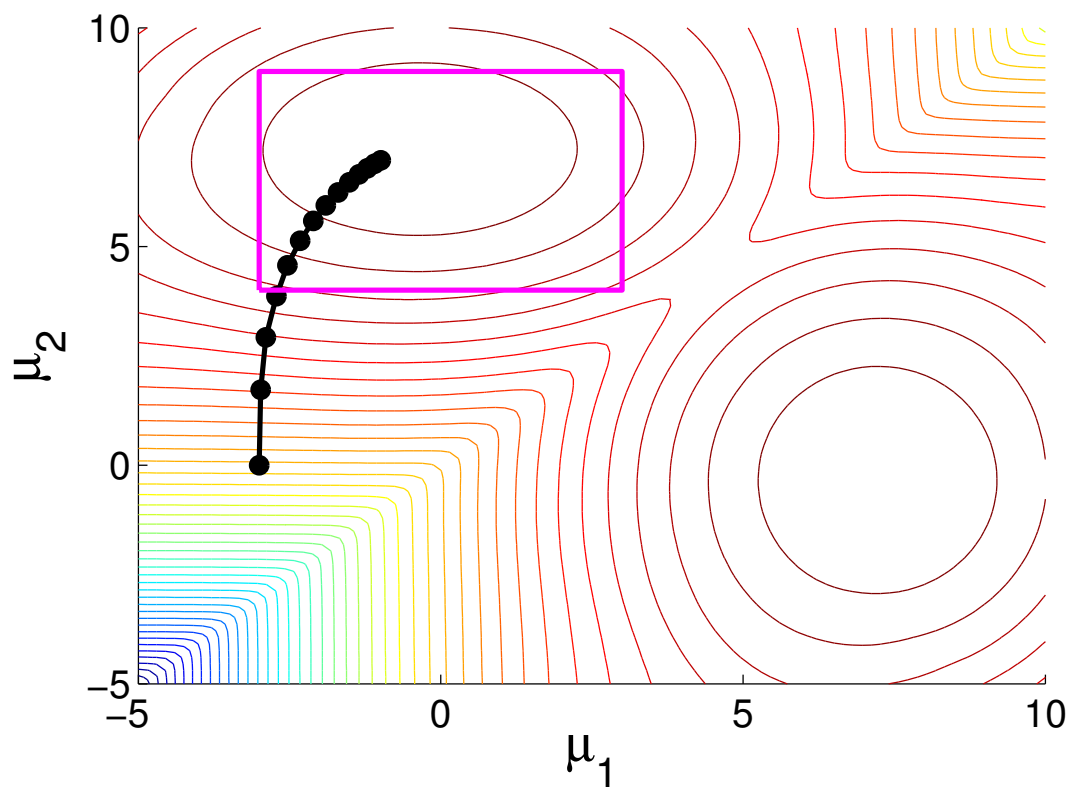
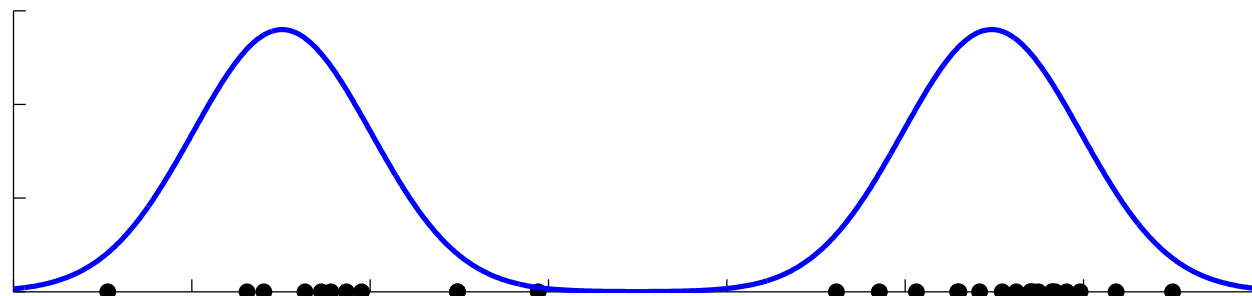
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