

# Probabilistic Modeling and Reasoning

## Homework — 5

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## Problem 1

After renaming the variables as following, the probability we want to calculate is:

Old variable name	New variable name
fuse assembly malfunction	$x_0$
drum unit	$x_1$
toner out	$x_2$
poor paper quality	$x_3$
worn roller	$x_4$
burning smell	$x_5$
poor print quality	$x_6$
wrinkled pages	$x_7$
multiple pages fed	$x_8$
paper jam	$x_9$

Table 1: Mapping of original variable names to new variable names.

$$p(x_1 = 1 \mid x_7 = 0, x_5 = 0, x_3 = 1) = 0.5075$$

It was calculated in python utilizing the "pyAgrum" library in the Python script named *Problem1-EM.py*

## Problem 2

The Python script is named *Gaussian-Mixture.py* and generates the output file named *em\_output.txt*

## Problem 3

The marginal likelihood is determined by summing over  $x_2$ :

$$p(x_1 = i) = \sum_{j=1}^2 \theta_{ij}.$$

1. For the first matrix:

$$\theta^{(1)} = \begin{pmatrix} 0.3 & 0.3 \\ 0.2 & 0.2 \end{pmatrix}$$

we compute:

$$p(x_1 = 1) = \theta_{11} + \theta_{12} = 0.3 + 0.3 = 0.6, \quad p(x_1 = 2) = \theta_{21} + \theta_{22} = 0.2 + 0.2 = 0.4.$$

2. For the second matrix:

$$\theta^{(2)} = \begin{pmatrix} 0.2 & 0.4 \\ 0.4 & 0 \end{pmatrix}$$

we compute:

$$p(x_1 = 1) = \theta_{11} + \theta_{12} = 0.2 + 0.4 = 0.6, \quad p(x_1 = 2) = \theta_{21} + \theta_{22} = 0.4 + 0 = 0.4.$$

Since the marginal probabilities  $p(x_1 = i)$  are identical for both matrices ( $p(x_1 = 1) = 0.6$  and  $p(x_1 = 2) = 0.4$ ), both  $\theta^{(1)}$  and  $\theta^{(2)}$  yield the same marginal likelihood score.

## Problem 4

Output of the "EM" function (implemented in *EM.m*) for input EM(1.9,0.2), where 1.9 is the starting value for  $\theta$  and 0.2 is the starting value for  $q(h = 2)$ .

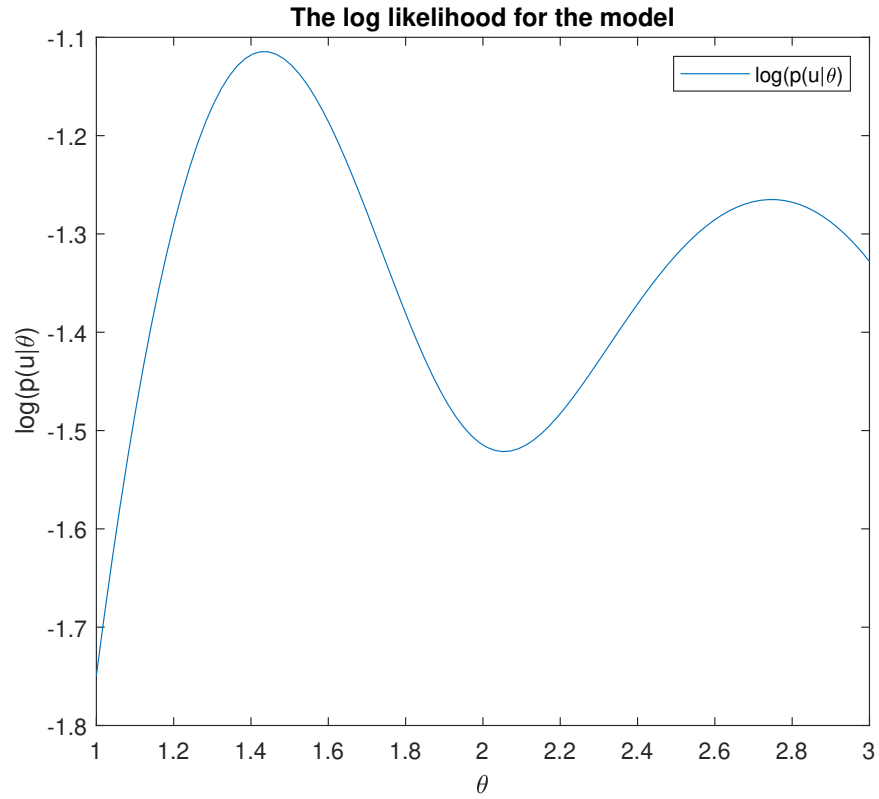


Figure 1: The log likelihood

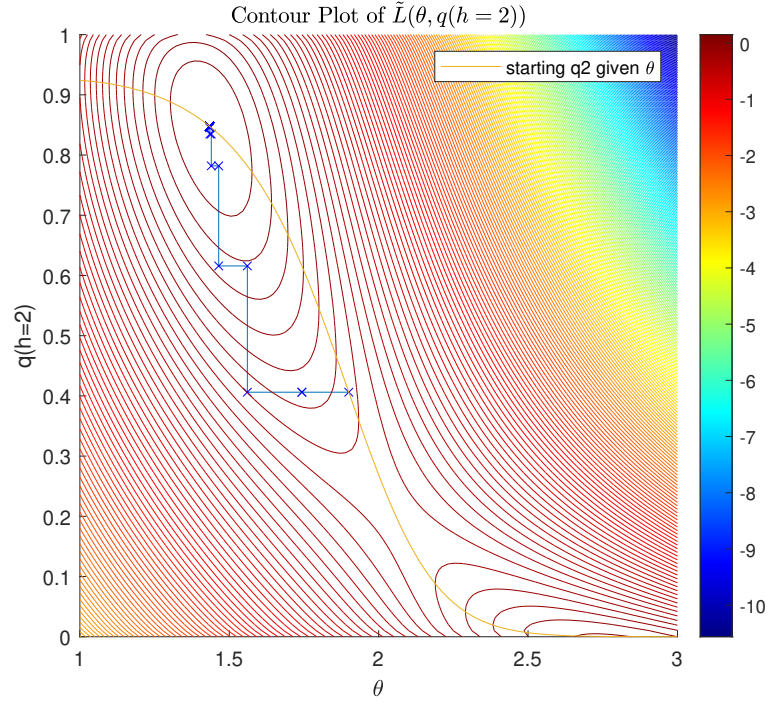


Figure 2: Convergence using 1.9 for the starting value for  $\theta$

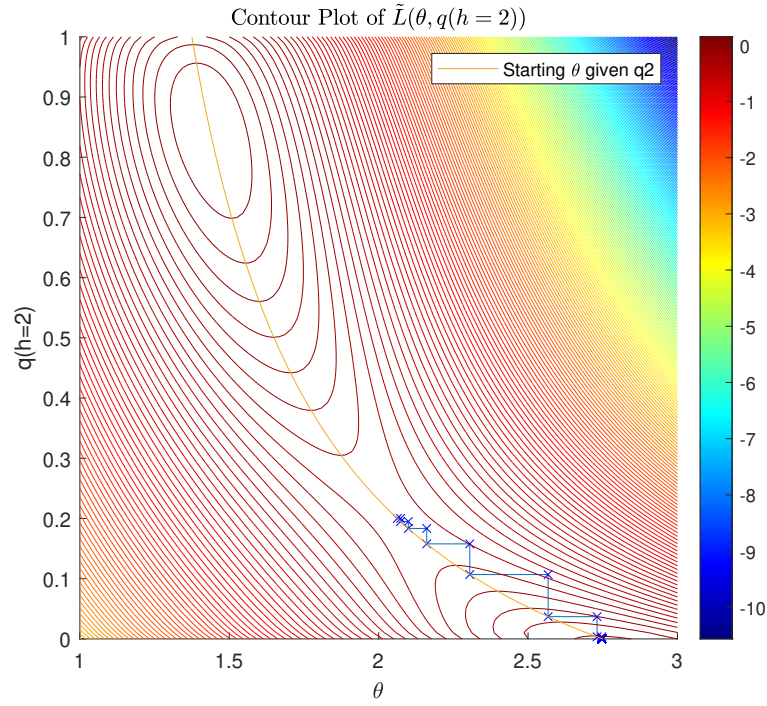


Figure 3: Convergence using 0.2 for the starting value for  $q(h=2)$