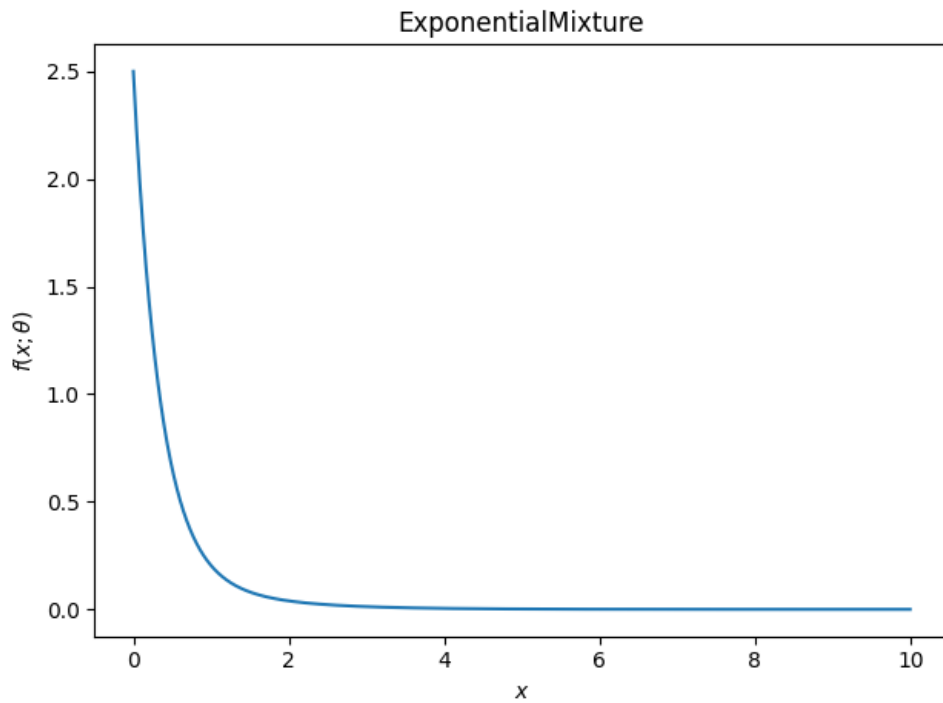


# ECSE 509 Project Report

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a) Pdf of the exponential mixture.



b) E and M update equations.

Let N be the number of observation we see, and J the number of exponential Pdf's we're mixing.

**E step**

$$P_{Y|X}(j|x_i) = \frac{\pi_j \lambda_j e^{-\lambda_j x_i}}{\sum_{j=1}^J \pi_j \lambda_j e^{-\lambda_j x_i}}$$

$$Q(\theta|\theta^t) = \sum_{i=1}^N \sum_{j=1}^J P_{Y|X}(j|x_i) \log[P_{X|Y}(x_i, \theta) \pi_j]$$

$$\boxed{Q(\theta|\theta^t) = \sum_{i=1}^N \sum_{j=1}^J P_{Y|X}(j|x_i) \log[\pi_j \lambda_j e^{-\lambda_j x_i}]}$$

**M step**

$$\theta^{t+1} = \operatorname{argmax} \sum_{i=1}^N \sum_{j=1}^J P_{Y|X}(j|x_i) \log[\pi_j \lambda_j e^{-\lambda_j x_i}]$$

$$\theta^{t+1} = \operatorname{argmax} \sum_{i=1}^N \sum_{j=1}^J P_{Y|X}(j|x_i) (\log[\pi_j \lambda_j] + \log[e^{-\lambda_j x_i}])$$

$$\theta^{t+1} = \operatorname{argmax} \sum_{i=1}^N \sum_{j=1}^J P_{Y|X}(j|x_i) (\log[\pi_j \lambda_j] + [-\lambda_j x_i])$$

$$\theta^{t+1} = \operatorname{argmin} - \sum_{i=1}^N \sum_{j=1}^J P_{Y|X}(j|x_i) (\log[\pi_j \lambda_j] + [-\lambda_j x_i])$$

$$\boxed{\theta^{t+1} = \operatorname{argmin} \sum_{i=1}^N \sum_{j=1}^J P_{Y|X}(j|x_i) (\lambda_j x_i - \log[\pi_j \lambda_j])}$$

We have the constraint that  $\sum_{j=1}^J \pi_j = 1$ , therefor to enforce the condition, we have to write the following Lagrangian:

$$L = \sum_{i=1}^N \sum_{j=1}^J P_{Y|X}(j|x_i) (\lambda_j x_i - \log[\pi_j \lambda_j]) + C(\sum_{j=1}^J \pi_j - 1)$$

The update rule of our parameters are then found by:

$$\begin{aligned} \frac{dL}{d\lambda_j} &= 0 \\ \frac{dL}{d\pi_j} &= 0 \end{aligned}$$

By solving the derivative with respect to both our terms, we get that the update rule is then:

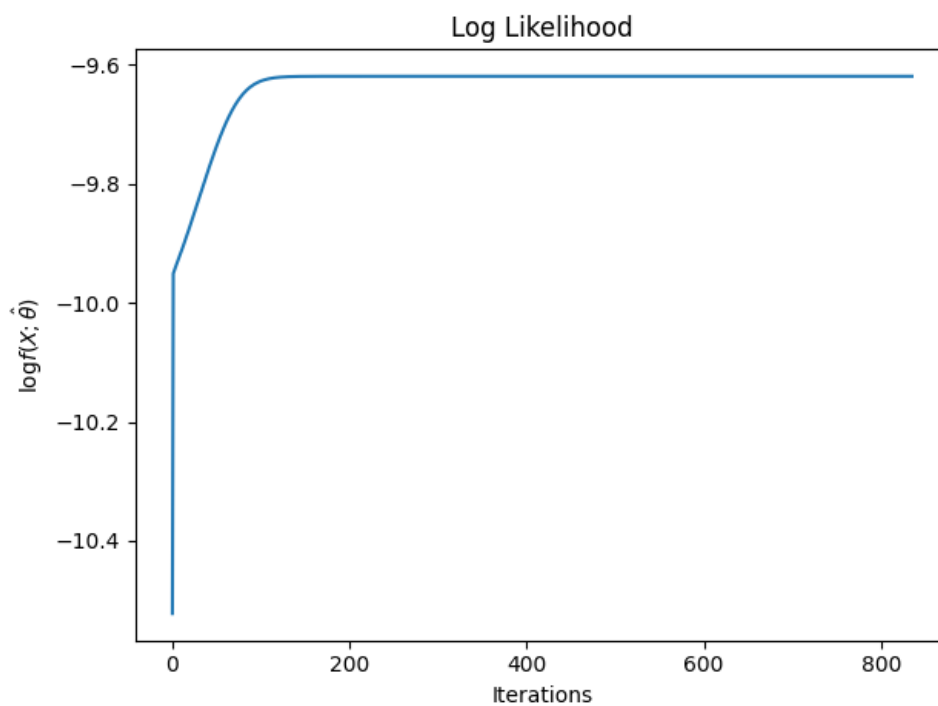
$$\boxed{\lambda_j = \frac{\sum_{i=1}^N P_{Y|X}(j|x_i)}{\sum_{i=1}^N P_{Y|X}(j|x_i) x_i}}$$

$$\boxed{\pi_j = \frac{\sum_{i=1}^N P_{Y|X}(j|x_i)}{\sum_{i=1}^N \sum_{j=1}^J P_{Y|X}(j|x_i)}}$$

**c) EM Algorithm.**

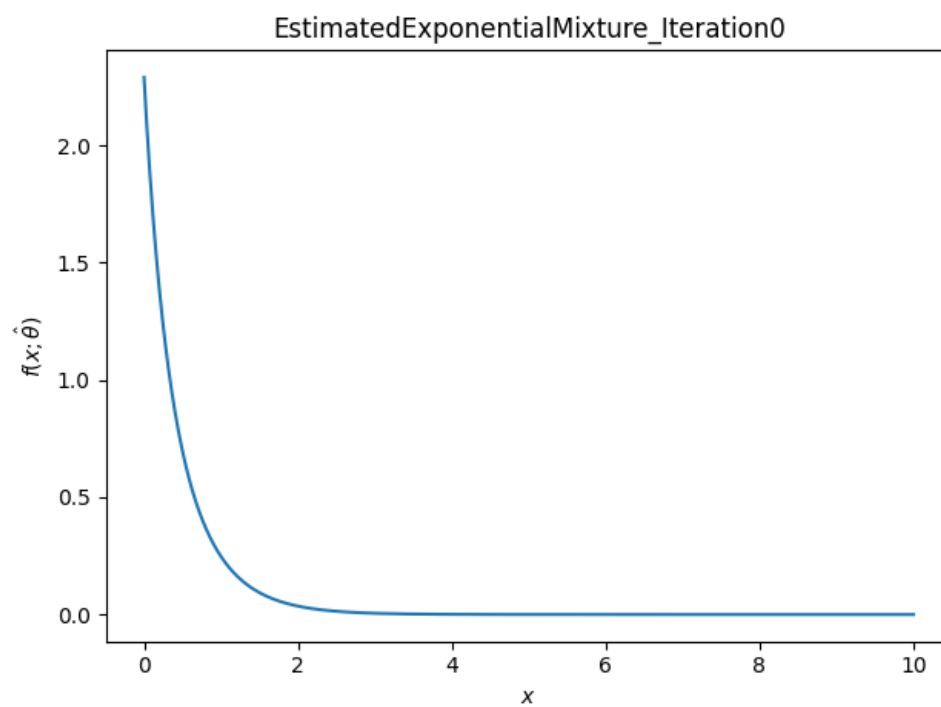
Please refer to the PDF containing the code uploaded separately on crowdmark.

**d) Log Likelihood.**



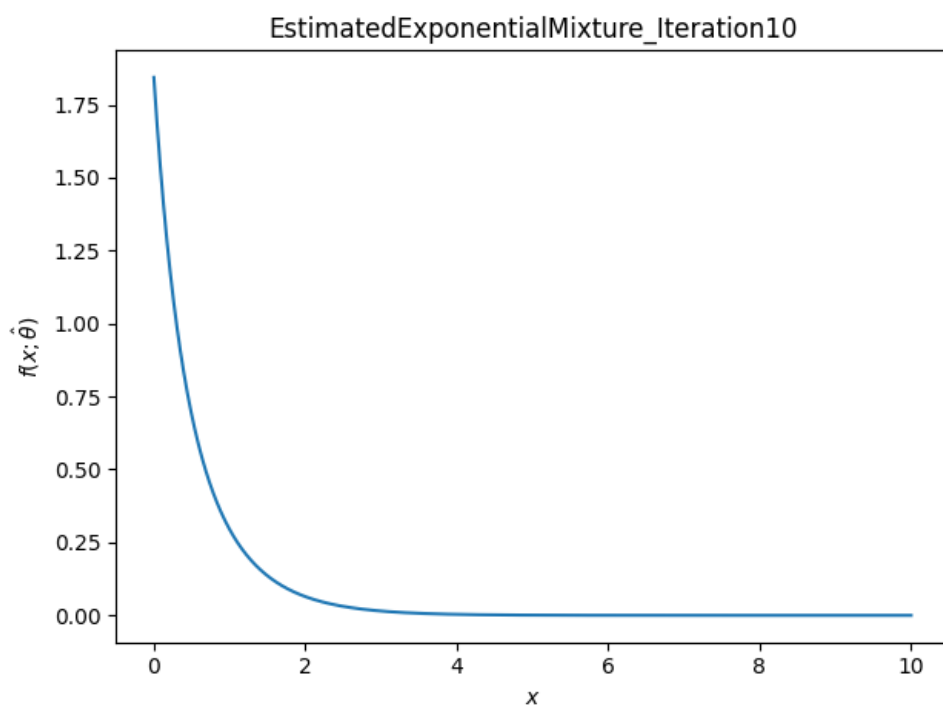
From the graph above, we can see that the log likelihood is monotonically increasing.

e) Estimated exponential mixture pdf.



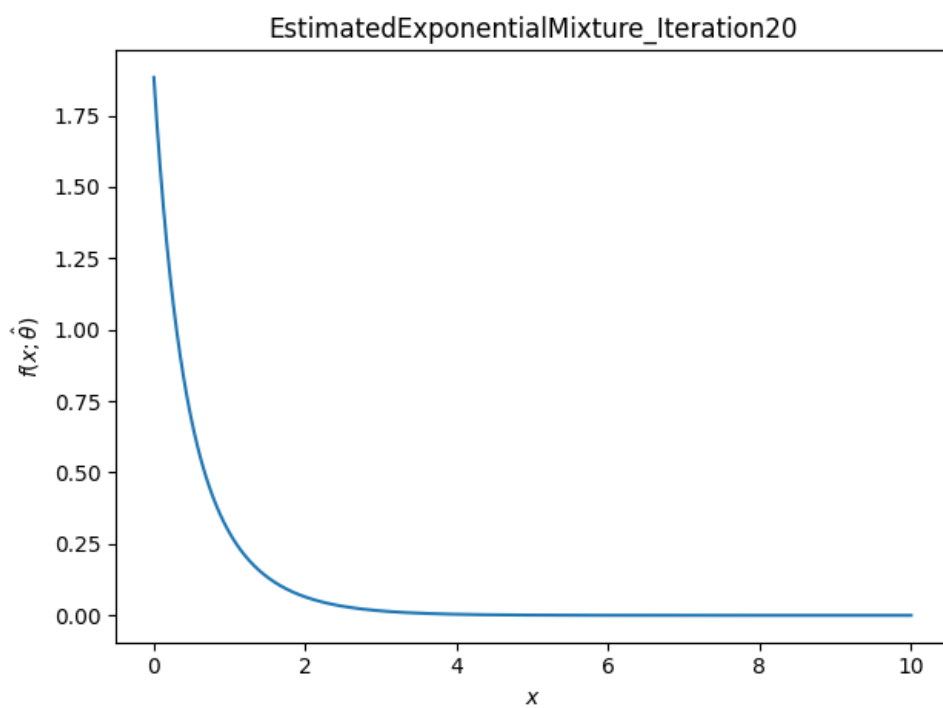
$\hat{\theta}$  estimate at iteration 0:

- $\hat{\lambda}_1 = 1.896595$
- $\hat{\lambda}_2 = 4.295177$
- $\hat{\pi}_1 = 0.8357424$
- $\hat{\pi}_2 = 0.1642576$



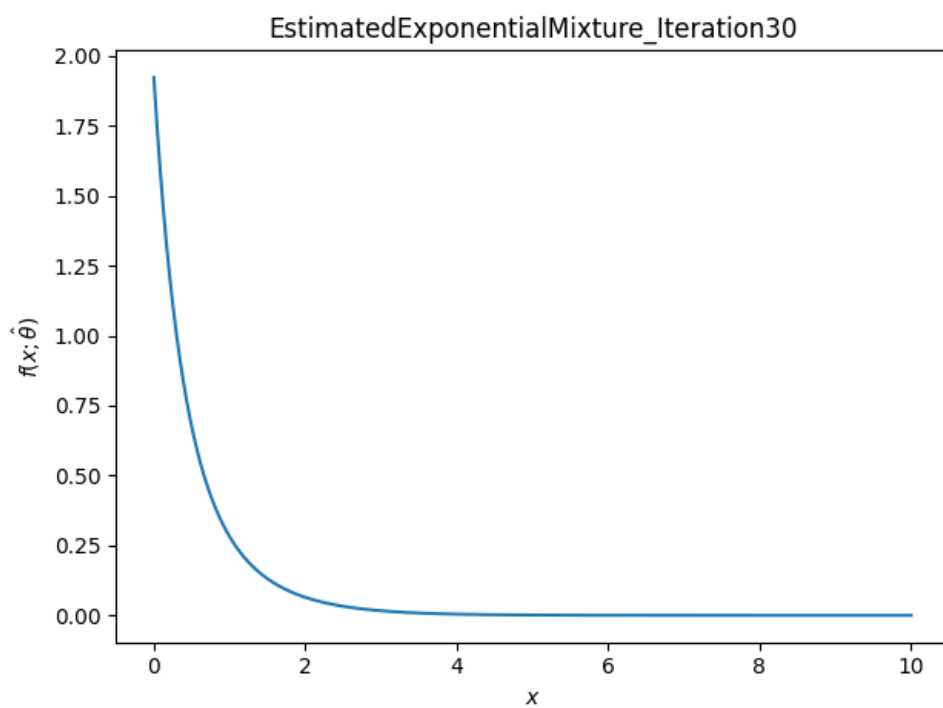
$\hat{\theta}$  estimate at iteration 10:

- $\hat{\lambda}_1 = 1.4575794$
- $\hat{\lambda}_2 = 3.436903$
- $\hat{\pi}_1 = 0.8049060$
- $\hat{\pi}_2 = 0.1950930$



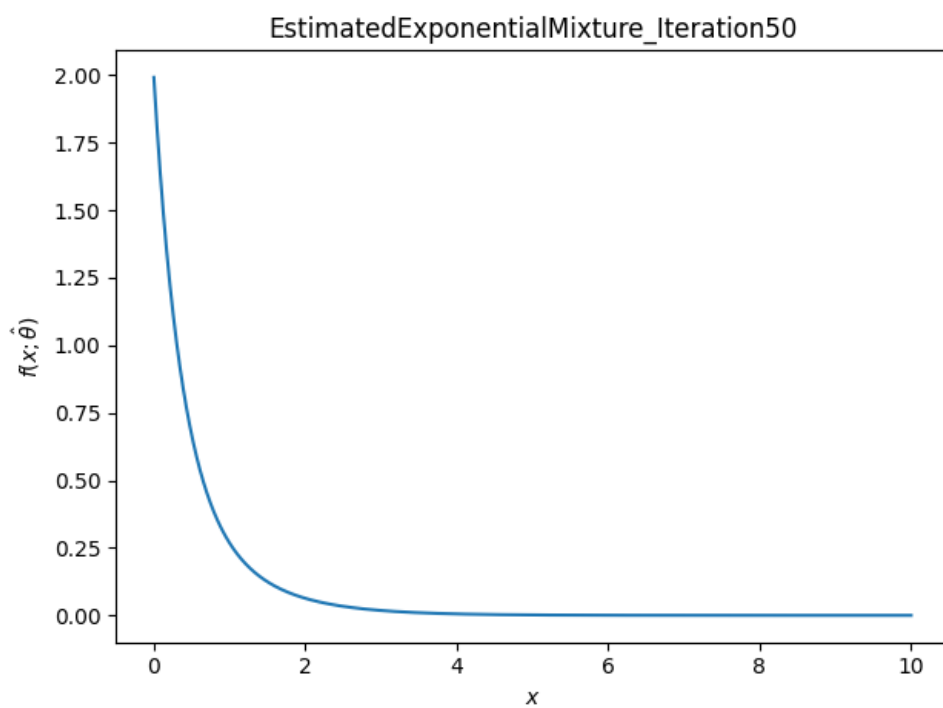
$\hat{\theta}$  estimate at iteration 20:

- $\hat{\lambda}_1 = 1.4025173$
- $\hat{\lambda}_2 = 3.295298$
- $\hat{\pi}_1 = 0.74601278$
- $\hat{\pi}_2 = 0.25398722$



$\hat{\theta}$  estimate at iteration 30:

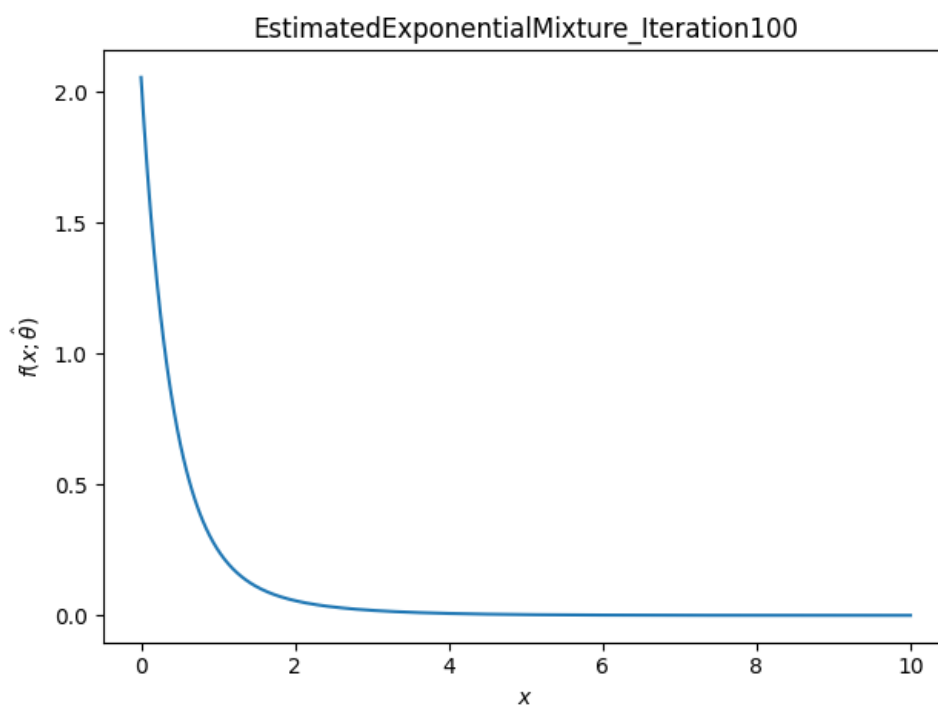
- $\hat{\lambda}_1 = 1.3397352$
- $\hat{\lambda}_2 = 3.1667288$
- $\hat{\pi}_1 = 0.680864$
- $\hat{\pi}_2 = 0.319136$



$\hat{\theta}$  estimate at iteration 50:

- $\hat{\lambda}_1 = 1.199646$
- $\hat{\lambda}_2 = 2.938034$
- $\hat{\pi}_1 = 0.544634$
- $\hat{\pi}_2 = 0.455369$

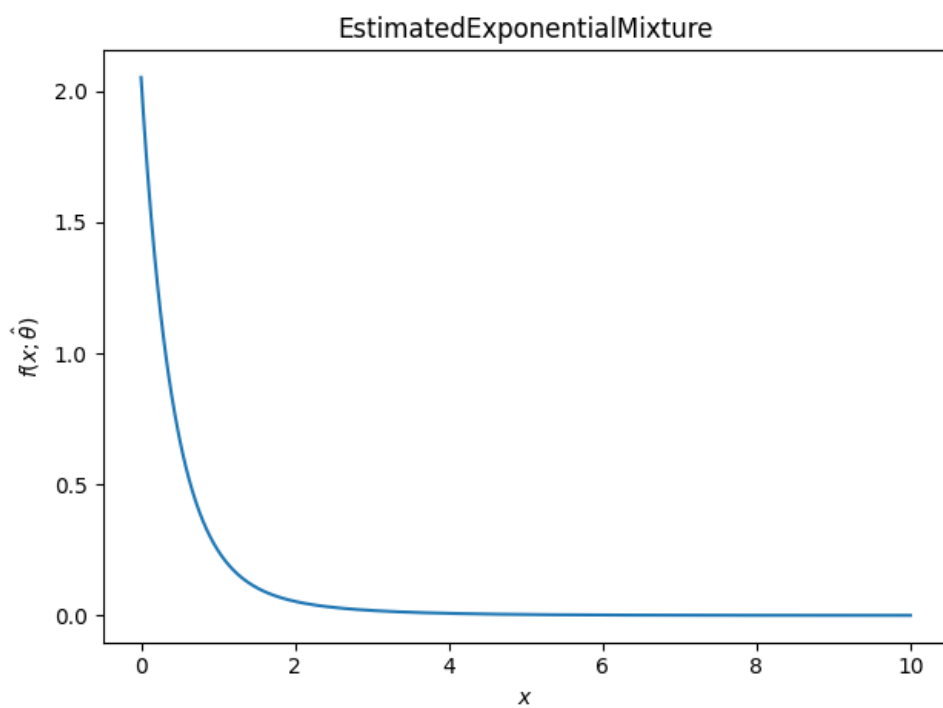




$\hat{\theta}$  estimate at iteration 100:

- $\hat{\lambda}_1 = 0.9189678$
- $\hat{\lambda}_2 = 2.5808147$
- $\hat{\pi}_1 = 0.316126$
- $\hat{\pi}_2 = 0.683873$

The final estimate is:



The final values found for  $\hat{\theta}$  are:

- $\hat{\lambda}_1 = 0.8330837$
- $\hat{\lambda}_2 = 2.4772141$
- $\hat{\pi}_1 = 0.2577022$
- $\hat{\pi}_2 = 0.7422978$