Homework 1 Responses

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1

I spent a lot of time on questions 3,4, and 5 and could not finish this.

2

Similarly, I could not finish this question either.

3.a

Plots are given in figure 1. I used the log-likelihood function as my stopping criteria. For the these non-stochastic experiments I stopped iterating when the improvement in the log-likelihood was less than .001.

3.b
Tabulated AIC and BIC scored below:

k bic iters 4 433.4* 455.1* 88 6 450.7 482.0 40 8 449.4 490.4 267 11 440.0 495.4 173 15 448.6 523.2 179 20 468.7 567.4 476				
4 433.4* 455.1* 88 6 450.7 482.0 40 8 449.4 490.4 267 11 440.0 495.4 173 15 448.6 523.2 179		aic	bic	iters
6 450.7 482.0 40 8 449.4 490.4 267 11 440.0 495.4 173 15 448.6 523.2 179	k			
8 449.4 490.4 267 11 440.0 495.4 173 15 448.6 523.2 179	4	433.4*	455.1*	88
11 440.0 495.4 173 15 448.6 523.2 179	6	450.7	482.0	40
15 448.6 523.2 179	8	449.4	490.4	267
	11	440.0	495.4	173
20 468.7 567.4 476	15	448.6	523.2	179
	20	468.7	567.4	476

Best values of each are marked with a *.

3.c

Plots for the location-scale models are given in figure 2. For the location-scale models I limited the σ^2 's to have a minimum of .001.

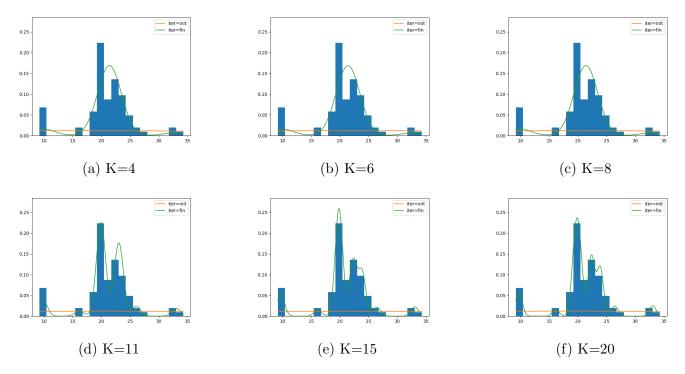


Figure 1: Non-Stochastic 1-D Location Mixtures

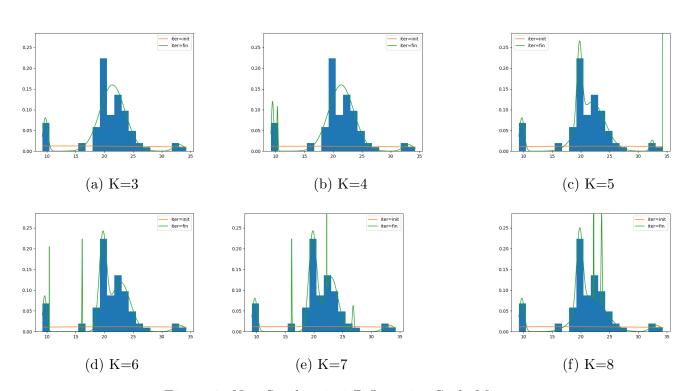


Figure 2: Non-Stochastic 1-D Location-Scale Mixtures

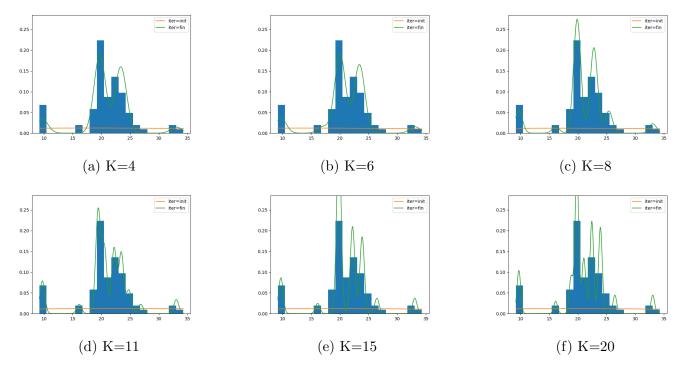


Figure 3: Stochastic 1-D Location Mixtures

3.d Tabulated AIC and BIC scored below:

	aic	bic	iters
k			
3	424.4	446.0*	64
4	425.1	454.0	71
5	410.1*	446.2	123
6	425.9	469.3	549
7	410.5	461.1	522
8	412.0	469.7	499

Best values of each are marked with a *.

3.e

It seemed that the location only models produced more reliable results. Allowing the variance parameters to update allowed the excess centroids to increase the log-likelihood arbitrarily by decreasing the variance to zero. This is why I limited the minimum σ^2 's.

It also appeared that initialization played a large role in the goodness of fit. The resulting mixture pdf's could vary substantially on repeated runs.

4.a

Plots given in figure 3. For the these *stochastic* experiments I stopped iterating when the improvement in the log-likelihood was less than .1. I lowered this threshold because I found that the stochasticity seemed to make the algorithm take thousands of iterations to converge in some cases.

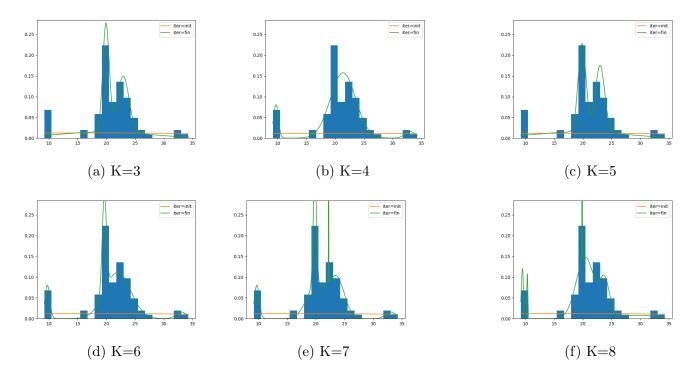


Figure 4: Stochastic 1-D Location-Scale Mixtures

4.bTabulated AIC and BIC scored below:

	aic	bic	iters
k			
4	478.7	500.4	29
6	435.2*	466.5*	53
8	443.2	484.1	98
11	455.2	510.6	78
15	448.6	523.2	117
20	475.6	574.3	77

Best values of each are marked with a *.

4.c

4.d

Tabulated AIC and BIC scored below:

aic	bic	iters
443.9	465.5	110
373.8	402.7	49
397.3	433.4	28
319.2*	362.5*	78
434.6	485.2	43
334.8	392.6	25
	443.9 373.8 397.3 319.2* 434.6	443.9 465.5 373.8 402.7 397.3 433.4 319.2* 362.5* 434.6 485.2

Best values of each are marked with a *.

4.e

The stochastic EM models did not seem to perform as well as the normal EM versions. The location only models tended to favor a high value of the sole $sigma^2$ parameter, but the location-scale models did tend to produce less centroids that became singularities.

5

I sadly did not have time to complete this.