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November 20, 2018

1 Bayesian Inference

- 1. (a)
 - (b)
 - (c)
 - (d)
 - . .
 - (e)
- 2. (a)
 - (b)

2 Kernel Regression

- 3. (a)
 - (b)
 - (c)
 - (d)
 - (e)

3 k-means clustering

4. (a) Figures below show cluster centers in the full MNIST set (train joined with test) from k-means for a variety of different cluster counts. In general, they all resemble identifiable numerals, but more clusters tends to lead to more distinct features. At k=5 we see centers somewhere in between similar numerals like 3 and 8 or 9 and 4, but at k=20 the algorithm has the freedom to center clusters around more specific forms like taller or rounder or more slanted 9s.

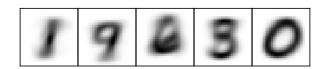


Figure 3.1: Final centers for k = 5 with uniform initialization

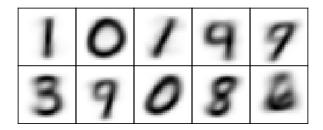


Figure 3.2: Final centers for k = 10 with uniform initialization

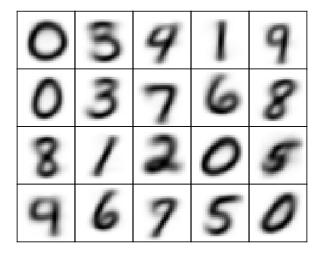


Figure 3.3: Final centers for k = 20 with uniform initialization

- (b) Figures below show centers obtained with the k-means++ intialization scheme. They look broadly similar to the uniform case, although perhaps modestly crisper.
 - Fig. 4b compare the convergence rate to uniform initialization through the mean squared distance from a point to its cluster center. We see that other than for k=5, the new initialization converges faster and to smaller mean distances.



Figure 3.4: Final centers for k = 5 with k-means++ initialization

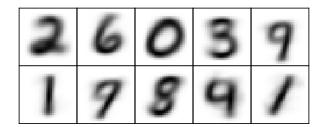


Figure 3.5: Final centers for k = 10 with k-means++ initialization

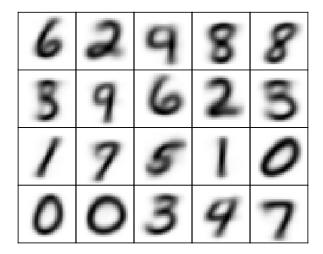


Figure 3.6: Final centers for k=20 with k-means++ initialization

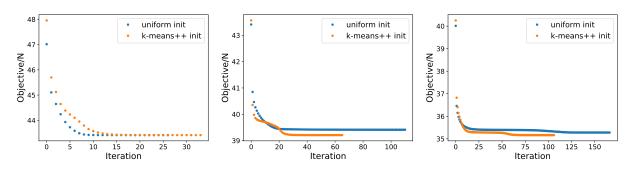


Figure 3.7: Objective function vs. iteration number for k = 5, 10, 20 from left to right

4 Joke Recommender System

- 5. (a)
 - (b)
 - (c)