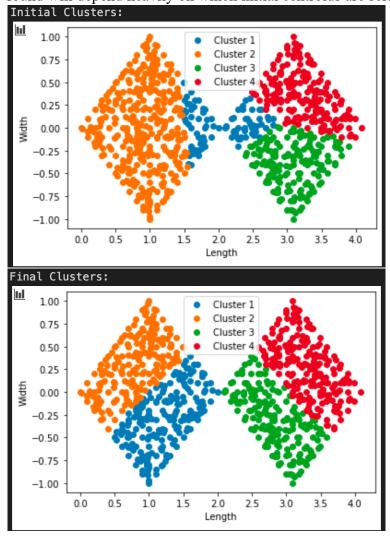
Artificial Intelligence HW 3

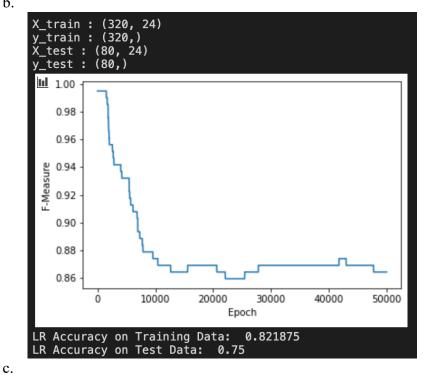
Problem 1

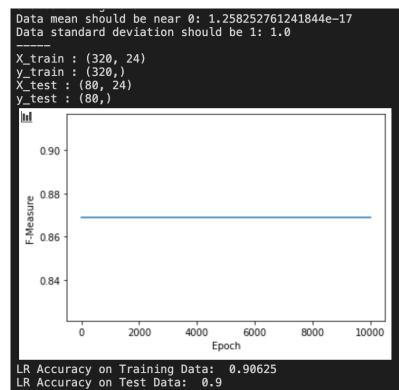
- a. See included Python implementation
- b. The Initial Clusters image shows the initial clusters (in the center of which lies the initial randomly chosen centroids). The Final Clusters image shows the clusters found by the k-means algorithm after 100 iterations. The results shown make sense, as the k-means algorithm will find the "local minimum" for its final centroid. Thus, the final clusters found will depend heavily on which initial centroids are selected.



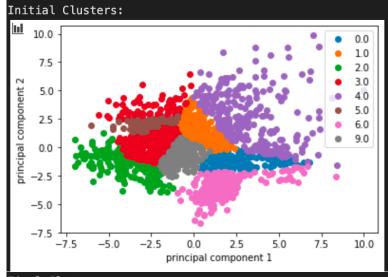
Problem 2

a. See included Python implementation

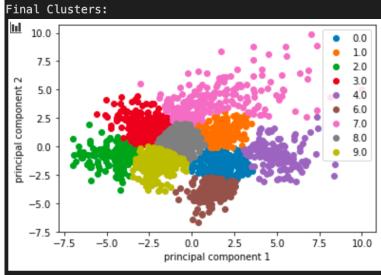




Problem 3



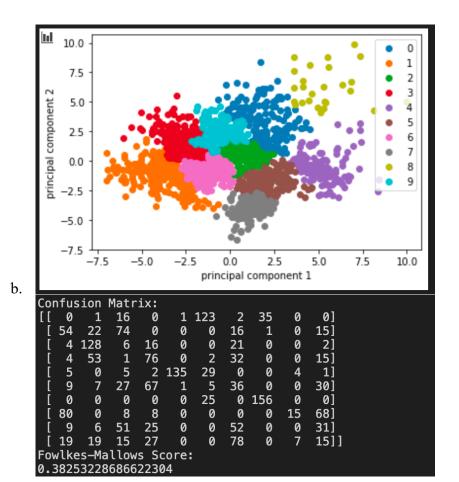


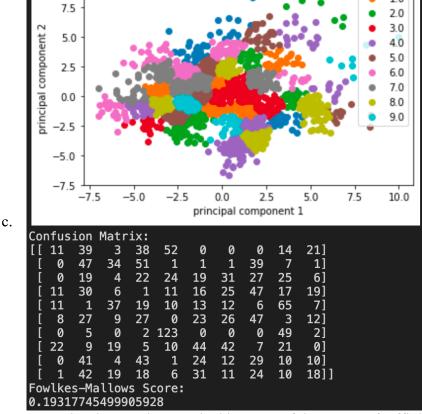


Actual vs. Predicted Data Array % Similarity: 90.0%

Confusion			Matrix:								
	[[:	124	6	0	0	5	0	40	0	1	2]
	[6	74	0	0	0	0	2	10	63	27]
	[1	8	106	16	0	0	0	1	10	35]
	[4	0	38	76	0	0	0	15	10	40]
	[12	8	0	2	150	0	0	8	1	0]
	[8	15	2	60	1	0	0	13	60	23]
	[21	1	0	0	0	0	159	0	0	0]
	[0	14	0	12	0	0	0	146	7	0]
	[9	27	2	25	0	0	0	10	75	26]
	[2	8	7	15	0	0	0	29	61	58]]
Fowlkes-Mallows Score:											

0.41134857985642637





Note: the above values are bad because of the nature of Affinity Propagation. It fails at classifying the values by clustering things correctly as it does not limit itself to 10 clusters. The above images are the results after normalizing the labels provided by the algorithm.

1.0

Problem 4:

```
X_train : (320, 24)
     train : (320,)
     test : (80, 24)
             (80,)
     Measure: 0.9885057471264368
a.
              (320, 24)
              (320,)
          : (80, 24)
             (80,)
     Measure: 0.9885057471264368
b.
     train : (320, 24)
     train : (320,)
             (80, 24)
             (80,)
   F-Measure: 1.0
```