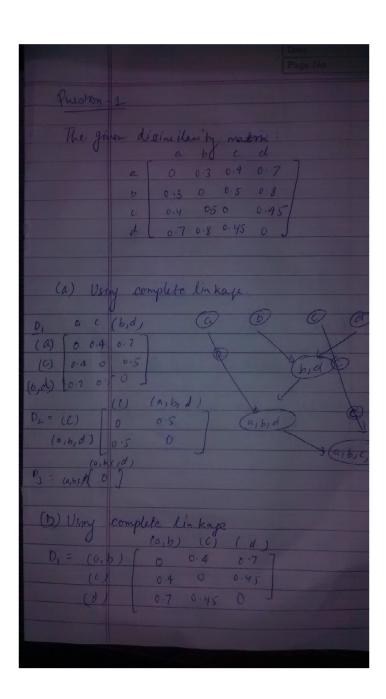
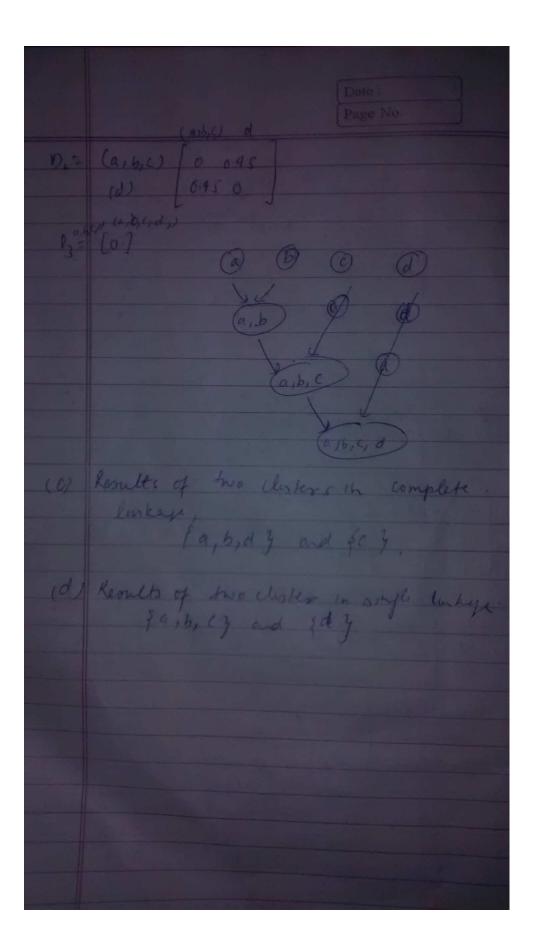
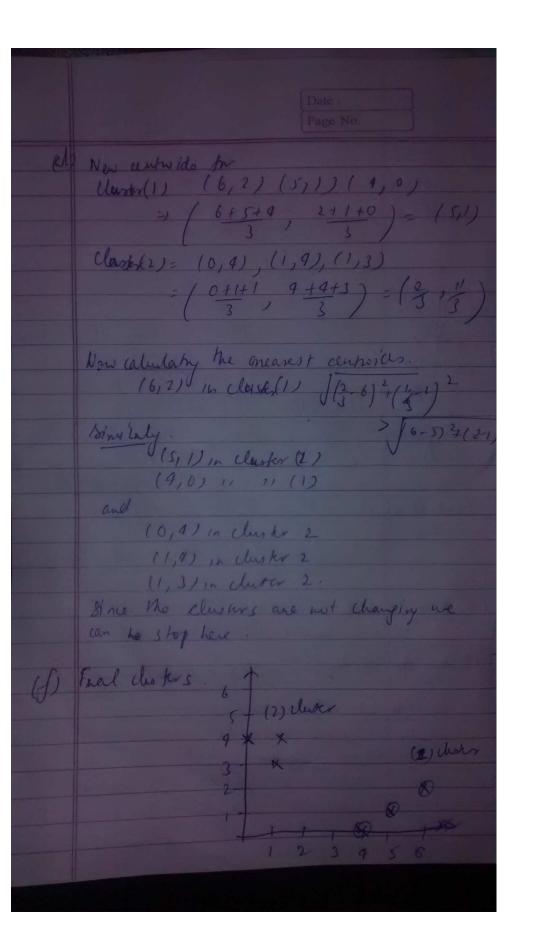
# Report

# Question 1

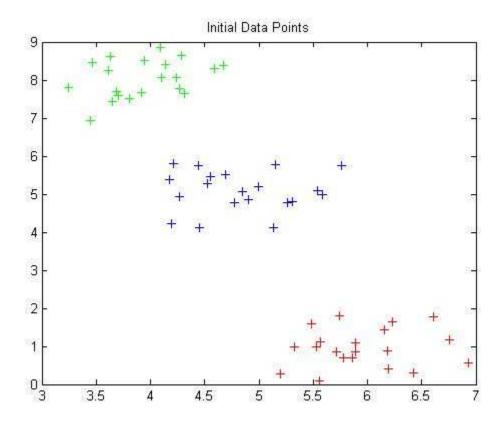




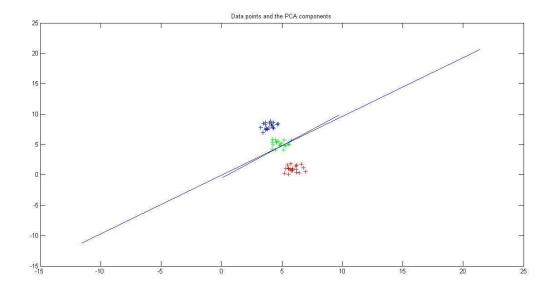
	Date:
	Page No.
(1) K=2	Obs. X1 X2
M= 6	A CONTRACTOR OF STATE
9=2	2 1 3
	3 0 9
1	4 6 2
6-	6 4 0
a) 5 2 2	(b) Random initial salvage
400	clusters
2	() (6,2)->
1	(a) Clockr(1)
	(6,2),(5,1),(1,3)
112	3 4 5 6 Clistis (2)
GIA IN	(6,4),(1,4),(4,0)
( Centrid of	
"	(2) - (0+14, 4:+4+6) = (5/3, 8/2)
(d) Clisest al	stree for (6,2)=(1)](6-9)\$12-2)2 4
	class 2 (2) (6-5/3)2/(8-2)2
Distance for	des kr(2) > Distance for closer 1
80 (6)	2) in chrskr(1)
	(5,1) in cluster (1) (1,3) incluster (2)
	(6,9) 11 1(2)
	(1,4) " (2)
	(4,0) (1)
	A Transferred and the second



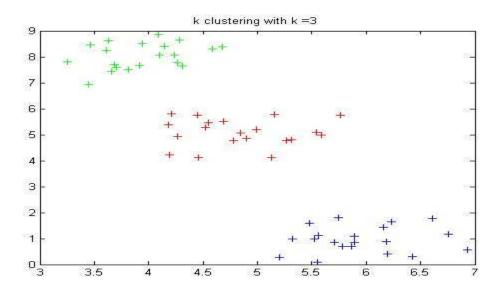
### (a) The 60 observation initially with the data labels



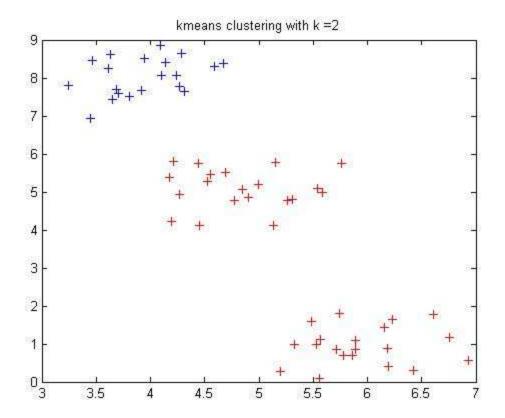
3(b) The data points with the PCA vector shown



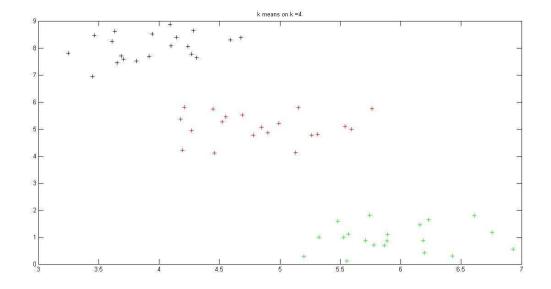
### 3(c) K means clustering on the data with k=3



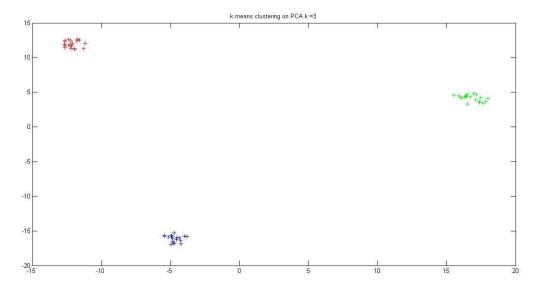
#### (d) k means clustering on K =2



### (e) k means clustering on the data with k=4 on the original data

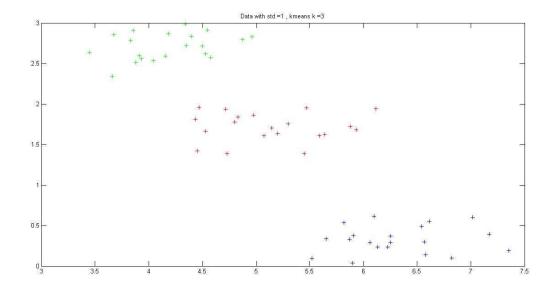


(f) k means clustering on the PCA points with k = 3.

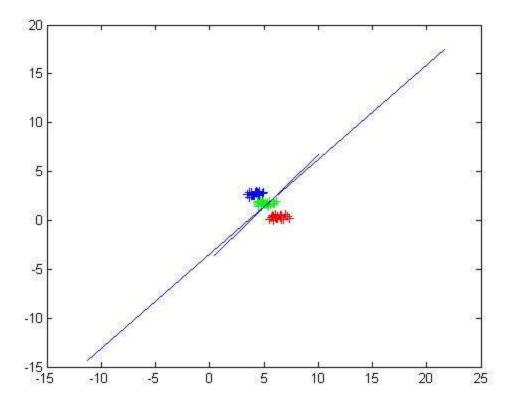


It can be seen in the PCA of the data points the points are more far separated than the original data .

(g) k means cluster on the points with std deviation = 1

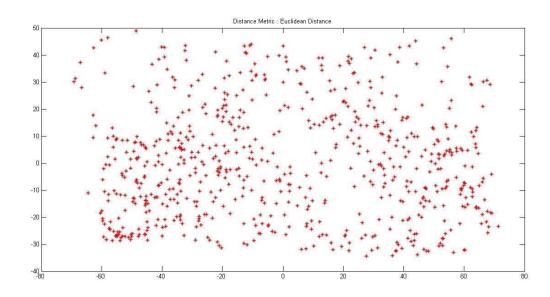


#### With std deviation and PCA



It can be seen by changing the standard deviation of the data to 1 the data points are more closer as the in the original data standard deviation was 0.5.

#### The 2D embedding with euclidean distance



### (c) Embedding with distance metric as hamming

