Problem1:

	Cluster1	Cluster2	Cluster3	Cluster4	Sum (O)
Label 1	0	1	4	0	5
Label 2	5	0	0	0	5
Label 3	0	5	0	0	5
Label 4	0	0	1	4	5
Sum (C)	5	6	5	4	20

	Same Cluster	Different Clusters
Same Class	TP = 32	FN = 8
Different Class	FP = 9	TN = 141

```
Purity = (5 + 5 + 4 + 4) / 20 = 0.9

Precision = TP / (TP + FP) = 0.78

Recall = TP / (TP + FN) = 0.8

F-measure = 2P * R / (P + R) = 0.79

I(C, O) = 5 / 20 * log2(20 * 5 / (5 * 5)) + 1 / 20 * log2(20 * 1 / (5 * 6)) + 5 / 20 * log2(20 * 5 / (5 * 6)) + 4 / 20 * log2(20 * 4 / (5 * 5)) + 1 / 20 * log2(20 * 4 / (5 * 5)) + 4 / 20 * log2(20 * 4 / (5 * 5)) + 4 / 20 * log2(20 * 4 / (5 * 4)) + 0.5 - 0.029 + 0.4342 + 0.3356 - 0.016 + 0.4 = 1.6248
H(O) = [(-5 / 20) * log2(5 / 20)] * 4 = 2 + 1.6248
H(C) = -[5 / 20 * log2(5 / 20)] * 2 + 6 / 20 * log2(6 / 20) + 4 / 20 * log2(4 / 20)] = 1.99
NMI(C,O) = I(C,O) / sqrt(H(C) * H(O)) = 0.814
```

This result can also be checked by:

python evaluate.py

purity: 0.9 TP: 32.0 TN: 141.0 FP: 9.0 FN: 8.0

precision: 0.780487804878

recall: 0.8

F-measure: 0.79012345679

Problem2 (K-means):

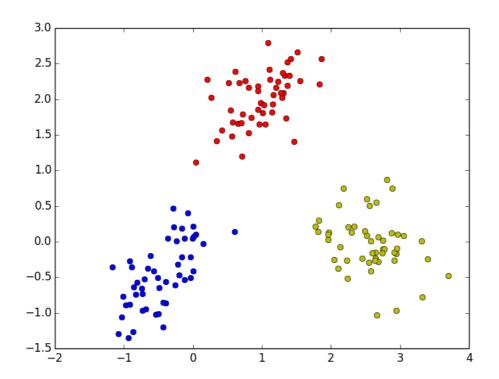
Strength: logic is clear and thus easy to implement. If not a lot of noise can do pretty well. **Weakness:** when the clusters have different sizes and variances, it doesn't give good result. It also doesn't do well in distinguish spherical shaped clusters. Easily affected by distant outliers.

Output:

For dataset1 Iteration :3

Purity is :1.0 NMI :1.0

Cluster 0 size :50 Cluster 1 size :50 Cluster 2 size :50



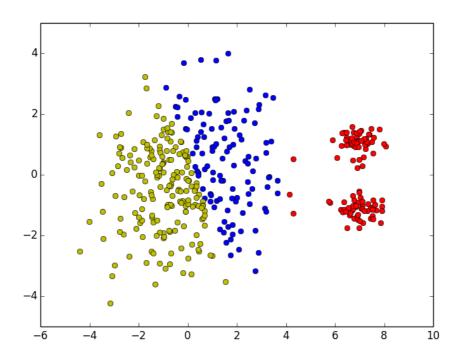
For dataset2

Iteration:9

Purity is :0.8675

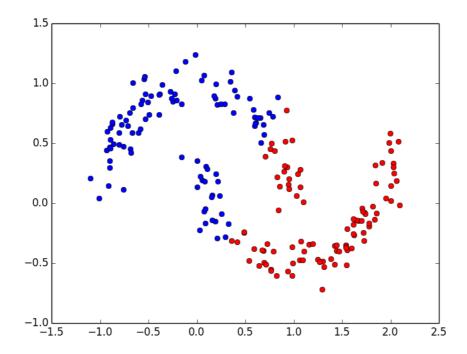
NMI :0.464256439333 Cluster 0 size :103 Cluster 1 size :112

Cluster 2 size :185



For dataset3 Iteration :6 Purity is :0.78

NMI :0.169704955284 Cluster 0 size :98 Cluster 1 size :102



Problem3 (DBSCAN):

Strength: if parameters are chosen well, it can output reasonable result, even for spherical shaped clusters. It can also detect noises and thus minimize their negative influence on the result.

Weakness: the result is extremely dependent to parameters. Chosen different Eps and Minpts can result in totally different outcomes, and some of them are bad clustering.

Output:

For dataset1

Esp: 0.477548092264

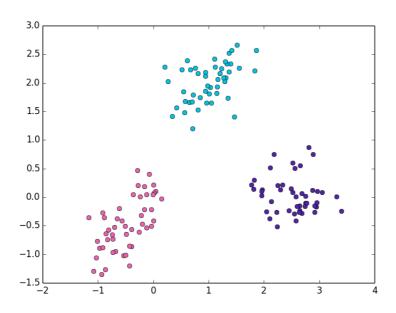
Number of clusters formed :3

Noise points:4

Purity is :0.973333333333

NMI:1.0

Cluster 0 size :50 Cluster 1 size :47 Cluster 2 size :49



For dataset2

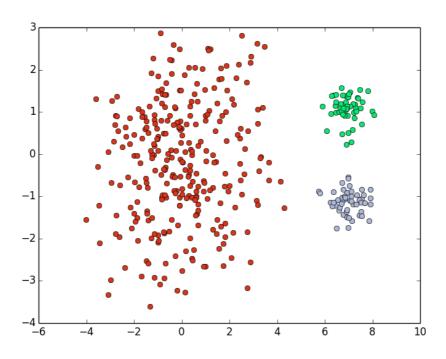
Esp: 0.7405776933

Number of clusters formed:3

Noise points :11 Purity is :0.9725

NMI:1.0

Cluster 0 size :50 Cluster 1 size :289 Cluster 2 size :50



For dataset3

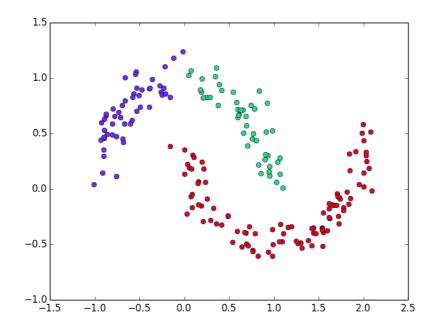
Esp:0.188103671919

Number of clusters formed :3

Noise points :4 Purity is :0.985

NMI:0.817348927469

Cluster 0 size :99 Cluster 1 size :47 Cluster 2 size :51



Problem4 (GMM):

Strength: It is robust and general, will not be affected too much by outliers. Mixture models are more general than portioning.

Weakness: compared to the previous two methods, it runs relatively slow and is much harder to implement. It also converges to a local optimum, not necessarily the global optimum.

Output:

For dataset1 Number of Iterations = 26

After Calculations

Final mean =

-0.462497803748 -0.463941570915

0.98985865093 2.01176299401

2.57343074153 -0.0271142473287

Final covariance =

For Cluster: 1

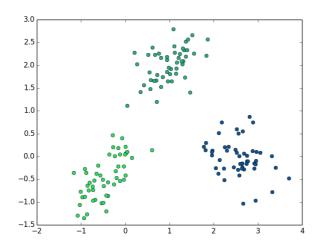
For Cluster : 2

For Cluster: 3

0.180387888855 -0.0467201611333 -0.0467201611333 0.152057292033

Purity is :1.0 NMI :1.0

Cluster 0 size :50 Cluster 1 size :50 Cluster 2 size :50



For dataset2 Number of Iterations = 60

After Calculations

Final mean =

6.97649235198 -0.0259625658049 -0.3059097563 0.0290544014694 2.02079952275 -1.28868763161

Final covariance =

For Cluster: 1

0.227976706733 -0.0302421268225 -0.0302421268225 1.30257250042

For Cluster: 2

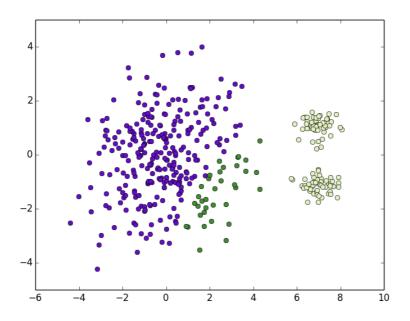
2.39135938542 0.6518938104930.651893810493 2.28436205204

For Cluster: 3

1.57511651604 0.663164323351 0.663164323351 1.21555931019

Purity is :0.875

NMI :0.603803956121 Cluster 0 size :100 Cluster 1 size :265 Cluster 2 size :35



For dataset3 Number of Iterations = 110

After Calculations

Final mean =

Final covariance =

For Cluster: 1

0.769219007365 -0.287852063191 -0.287852063191 0.19019007583

For Cluster : 2

0.682736130041 -0.300607450603 -0.300607450603 0.175861444996

Purity is :0.69

NMI :0.075947839504 Cluster 0 size :106 Cluster 1 size :94

