## Problem 1

```
[1]: import numpy as np
     from numpy.linalg import norm
     from matplotlib import pyplot as plt
     pi = np.array([0.04, 0.3, 0.3, 0.36])
     mu = np.array([[0., 0.],
                     [0., 5.],
                     [5., 0.],
                     [5., 5.]])
     sigma1, sigma2 = np.diag([0.4, 0.4]), np.diag([2.0, 2.0])
     sigma = np.stack([sigma1, sigma2, sigma2, sigma2], axis=0)
[2]: def rnormmix(pi, mu, sigma):
         _pi = pi.cumsum()
         k = np.random.rand()
         for i, pii in enumerate(_pi):
             if k < pii:</pre>
                 return np.random.multivariate_normal(mu[i], sigma[i])
     def J(X, mu):
         k = mu.shape[0]
         1 = np.array([
                 np.argmin([
                     norm(X[i, :] - mu[j]) for j in range(k)
                 ])
                  for i in range(X.shape[0])
         ])
         X, s = np.array([]), 0.
         for j in range(k):
             _{X} = X[1 == j, :]
             s += np.sum([
                     np.sum([
                         norm(_X[i, :] - _X[j, :])**2 for j in range(i+1)
                      for i in range(_X.shape[0])
             ]) * 0.5 / np.sum(1 == j)
         return s
```

```
[3]: def cluster(X, k, TOL=1e-5):
    n, d = X.shape
    mu = 5 * np.random.rand(k, d)
    mu_old = mu + 1000
```

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[4]: X, Js, mu_train = np.array([]), np.array([]), np.array([])
for _ in range(20):
    X = np.array([rnormmix(pi, mu, sigma) for _ in range(1000)])
    mu_train = cluster(X, 4)
    Js = np.append(Js, J(X, mu_train))
Js.mean(), Js.std()
```

[4]: (1734.0371260112297, 80.20126748064264)

```
[5]: def rdiscrete(p):
         _p = p.cumsum()
         k = np.random.rand()
         for i, pii in enumerate(_p):
             if k < pii:</pre>
                 return i
     def cluster(X, k, TOL=1e-5):
         n = X.shape[0]
         mu = X[np.random.choice(np.arange(n)), :]
         for _ in range(k-1):
             d = np.array([
                     np.min([ norm(X[i, :] - muj)**2 for muj in mu ])
                      for i in range(n)
             ])
             mu = np.vstack([mu, X[rdiscrete(d / d.sum()), :]])
         mu old = mu + 1000
         l = np.array([])
         while np.abs((mu_old - mu).mean()) > TOL:
             1 = np.array([
                     np.argmin([
                         norm(X[i, :] - mu[j]) for j in range(k)
                     ])
```

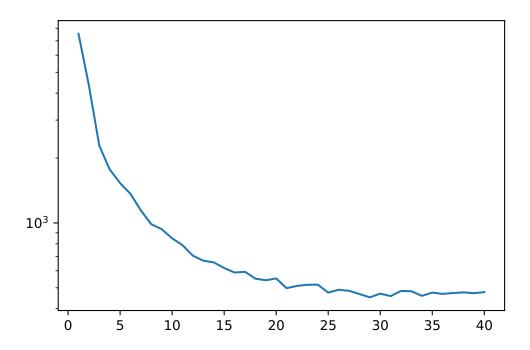
```
[6]: X, Js, mu_train = np.array([]), np.array([]), np.array([])
for _ in range(20):
    X = np.array([rnormmix(pi, mu, sigma) for _ in range(1000)])
    mu_train = cluster(X, 4, TOL=1e-8)
    Js = np.append(Js, J(X, mu_train))
Js.mean(), Js.std()
```

[6]: (1736.5714743691067, 99.26874328463809)

```
[7]: Js = np.array([])
for k in range(40):
    X, mu_train = np.array([]), np.array([])
    X = np.array([rnormmix(pi, mu, sigma) for _ in range(1000)])
    mu_train = cluster(X, k+1)
    Js = np.append(Js, J(X, mu_train) + (k+1)*np.log(1000))
    np.argmin(Js)
```

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```
[12]: fig = plt.semilogy(np.arange(40)+1, Js)
plt.savefig("tmp.png")
```



We wish to set  $\mu_i = \frac{1}{|\mathcal{S}_i|} \sum_{x \in \mathcal{S}_i} \phi(x)$  in order to calculate  $class(i) = \arg\min_{j \in [k]} \|\phi(x_i) - \mu_j\|^2$ .

$$\|\phi(x_i) - \mu_j\|^2 = \langle \phi(x_i) - \mu_j, \phi(x_i) - \mu_j \rangle$$

$$= \langle \phi(x_i), \phi(x_i) \rangle - 2 \langle \phi(x_i), \mu_j \rangle + \langle \mu_j, \mu_j \rangle$$

$$= K(x_i, x_i) - \frac{2}{|\mathcal{S}_j|} \sum_{x \in \mathcal{S}_j} K(x_i, x) + \frac{1}{|\mathcal{S}_j|^2} \sum_{x, x' \in \mathcal{S}_j} K(x, x')$$

$$=: K(x_i, \mathcal{S}_j)$$

Therefore, we instead:

Set class(i) = 
$$\underset{j \in [k]}{\operatorname{arg\,min}} K(x_i, \mathcal{S}_j)$$