HW6_P3

November 12, 2018

```
In [1]: import numpy as np
        import matplotlib.pyplot as plt
In [2]: # Problem 3.5
In [3]: d = 1
        n = 500
        k = 2
        pi = np.array([0.7, 0.3])
        w = np.array([-2, 1])
        b = np.array([0.5, -0.5])
        sigma = np.array([0.4, 0.3])
        X = np.random.uniform(0, 1, n)
        mu = 0.5 # np.mean(X)
        #mu = mu * w + b
In [4]: y = np.zeros(n)
        for l in range(k):
            for i in range(n):
                y[i] += pi[1] * (np.random.normal(w[1]*X[i] + b[1], sigma[1], 1))
In [5]: def E_step(n, k, mu_hat, y, pi_hat, w_hat, b_hat, sigma_square_hat):
            W_temp = np.zeros((n, k))
            W = np.zeros((n, k))
            for l in range(k):
                for i in range(n):
                    \#W_{temp[i][l]} = pi_{hat[l]} * (1/(2*np.pi*(sigma_square_hat[l]))**0.5) * np.
                    W_{temp[i][1]} = pi_{hat[1]} * (1/(2*np.pi*(sigma_square_hat[1]))**0.5) * np.ext[1]
            for i in range(n):
                W[i] = W_temp[i] / sum(W_temp[i])
            return W
In [59]: def M_step(n, k, mu, y, pi_hat, w_hat, b_hat, sigma_hat, W):
             # calculate pi
             for l in range(k):
```

```
'''# calculate w
             w_hat_new = np.zeros(k)
             w_hat_temp = np.zeros((n, k))
             for l in range(k):
                 w_hat_temp[:,l] = W[:,l] * (X - b_hat[l])
                 w_hat_new[l] = sum(w_hat_temp[:,l]) / (mu * sum(W[:,l]))
             # calculate b
             b_hat_new = np.zeros(k)
             b_hat_temp = np.zeros((n, k))
             for l in range(k):
                 b_hat_temp[:,l] = W[:,l] * (X - w_hat[l]*mu)
                 b_hat_new[l] = sum(b_hat_temp[:,l]) / sum(W[:,l])'''
             mu_hat = np.zeros(k)
             mu_hat_temp = np.zeros((n, k))
             for l in range(k):
                 mu_hat_temp[:,1] = W[:,1] * y
                 mu_hat[1] = sum(mu_hat_temp[:,1]) / sum(W[:,1])
             w_hat = (mu_hat - b_hat)/mu
             b_hat = mu_hat - w_hat*mu
             # calculate sigma square!
             sigma_square_hat_temp = np.zeros((n, k))
             for 1 in range(k):
                 \#sigma\_square\_hat\_temp[:,l] = W[:,l] * ((X - w\_hat\_new[l]*mu - b\_hat\_new[l])*
                 \#sigma\_square\_hat\_temp[:,l] = W[:,l] * ((y - w\_hat[l]*mu - b\_hat[l])**2)
                 sigma_square_hat_temp[:,1] = W[:,1] * ((y - mu_hat[1])**2)
                 sigma_square_hat[1] = sum(sigma_square_hat_temp[:,1]) / sum(W[:,1])
             #return pi_hat, w_hat_new, b_hat_new, sigma_square_hat, mu_hat
             return pi_hat, w_hat, b_hat, sigma_square_hat, mu_hat
In [60]: # main function
         # initializing parameters
         pi_hat = np.array([0.5, 0.5])
         w_{hat} = np.array([1, -1])
         b_hat = np.array([0, 0])
         sigma_square_hat = np.array([np.std(y), np.std(y)])**2
         log_like = []
         log_like_prev = -1000
         increase = 1
         mu_hat = w_hat * mu + b_hat
```

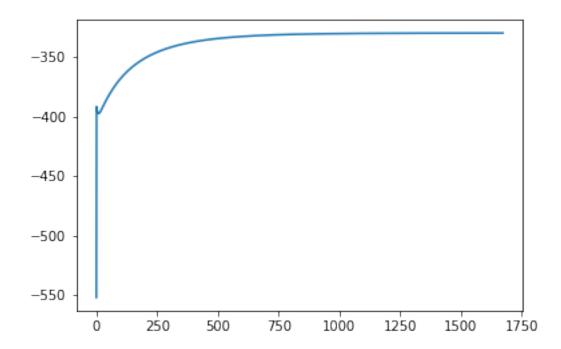
pi_hat[1] = sum(W[:,1]) / n

```
while abs(increase) > 1e-04:
    W = E_step(n, k, mu_hat, y, pi_hat, w_hat, b_hat, sigma_square_hat)

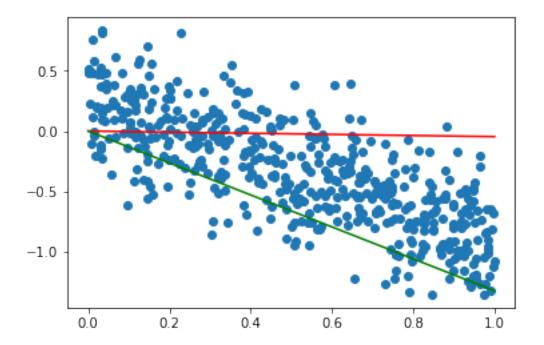
J = 1
for i in range(n):
    J_temp = 0
    for l in range(k):
        #J += W[i][l] * pi[l] * (1/(2*np.pi*(sigma_square_hat[l]))**0.5) * np.exp
        J_temp += pi[l] * (1/(2*np.pi*(sigma_square_hat[l]))**0.5) * np.exp(-(y[i] J *= J_temp))

log_like.append(np.log(J))
increase = np.log(J) - log_like_prev
log_like_prev = np.log(J)

[pi_hat, w_hat, b_hat, sigma_square_hat, mu_hat] = M_step(n, k, mu, y, pi_hat, w_i)
```



```
In [62]: [pi, w, b, sigma]
Out[62]: [array([0.7, 0.3]), array([-2, 1]), array([ 0.5, -0.5]), array([0.4, 0.3])]
In [63]: [pi_hat, w_hat, b_hat, sigma_square_hat**0.5, mu_hat]
```



In []:
In []: