Homework 2016-03-21

Chuan Lu 13300180056

March 23, 2016

Problem 1.

Calculate the confidence interval with a confidence level of 0.95 of the samples given.

Proof. Firstly, the mean of the samples $\overline{x} = \frac{1}{10} \sum_i x_i = 10.05$, the variance of the samples $\sigma = \frac{1}{10} \sum_i (x_i - \overline{x})^2 = 0.0525$. We can then check from the t-distribution table that the confidence interval is $[\overline{X} - t_{\alpha/2}(n-1)\frac{\sigma}{\sqrt{n}}, \overline{X} + t_{\alpha/2}(n-1)\frac{\sigma}{\sqrt{n}}]$ $t_{\alpha/2}(n-1)\frac{\sigma}{\sqrt{n}}$ = [9.9854, 10.1146]

Problem 2.

If the rate of abnormality in this area is below the average with the information provided.

Proof. P(Only one person is of abnormality|The rate of abnormality is 0.01) = $\binom{400}{1}*(1-0.01)^{399}*0.01$ = 0.0725 > 0.05, which implies that this phenomenon is just possible, hence the rate of abnormality in this area can NOT be seen as below the average.

Derive out the EM algorithm of a distribution mixed by three normal distributions.

Proof. Let

$$z_{ij} = \begin{cases} 1 & X_i \text{ belongs to the } j^{th} \text{ distribution,} \\ 0 & \text{others,} \end{cases}$$

and $P(z_i=1)=t_i$, where $\sum_{i=1}^3 t_i=1$. Then the likelihood function is $L(\theta;X,Z)=\prod_{i=1}^n \prod_{j=1}^3 t_j \frac{1}{(\sqrt{2\pi})^n \sqrt{(\det(\Sigma_j))}} e^{-\frac{1}{2}(x-\mu)^T \Sigma^{-1}(x-\mu)}$, the log-likelihood

of which is $l(\theta; x, z) = \sum_{i=1}^{n} \sum_{j=1}^{3} z_{ij} [log(t_j) - \frac{d}{2} log(2\pi) - \frac{1}{2} (x_i - \mu_j)^T \Sigma^{-1} (x - \mu)].$ E-step:

Assume there are a vector of θ_n , then $Q(\theta|\theta_m) = E(l(\theta;x,z)) = \sum_{i=1}^n \sum_{j=1}^3 T_{i,j}^{(m)} [log(t_j) - \frac{d}{2}log(2\pi) - \frac{1}{2}(x_i - \mu_j)^T \Sigma^{-1}(x - \mu)]$ in which $T_{ij}^{(m)} = P(z_{ij} = 1|X_i = x_i; \theta_m) = \frac{t_j^{(m)} f(x_i; \mu_j^{(m)}, \Sigma_j^{(m)})}{\sum_{j=1}^3 t_i^{(m)} f(x_i; \mu_j^{(m)}, \Sigma_j^{(m)})}, j = 1, 2, 3.$

M-step:

$$t_{m+1} = argmax_t Q(\theta|\theta_m) = argmax_t \left[\sum_{i=1}^n T_{i,1}^{(m)}]log(t_1) + \left[\sum_{i=1}^n T_{i,2}^{(m)} \right] log(t_2) + \left[\sum_{i=1}^n T_{i,3}^{(m)} \right] log(t_3) \right], \text{ so } t_j^{(m+1)} = \frac{1}{n} \sum_{i=1}^n T_{i,j}^{(m)}.$$
 For μ_{m+1}, σ_{m+1} ,

$$(\mu_{m+1}, \sigma_{m+1}) = argmax_{\mu,\sigma}Q(\theta|\theta_m), \text{ so } \mu_j^{(m+1)} = \frac{\sum_{i=1}^n T_{i,j}^{(m)} x_i}{\sum_{i=1}^n T_{i,j}^{(m)}}, \text{ and } \sigma_j^{(m+1)} = \frac{\sum_{i=1}^n T_{i,j}^{(m)} (x_i - \mu_j^{(m)})(x_i - \mu_j^{(m)})^T}{\sum_{i=1}^n T_{i,j}^{(m)}}$$

Problem 4.

Estimate the arguments with the data given.

Proof. **0.1** The code is shown as follows.

```
1 source ('em_func.R')
     em = function(x, dim, mu, sigma, t, maxit = 100, error = 1e-6) {
  3
        # mu, sigma, t are lists a containing init data
        \# Init
  4
        n = \frac{nrow}{x}
  5
        T = matrix(rep(0, n*dim), ncol=dim)
  6
  7
  8
         for(step in 1:maxit) {
  9
 10
           \# E - step
            for(j in 1:dim) {
 11
              for(i in 1:n) {
 12
 13
                 \# print(x/i, /)
 14
                 \mathbf{T}[\,\mathbf{i}\;,\;\;\mathbf{j}\,]\;=\;\mathbf{em}.\mathbf{FUNC}(\,\mathbf{x}\,[\,\mathbf{i}\;,\;\;]\;,\;\;\mathbf{mu}[\,[\,\mathbf{j}\,]\,]\;,\;\;\mathbf{sigma}\,[\,[\,\mathbf{j}\;]\,]\,)\;\;*\;\;\mathbf{t}\,(\,\mathbf{j}\,)
 15
 16
 17
           T = T/rowSums(T)
 18
 19
           sigma0 = sigma
 20
            \mathbf{t0} = \mathbf{t}
 21
           mu0 = mu
 22
 23
           # M-step
 24
            for(j in 1:dim) {
 25
              \mathbf{v1} = \mathbf{sum}(\mathbf{T}[\ ,\ \mathbf{j}\ ])
 26
              v2 = matrix(rep(0, ncol(x)), nrow=1)
 27
              for(i in 1:n) {
                 \mathbf{v2} = \mathbf{v2} + (\mathbf{T}[\mathbf{i}, \mathbf{j}] * \mathbf{x}[\mathbf{i}, ])
 28
 29
 30
              mu[[j]] = v2 / v1
 31
              \mathbf{t} [\mathbf{j}] = \mathbf{v1} / \mathbf{n}
 32
              v3 = matrix(rep(0, ncol(x)^2), nrow=ncol(x))
 33
              for(i in 1:n) {
 34
                 temp1 = matrix(x[i, ])
 35
                 temp2 = matrix(mu[[j]])
 36
                 delta = temp1 - temp2
 37
                 v3 = v3 + (T[i, j] * delta \%\% t(delta))
 38
 39
              sigma[[j]] = (v3/v1)
 40
 41
           \mathbf{mu}_{\mathbf{sum}} = 0
 42
           sigma_sum = 0
 43
            for(i in 1: length(mu)) {
 44
              mu\_sum = mu\_sum + sum(abs(mu0[[i]] - mu[[i]]))
 45
 46
            for(i in 1:length(sigma)) {
 47
              sigma_sum = sigma_sum + sum(abs(sigma0[[i]] - sigma[[i]]))
 48
 49
            # Check if converged
 50
            if(mu\_sum < error \& sigma\_sum < error \& sum(abs(t-t0)) < error)
 51
              break
 52
```

```
returnlist = list (mu, sigma, t)
 54
 55
        return(returnlist)
 56 }
     em_FUNC <- function(x, mu, sigma) {
        n = ncol(t(matrix(x)))
        res = 1/((sqrt(2*pi))^n * sqrt(abs(det(sigma)))) * exp(-1/2 * (x-mu)%%solve(sigma)%%(t))
  3
        return (res)
  4
1 # Exercise 4
  2 source( 'EM.R')
  3 \quad data = read.csv('Data1.csv')
  4 \mathbf{x} \mathbf{1} = \mathbf{data}["V1"]
  5 \quad \mathbf{x2} = \mathbf{data}["V2"]
  6 \mathbf{x} = \mathbf{data}.\mathbf{frame}(\mathbf{x1}, \mathbf{x2})
  9 n = 2
 10 \mathbf{mu} = \mathbf{list}(\mathbf{t}(\mathbf{matrix}(\mathbf{runif}(\mathbf{n}))), \mathbf{t}(\mathbf{matrix}(\mathbf{runif}(\mathbf{n}))), \mathbf{t}(\mathbf{matrix}(\mathbf{runif}(\mathbf{n}))))
 11 sigma = list(matrix(runif(n*n), ncol = n), matrix(runif(n*n), ncol = n), matrix(runif(n*n))
 12 \mathbf{t} = \mathbf{c} (0.1, 0.7, 0.2)
 13 rlist = em(x, dim, mu, sigma, t)
 14 print (rlist [[1]])
 15 print (rlist [[2]])
 16 print(rlist[[3]])
```

0.2 The result is shown as follows, in which rlist[[1]] is mu, rlist[[2]] is sigma, rlist[[3]] is t.

```
> print(rlist[[1]])
   [[1]]
3
              [,1]
4
   [1,] -2.274075 -3.209265
5
6
   [[2]]
7
             [,1]
   [1,] 5.271368 -2.079697
9
10
   [[3]]
11
              [,1]
   [1,] 0.8930116 -0.6144073
12
13
14 > print(rlist[[2]])
   [[1]]
15
16
               [,1]
   [1,] 0.6014583 -0.1520107
17
18
   [2,] -0.1520107 0.8660930
19
20
   [[2]]
21
                 [,1]
[1,]
        1.577730561 -0.009383941
   [2,] -0.009383941 1.104548710
```

Problem 5.

Estimate the average height of men and women with the data givem.

Proof. **0.3** The code is shown as follows, and the function 'EM' is given within the answer to the former exercise.

```
# Exercise 5
 2 source( 'EM.R')
    \mathbf{x} = \mathbf{matrix}(\mathbf{c}(171, 174, 159, 176, 164, 169, 170, 173, 159,
                     172, 166, 175, 161, 186, 160, 168, 166, 174,
 4
 5
                     159, 178, 165, 189, 164, 168, 165, 185, 160,
                     175, 172, 168, 167, 171, 160, 174, 168, 174,
 6
 7
                     167, 175, 162, 177)
 8
    dim = 2
 9
   \mathbf{mu} = \mathbf{list} (\mathbf{matrix} (\mathbf{c}(163)), \mathbf{matrix} (\mathbf{c}(176)))
10 sigma = list(matrix(runif(1)), matrix(runif(1)))
11 \mathbf{t} = \mathbf{c} (0.2, 0.8)
   rlist = em(x, dim, mu, sigma, t)
13 print (rlist [[1]])
    print(rlist[[2]])
15 print(rlist[[3]])
```

0.4 The result is shown as follows, in which rlist[[1]] is mu, rlist[[2]] is sigma, rlist[[3]] is t.

```
1 > \mathbf{print}(\mathbf{rlist}[[1]])
 2
    [[1]]
3
               [,1]
 4
    [1,] 163.6147
 5
 6
    [[2]]
 7
               [,1]
    [1,] 172.5814
 8
9
10 > print(rlist[[2]])
11
    [[1]]
12
              [,1]
    [1,] 14.6682
13
14
15
    [[2]]
16
               [,1]
    [1,] 46.11688
17
18
19 > print(rlist[[3]])
```