

MLE, EM, EM acceleration

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Suppose that there are 4 kinds of animals (A, B, C, D) in a farm. And their probability are

$$P(A) = \frac{1}{2} + \frac{\theta}{4}$$

$$P(B) = \frac{1}{4}(1 - \theta)$$

$$P(C) = \frac{1}{4}(1 - \theta)$$

$$P(D) = \frac{\theta}{4}$$

$$\theta \in (0, 1)$$

. Now we randomly choose one animal 197 times and get the result

$$N(A) = 125$$

$$N(B) = 18$$

$$N(C) = 20$$

$$N(D) = 34$$

MLE

We let $X = (x_1, x_2, x_3, x_4) = (125, 18, 20, 34)$

$$\begin{aligned} L(x_i; \theta) &= \left(\frac{1}{2} + \frac{\theta}{4}\right)^{x_1} \left[\frac{1}{4}(1 - \theta)\right]^{x_2} \left[\frac{1}{4}(1 - \theta)\right]^{x_3} \left(\frac{\theta}{4}\right)^{x_4} \\ &\propto (2 + \theta)^{x_1} (1 - \theta)^{x_2 + x_3} (\theta)^{x_4} \end{aligned}$$

$$\begin{aligned} l(x_i; \theta) &= \ln[(2 + \theta)^{x_1} (1 - \theta)^{x_2 + x_3} (\theta)^{x_4}] \\ &= x_1 \ln(2 + \theta) + (x_2 + x_3) \ln(1 - \theta) + x_4 \ln \theta \\ &= 125 \ln(2 + \theta) + 38 \ln(1 - \theta) + 34 \ln \theta \end{aligned}$$

$$l'(x_i; \theta) = \frac{125}{2 + \theta} - \frac{38}{1 - \theta} + \frac{34}{\theta} = 0$$

We use Newton's Method to find the result

```
f=function(x){
  125/(2+x)-38/(1-x)+34/x
}
f1=function(x){
  -125/(2+x)^2-38/(1-x)^2-34/x^2
}
x=0.1
count=0
while (abs(f(x))>0.001&&count<1000) {
```

```

temp=x
x=temp-f(temp)/f1(temp)
count=count+1
}
x

```

[1] 0.6268216

We get $\theta = 0.6268216$

EM

Let's suppose that we have two subspecies of A including A_1, A_2 and

$$P(A_1) = \frac{1}{2}$$

$$P(A_2) = \frac{\theta}{4}$$

Further more, suppose that

$$N(A_1) = Y$$

$$N(A_2) = x_1 - Y = 125 - Y$$

In this situation, if we want to use MLE to get θ , we will have

$$\begin{aligned}
L(X, Y; \theta) &= \left(\frac{1}{2}\right)^Y \left(\frac{\theta}{4}\right)^{x_1 - Y} \left[\frac{1}{4}(1 - \theta)\right]^{x_2} \left[\frac{1}{4}(1 - \theta)\right]^{x_3} \left(\frac{\theta}{4}\right)^{x_4} \\
&\propto (1 - \theta)^{x_2 + x_3} (\theta)^{x_1 - Y + x_4}
\end{aligned}$$

We can not calculate θ through this formula. Then we need EM algorithm

$$\begin{aligned}
Q(\theta|\theta^{(i)}) &= E[\ln L(X, Y; \theta|X, \theta^{(i)})] \\
&= E[(x_1 - Y + x_4) \ln \theta + (x_2 + x_3) \ln(1 - \theta) | X, \theta^{(i)}] \\
&= (x_1 + x_4) \ln \theta + (x_2 + x_3) \ln(1 - \theta) - E(Y|X, \theta^{(i)}) \ln \theta
\end{aligned}$$

When given X and $\theta^{(i)}$, $(Y|X, \theta^{(i)}) \sim b(x_1, p_1)$, where $p_1 = \frac{\frac{1}{2}}{\frac{1}{2} + \frac{\theta^{(i)}}{4}} = \frac{2}{2 + \theta^{(i)}}$

$$E(Y|X, \theta^{(i)}) = np = x_1 p_1 = \frac{2x_1}{2 + \theta^{(i)}}$$

So

$$\begin{aligned}
Q(\theta|\theta^{(i)}) &= (x_1 + x_4) \ln \theta + (x_2 + x_3) \ln(1 - \theta) - E(Y|X, \theta^{(i)}) \ln \theta \\
&= (x_1 + x_4 - \frac{2x_1}{2 + \theta^{(i)}}) \ln \theta + (x_2 + x_3) \ln(1 - \theta) \\
&= (159 - \frac{250}{2 + \theta^{(i)}}) \ln \theta + 38 \ln(1 - \theta)
\end{aligned}$$

Then we calculate the maximum of $Q(\theta|\theta^{(i)})$

$$\begin{aligned}
Q'(\theta|\theta^{(i)}) &= (159 - \frac{250}{2 + \theta^{(i)}}) \frac{1}{\theta} - 38 \frac{1}{(1 - \theta)} = 0 \\
\Rightarrow \theta^{(i+1)} &= \frac{68 + 159\theta^{(i)}}{144 + 197\theta^{(i)}}
\end{aligned}$$

```

EM=function(theta.init,tol,count.max){
  theta=theta.init
  count=0
  for(i in 1:count.max){
    count=count+1
    theta[i+1]=(68+159*theta[i])/(144+197*theta[i])
    if(abs(theta[i+1]-theta[i])<tol&&count<count.max)break
  }
  theta
}
theta=EM(0.2,1e-5,1000)
count=length(theta)-1
theta

## [1] 0.2000000 0.5441658 0.6151352 0.6252563 0.6266134 0.6267939 0.6268178
## [8] 0.6268210

count

## [1] 7

```

MCCEM

We can also use Monte Carlo Method to do this EM algorithm.

$$\begin{aligned}
 Q^{(i+1)}(\theta|\theta^{(i)}) &= \frac{1}{m} \sum_{j=1}^m \ln L(X, Y; \theta) \\
 &= \frac{1}{m} \sum_{j=1}^m \ln(\theta^{x_1+x_2-Y} (1-\theta)^{x_2+x_3}) \\
 &= \frac{1}{m} \sum_{j=1}^m (x_1 + x_2 - Y) \ln \theta + (x_2 + x_3) \ln(1 - \theta) \\
 &= (x_1 + x_2 - E(Y)) \ln \theta + (x_2 + x_3) \ln(1 - \theta)
 \end{aligned}$$

$$\begin{aligned}
 Q'^{(i+1)}(\theta|\theta^{(i)}) &= \frac{(x_1 + x_2 - E(Y))}{\theta} - \frac{(x_2 + x_3)}{1 - \theta} = 0 \\
 \Rightarrow \theta^{(i+1)} &= \frac{x_1 + x_2 - E(Y)}{x_1 + x_2 + x_3 + x_4 - E(Y)} \\
 Y &\sim b(x_1, \frac{2}{2 + \theta^{(i)}})
 \end{aligned}$$

```

x1=125
x2=18
x3=20
x4=34
MCCEM=function(theta.init,m,tol,count.max){
  theta=theta.init
  count=0
  for(i in 1:count.max){
    count=count+1
    p1=2/(2+theta.init)
    Y=rbinom(m,x1,p1)

```

```

    theta[i+1]=(x1+x4-mean(Y))/(x1+x2+x3+x4-mean(Y))
    theta.init=theta[i+1]
    if(abs(theta[i+1]-theta[i])<tol&&count<count.max)break
  }
  theta
}
theta=MCEM(0.2,10000,1e-5,1000)
count=length(theta)-1
theta

## [1] 0.2000000 0.5442656 0.6150577 0.6254549 0.6266795 0.6266006 0.6266706
## [8] 0.6268877 0.6271542 0.6268059 0.6268653 0.6268708

count

## [1] 11

```

SQUAREM

```

x1=125
x2=18
x3=20
x4=34
loglik=function(theta){
  y=rbinom(1,x1,2/(2+theta))
  log((1-theta)^(x2+x3)*theta^(x1+x4-y))
}
em=function(theta){
  theta.new=(68+159*theta)/(144+197*theta)
  theta=theta.new
  return(theta)
}
library(SQUAREM)
theta0=0.2
result=squarem(c(theta=theta0), fixptfn=em, objfn=loglik, control=list(tol=1e-10,mstep=5))
theta=result$par
count=result$iter
theta

##      theta
## 0.6268215

count

## [1] 4

```

Compare

Let's choose a smaller tol to see the speed of the algorithms we talk above.

```

tol=1e-7
count.max=1000
theta.init=0.2

```

EM

```
t1=Sys.time()
theta_EM=EM(theta.init,tol,count.max)
t2=Sys.time()
iter_EM=length(theta_EM)-1
theta_EM

## [1] 0.2000000 0.5441658 0.6151352 0.6252563 0.6266134 0.6267939 0.6268178
## [8] 0.6268210 0.6268214 0.6268215

print(paste("iteration=",iter_EM))

## [1] "iteration= 9"

print(paste("time=",t2-t1))

## [1] "time= 0.00199389457702637"
```

MCEM

```
t3=Sys.time()
theta_MCEM=MCEM(theta.init,10000,tol,count.max)
t4=Sys.time()
iter_MCEM=length(theta_MCEM)-1
print("MCEM")

## [1] "MCEM"

theta_MCEM

## [1] 0.2000000 0.5443356 0.6149481 0.6251049 0.6266519 0.6267411 0.6268052
## [8] 0.6269305 0.6270411 0.6268877 0.6266134 0.6265342 0.6267616 0.6268668
## [15] 0.6271688 0.6266560 0.6268931 0.6268254 0.6269452 0.6268880 0.6267183
## [22] 0.6268495 0.6267044 0.6267044

print(paste("iteration=",iter_MCEM))

## [1] "iteration= 23"

print(paste("time=",t4-t3))

## [1] "time= 0.0339090824127197"
```

SQUAREM

```
t5=Sys.time()
result=squarem(c(theta=theta0), fixptfn=em, objfn=loglik, control=list(tol=1e-10,mstep=5))
t6=Sys.time()
theta_SQUAREM=result$par
iter_SQUAREM=result$iter
print("SQUAREM")

## [1] "SQUAREM"
```

```
theta_SQUAREM
```

```
##      theta  
## 0.6268215
```

```
print(paste("iteration=", iter_SQUAREM))
```

```
## [1] "iteration= 4"
```

```
print(paste("time=", t6-t5))
```

```
## [1] "time= 0.0029909610748291"
```