

# Gaussian EM Algorithm

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```
EM_normal<-function(data,mu0=c(0,0,0),E0=diag(c(1,1,1)),tol=1e-6){
  muhat<-mu0
  Ehat<-E0

  X<-data
  n<-nrow(X)

  mre<-10000 #modified relative error - something large/arbitrary to enter the while loop
  while(mre>tol){
    #predict

    T1<-0
    T2<-0

    for(j in 1:n){
      if(sum(is.na(X[j,]))==0){next}
      mis<-is.na(X[j,])
      x1<-X[j,mis] #missing components
      x2<-X[j,!mis] #not missing
      #we estimate x1 with the conditional expectation of x1 given x2
      muhat1<-muhat[mis]
      muhat2<-muhat[!mis]
      Ehat11<-Ehat[mis,mis]
      Ehat12<-Ehat[mis,!mis]
      Ehat21<-Ehat[!mis,mis]
      Ehat22<-Ehat[!mis,!mis]

      x1est<-muhat1+Ehat12%*%solve(Ehat22)%*%(x2-muhat2)

      x<-X[j,]
      x[mis]<-x1est
      T1<-T1+x #estimated complete data contribution to the sufficient statistic T1

      x1x1Test<-Ehat11-Ehat12%*%solve(Ehat22)%*%Ehat21+x1est%*%t(x1est)
      xxT<-X[j,]%*%t(X[j,])
      xxT[mis,mis]<-x1x1Test
      xxT[!mis,mis]<-x2%*%t(x1est)
      xxT[mis,!mis]<-x1est%*%t(x2)

      T2<-T2+xxT #estimated complete data contribution to the sufficient statistic T2
    }
    #save initial estimate to compute mre
    muhat_old<-muhat
    Ehat_old<-Ehat

    #estimate the new
    muhat<- 1/n*T1
```

```

Ehat<-1/n*T2-muhat%*%t(muhat)

#calculate mre
par_old<-c(muhat_old,Ehat_old[lower.tri(Ehat_old,diag=TRUE)])
par_new<-c(muhat,Ehat[lower.tri(Ehat,diag=TRUE)])
mre<-sqrt(sum((par_old-par_new)^2))/max(1,sqrt(sum(par_new^2)))
}
return(list(muhat,Ehat))
}

```

Now let's test our EM-algorithm on some sample data:

```

X<-matrix(c(3,6,0,4,4,3,NA,8,3,5,NA,NA),4,byrow=TRUE)

```

For our initial estimate of  $\mu$ , we'll compute sample estimates with the data points we have. For our initial estimate of  $\Sigma$ , we'll impute the missing data with the respective initial estimate of  $\mu$ .

```

mu0<-apply(X,2,mean,na.rm=TRUE)
Xtemp<-X
Xtemp[3,1]<-mu0[1]
Xtemp[4,2]<-mu0[2]
Xtemp[4,3]<-mu0[3]
E0<-cov(Xtemp)
EM_normal(X,mu0,E0)

```

```

## [[1]]
## [1] 2.489376 4.017141 1.506429
##
## [[2]]
##      [,1]      [,2]      [,3]
## [1,]  6.197220 10.000536  3.750204
## [2,] 10.000536 16.138010  6.051758
## [3,]  3.750204  6.051758  2.269411

```