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classification_heterogeneity

Description

This function classify data under the distribution heterogeneity in one class by the proposed classification method. The inputs are the historical data of two classes in each situation, the designed error rates of two classes in each situation, and the data to be classified. The outputs are the thresholds t_0, t_1 and the classifier $\eta(x)$.

Inputs

XM_1 historical data of class 1 in situation 1
XF_1 historical data of class 1 in situation 2
XM_0 historical data of class 0 in situation 1
XF_0 historical data of class 0 in situation 2
Xtest data to be classified
alpha11 designed classification error rate of class 1 in situation 1
alpha21 designed classification error rate of class 1 in situation 2
alpha10 designed classification error rate of class 0 in situation 1
alpha20 designed classification error rate of class 0 in situation 2

Outputs

t0 ,t1 the thresholds of the classification
eta the classifier of the classification

Function

```
classification_heterogeneity<-function(XM_1,XM_0,XF_1,XF_0,Xtest, alpha11,  
alpha10, alpha21, alpha20){
```

```
X1<-c(XM_1,XF_1)  
X0<-c(XM_0,XF_0)
```

```
n11<-length(XM_1)  
n10<-length(XM_0)  
n21<-length(XF_1)  
n20<-length(XF_0)  
n1<-n11+n21  
n0<-n10+n20  
nt<-length(Xtest)
```

```
hM_1<-bw.ucv(XM_1)  
hM_0<-bw.ucv(XM_0)  
hF_1<-bw.ucv(XF_1)  
hF_0<-bw.ucv(XF_0)  
h1<-bw.ucv(X1)  
h0<-bw.ucv(X0)
```

```
#####situation 1#####
```

```

#####estimate t1#####

aM1<-matrix(0 ,nrow=n11 ,ncol=n11)
bM1<-matrix(0 ,nrow=n11 ,ncol=n10)
aaM1<-c()
bbM1<-c()
etaM1<-c()
for ( i in 1:( n11 )){
for ( j in 1:( n11 )){
for ( k in 1:( n10 ))
{
aM1[ i , j]<-1/(n11*hM_1)*exp((-1/2)*((XM_1[ i ]-XM_1[ j ])/hM_1)^2)
bM1[ i , k]<-1/(n10*hM_0)*exp((-1/2)*((XM_1[ i ]-XM_0[ k ])/hM_0)^2)
}
}
}

aaM1<-rowSums(aM1)
bbM1<-rowSums(bM1)
etaM1<-aaM1/bbM1
tM1<-sort(etaM1)[( n11)*( alpha11 )]

#####estimate t0#####

aM0<-matrix(0 ,nrow=n10 ,ncol=n11)
bM0<-matrix(0 ,nrow=n10 ,ncol=n10)
aaM0<-c()
bbM0<-c()
etaM0<-c()
for ( i in 1:( n10 )){
for ( j in 1:( n11 )){
for ( k in 1:( n10 ))
{
aM0[ i , j]<-1/(n11*hM_1)*exp((-1/2)*((XM_0[ i ]-XM_1[ j ])/hM_1)^2)
bM0[ i , k]<-1/(n10*hM_0)*exp((-1/2)*((XM_0[ i ]-XM_0[ k ])/hM_0)^2)
}
}
}

aaM0<-rowSums(aM0)
bbM0<-rowSums(bM0)
etaM0<-aaM0/bbM0
tM0<-sort(etaM0,decreasing=T)[( n10)*( alpha10 )]

#####situation 2#####
#####estimate t1#####

aF1<-matrix(0 ,nrow=n21 ,ncol=n21)

```

```

bF1<-matrix(0 ,nrow=n21 ,ncol=n20)
aaF1<-c()
bbF1<-c()
etaF1<-c()
for (i in 1:(n21)){
for (j in 1:(n21)){
for (k in 1:(n20))
{
aF1[i,j]<-1/(n21*hF_1)*exp((-1/2)*((XF_1[i]-XF_1[j])/hF_1)^2)
bF1[i,k]<-1/(n20*hF_0)*exp((-1/2)*((XF_1[i]-XF_0[k])/hF_0)^2)
}
}
}

aaF1<-rowSums(aF1)
bbF1<-rowSums(bF1)
etaF1<-aaF1/bbF1
tF1<-sort(etaF1)[(n21)*(alpha21)]

#####estimate t0#####

aF0<-matrix(0 ,nrow=n20 ,ncol=n21)
bF0<-matrix(0 ,nrow=n20 ,ncol=n20)
aaF0<-c()
bbF0<-c()
etaF0<-c()
for (i in 1:(n20)){
for (j in 1:(n21)){
for (k in 1:(n20))
{
aF0[i,j]<-1/(n21*hF_1)*exp((-1/2)*((XF_0[i]-XF_1[j])/hF_1)^2)
bF0[i,k]<-1/(n20*hF_0)*exp((-1/2)*((XF_0[i]-XF_0[k])/hF_0)^2)
}
}
}

aaF0<-rowSums(aF0)
bbF0<-rowSums(bF0)
etaF0<-aaF0/bbF0
tF0<-sort(etaF0,decreasing=T)[(n20)*(alpha20)]

#####choose final boundary#####
t1<-min(tM1,tF1)
t0<-max(tM0,tF0)

#####construct the classifier eta(x)#####
at<-matrix(0 ,nrow=nt ,ncol=n1)

```

```

bt<-matrix(0 , nrow=nt , ncol=n0)
aat<-c()
bbt<-c()
eta<-c()

for (i in 1:(nt)){
for (j in 1:(n1)){
for (k in 1:(n0))
{
at [i , j]<-1/(n1*h1)*exp((-1/2)*((Xtest [i]-X1 [j])/h1)^2)
bt [i , k]<-1/(n0*h0)*exp((-1/2)*((Xtest [i]-X0 [k])/h0)^2)
}
}
}

aat<-rowSums(at)
bbt<-rowSums(bt)
eta<-aat/bbt

result = list(t0=t0 , t1=t1 , eta=eta)
return(result)
}

```

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classification_homogeneity

Description

This function classify data by the method regarding data in one class follow homogeneous distribution. The inputs are the historical data, the designed error rates of two classes and the data to be classified. The outputs are the thresholds t_0 , t_1 and the classifier $\eta(x)$.

Inputs

X1	historical data of class 1
X0	historical data of class 0
Xtest	data to be classified
alpha1	designed classification error rate of class 1
alpha0	designed classification error rate of class 0

Outputs

t0 , t1	the thresholds of the classification
eta	the classifier of the classification

Function

```

classification_homogeneity<-function(X1,X0,Xtest , alpha1 , alpha0){

n1<-length(X1)
n0<-length(X0)

```

```

nt<-length ( Xtest )

h1<-bw . ucv ( X1 )
h0<-bw . ucv ( X0 )

#####estimate t1#####

a1<-matrix ( 0 , nrow=n1 , ncol=n1 )
b1<-matrix ( 0 , nrow=n1 , ncol=n0 )
aa1<-c ()
bb1<-c ()
eta1<-c ()
for ( i in 1:( n1 )){
for ( j in 1:( n1 )){
for ( k in 1:( n0 )) {
{
a1 [ i , j ]<-1/((n1)*h1)*exp((-1/2)*((X1 [ i ]-X1 [ j ])/h1)^2)
b1 [ i , k ]<-1/((n0)*h0)*exp((-1/2)*((X1 [ i ]-X0 [ k ])/h0)^2)
}
}
}
}

aa1<-rowSums ( a1 )
bb1<-rowSums ( b1 )
eta1<-aa1/bb1
t1<-sort ( eta1 )[( n1)*( alpha1 )]

#####estimate t0#####

a0<-matrix ( 0 , nrow=n0 , ncol=n1 )
b0<-matrix ( 0 , nrow=n0 , ncol=n0 )
aa0<-c ()
bb0<-c ()
eta0<-c ()
for ( i in 1:( n0 )){
for ( j in 1:( n1 )){
for ( k in 1:( n0 )) {
{
a0 [ i , j ]<-1/((n1)*h1)*exp((-1/2)*((X0 [ i ]-X1 [ j ])/h1)^2)
b0 [ i , k ]<-1/((n0)*h0)*exp((-1/2)*((X0 [ i ]-X0 [ k ])/h0)^2)
}
}
}
}

aa0<-rowSums ( a0 )
bb0<-rowSums ( b0 )
eta0<-aa0/bb0
t0<-sort ( eta0 , decreasing=T )[( n0)*( alpha0 )]

```

```

#####construct the classifier eta(x)#####
at<-matrix(0,nrow=nt,ncol=n1)
bt<-matrix(0,nrow=nt,ncol=n0)
aat<-c()
bbt<-c()
eta<-c()

for (i in 1:(nt)){
for (j in 1:(n1)){
for (k in 1:(n0))
{
at[i,j]<-1/(n1*h1)*exp((-1/2)*((Xtest[i]-X1[j])/h1)^2)
bt[i,k]<-1/(n0*h0)*exp((-1/2)*((Xtest[i]-X0[k])/h0)^2)
}
}
}

aat<-rowSums(at)
bbt<-rowSums(bt)
eta<-aat/bbt

result = list(t0=t0,t1=t1,eta=eta)
return(result)
}

```

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classification run.R

Description

There are some examples in this R file to run the above functions, both the simulation data and the real data can be realized by the above two functions.