

## nQTL instruction

### A、 Data preparation:

#### 1、 Require data:

SNP data matrix, gene expression matrix, gene symbol list, PPI net.

#### 2、 Data format:

SNP data matrix and gene expression matrix should have equal number of columns representing the same samples; gene symbol list should be corresponding to the feature/row id of gene expression matrix, e.g. the first list is feature id, the second list is related gene symbol; PPI net contains list of gene-pairs, each row represent one gene-pair including two gene symbols.

### B、 Code usage:

#### 1、 Load R packages:

```
Library(MatrixEQTL)
library(R.matlab)
```

#### 2、 Parameter setting:

When SNP data is continuous, the modelLINEAR is used; and when SNP data is discrete, the modelANOVA is used.

```
useModel1= modelANOVA;
useModel2= modelLINEAR;
pvOutputThreshold1 = 1e-100;
pvOutputThreshold2 = 1e-45;
```

#### 3、 Read data input:

```
exp<-read.csv(file = " /example/exp.csv",row.names = 1)
snpc<-read.csv(file=" /example/snp.csv",row.names = 1)
expname<-read.csv(file = " /example/genename.csv")
ppi<-read.csv("/expample/String_9606_v10.5_s900_new.csv",header= F)
```

The input data format is as follows:

```
> exp[1:5,1:5]
           s7      s41      s23      s12      s1
ENSG00000000003 -0.04575573 -0.035124766 -0.004217641 -0.03172192 -0.04700507
ENSG000000000419 -0.01379910 -0.002979931 -0.005079183  0.06630001 -0.00008860
ENSG000000000457 -0.05069286 -0.016010936  0.009574341  0.02545820 -0.01748970
ENSG000000000460  0.02297084 -0.025389703 -0.006388585 -0.04028415  0.01960859
ENSG000000000938 -0.14934426  0.048292682 -0.026677878  0.06876903  0.02196986
> snpc[1:5,1:5]
           s7 s41 s23 s12 s1
rs61769350  2  2  2  2  2
rs4951859   2  2  2  2  2
rs142557973 2  2  2  2  2
rs141242758 2  2  2  2  2
rs79010578  2  2  2  2  2
> head(expname)
  Ensembl.Gene.ID Gene.Symbol
1 ENSG00000000003    TSPAN6
2 ENSG000000000419    DPM1
3 ENSG000000000457    SCYL3
4 ENSG000000000460    C1orf112
5 ENSG000000000938    FGR
6 ENSG000000000971    CFH
```

#### 4、 Data format adaption:

```
expdata<-apply(exp,2,as.numeric)
```

```
colnames(expdata)<-NULL
```

```
snpdata<-apply(snp,2,as.numeric)
```

```
colnames(snpdata)<-NULL
```

The intra data format is as follows:

```
> expdata[1:5,1:5]
      [,1]      [,2]      [,3]      [,4]      [,5]
[1,] -0.04575573 -0.035124766 -0.004217641 -0.03172192 -0.04700507
[2,] -0.01379910 -0.002979931 -0.005079183  0.06630001 -0.00008860
[3,] -0.05069286 -0.016010936  0.009574341  0.02545820 -0.01748970
[4,]  0.02297084 -0.025389703 -0.006388585 -0.04028415  0.01960859
[5,] -0.14934426  0.048292682 -0.026677878  0.06876903  0.02196986
> snpdata[1:5,1:5]
      [,1] [,2] [,3] [,4] [,5]
[1,]    2    2    2    2    2
[2,]    2    2    2    2    2
[3,]    2    2    2    2    2
[4,]    2    2    2    2    2
[5,]    2    2    2    2    2
```

#### 5、 Edge co-expression calculation:

```
exp$mean<-apply(exp, 1, mean)
```

```
myfuntion1<-function(ppi,expname){
```

```
if (length(which(expname[2]==as.character(ppi[1]))[1])==0){
```

```
    x<-0
```

```
    }else
```

```
        x<-which(expname[2]==as.character(ppi[1]))[1]
```

```
}
```

```
x<-apply(ppi, 1, myfuntion1,expname)
```

```
    myfuntion2<-function(ppi,expname){
```

```
if (length(which(expname[2]==as.character(ppi[2]))[1])==0) {
```

```
    x<-0
```

```
    }else
```

```
        x<-which(expname[2]==as.character(ppi[2]))[1]
```

```
}
```

```
y<-apply(ppi,1, myfuntion2,expname)
```

```
expsiteppi<-data.frame(cbind(x,y))
```

```
expsiteppi<-na.omit(expsiteppi)
```

```
save(expsiteppi,file="/example/expsiteppi.RData")
```

```
c<-dim(exp)[2]
```

```
pair<-data.frame()
```

```
for (s in 1:(dim(exp)[2]-1)){
```

```
    for (i in 1:dim(expsiteppi)[1]){
```

```
        pair[i,s]<-(exp[expsiteppi[i,1],s]-exp[expsiteppi[i,1],c])*
                    (exp[expsiteppi[i,2],1]-exp[expsiteppi[i,2],c])
```

```

    }
}
save(pair,file="/example/pair.RData")

```

## 6、 nQTL calculation:

```

snps1 = SlicedData$new( snpdata );
gene1 = SlicedData$new( pair );
cvrt1 = SlicedData$new( );
snps1$ResliceCombined(500);
gene1$ResliceCombined(500);
filename = tempfile();
errorCovariance = numeric();
meh = Matrix_eQTL_main( snps = snps1, gene = gene1, cvrt = cvrt1, output_file_name =
filename, pvOutputThreshold = pvOutputThreshold1, useModel = useModel1,
errorCovariance = errorCovariance, verbose = TRUE, pvalue.hist = 100);
unlink( filename );
# png(filename = "histogram.png", width = 650, height = 650)
plot(meh, col="grey")
# dev.off();

# a Q-Q plot
meq = Matrix_eQTL_main( snps = snps1, gene = gene1, cvrt = cvrt1, output_file_name =
filename, pvOutputThreshold = pvOutputThreshold2, useModel = useModel1,
errorCovariance = errorCovariance, verbose = TRUE, pvalue.hist = "qqplot");
unlink( filename );
# png(filename = "QQplot.png", width = 650, height = 650)
plot(meq, pch = 16, cex = 0.7)
# dev.off();

```

## C、 Demo experiment

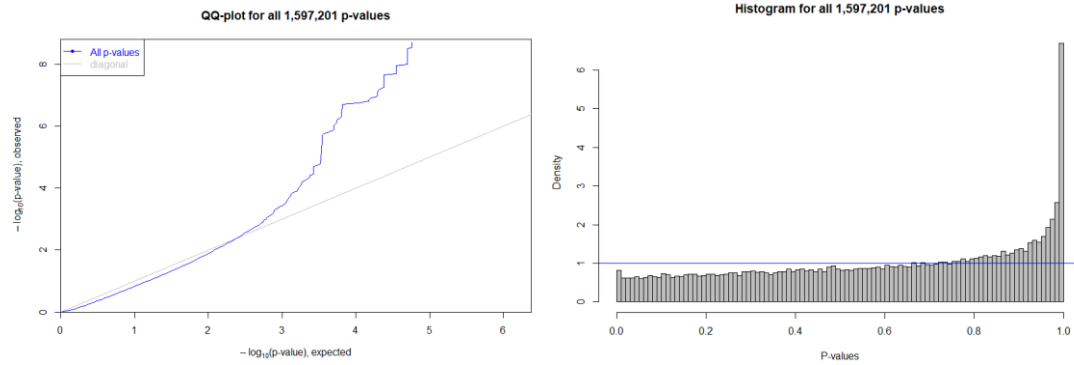
### 1、 Demo data:

Size of SNP data: 1999\*10

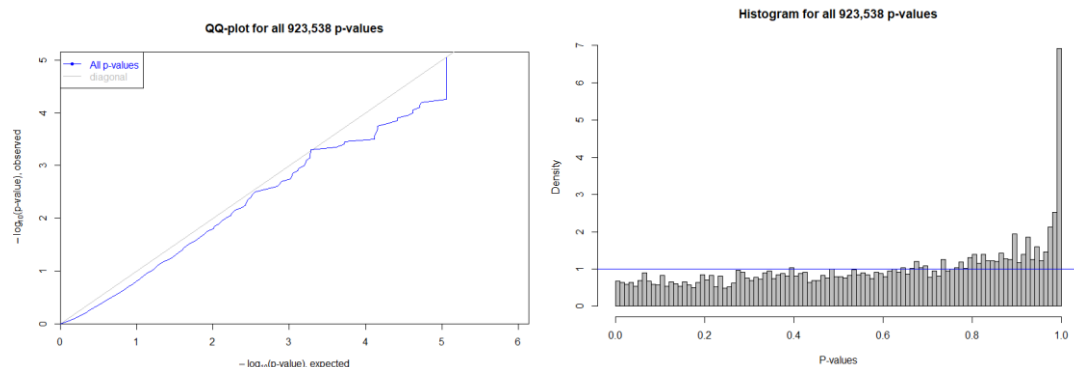
Size of expression data: 799\*10

Size of co-expression data: 462\*10 (ppi>900)

### 2、 eQTL outcomes:



### 3、nQTL outcomes:



## D、List of software used in down-stream analysis of nQTL framework

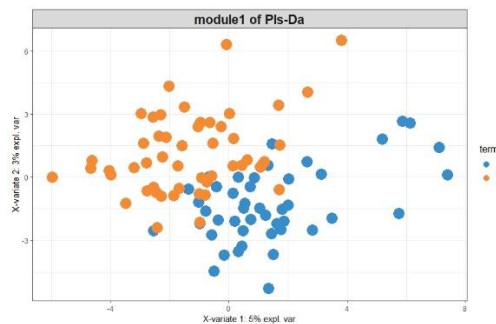
### 1、PLSDA

```
library(mixOmics)
#set working path(setwd(working path))
exp<-read.csv("./expdata.csv",row.names = 1)
#Delete outliers
exp<-exp[,-24]
module<-read.csv("./modulegene.csv")
samplotype<-read.csv("./samplotype.csv")
samplotype<-samplotype[samplotype$X.Sample_title%in%colnames(exp),]
exp<-as.data.frame(t(exp))
exp$term<-samplotype$term
exp<-exp[order(exp$term),]
a <- dim(exp)[2]
exp<-as.data.frame(t(exp[,-a]))

exp$term<-samplotype$type.cancer
exp<-exp[order(exp$term),]
exp<-as.data.frame(t(exp[,-a]))

data<-exp[rownames(exp)%in%module$cluster1,]
data<-apply(data,2,as.numeric)
XXt<-t(scale(data))
```

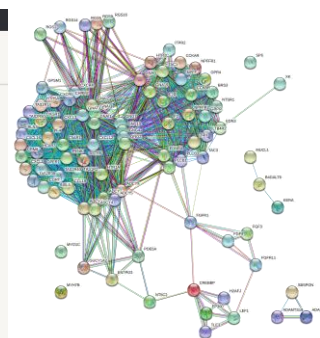
```
YY <- c(rep(c("adenocarcinoma", "large cell carcinoma", "squamous cell carcinoma"), c(50, 20, 26)))
YY <- c(rep(c("L", "S"), c(45, 51))) #names(datat)
plsda.datatm <- plsda(XXt, YY, ncomp = 2)
n <- dim(exp)[1]
plsda.datatm$names$sample <- c(1:n)
plotIndiv(plsda.datatm, ind.names = F, pch = 16, plot.ellipse = TRUE, title = "module1 of Pls-Da", legend = T, legend.title = "term", cex = 6, add.legend = TRUE, style = "ggplot2")
a <- vip(plsda.datatm)
```



## 2、PPI network and module

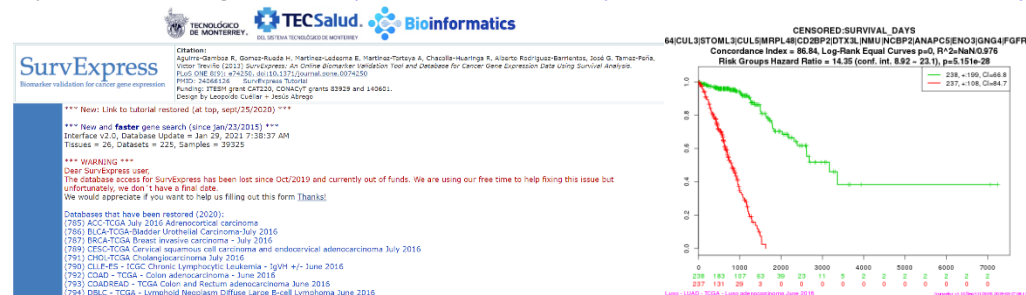
Input the module genes to

[https://string-db.org/cgi/input?sessionId=bUDUQJ46kRyH&input\\_page\\_active\\_form=multiple\\_id\\_entifiers](https://string-db.org/cgi/input?sessionId=bUDUQJ46kRyH&input_page_active_form=multiple_id_entifiers)



## 3、Survival analysis

Input the module genes to <http://bioinformatica.mty.itesm.mx:8080/Biomatec/SurvivaX.jsp>



## 4、CCA

```
##set working path(setwd(working path))
```

```

library(vegan)
load("./pair.RData")
load("./exp.RData")
genename<-read.csv("./genename.csv")
modulepair<-read.csv("./geneorder.csv")
inectgene<-c('IFITM1')
#Select relevant factors : 'ISG15', 'IFI35', 'IFIT5', 'IFIT3', 'ISG20', 'HERC6', 'IFIT2', 'IFI6', 'TNF'
inectgene<-genename[genename$Gene.Symbol%in%inectgene,]
inectexp<-exp[rownames(exp)%in%inectgene$Ensembl.Gene.ID,]
rownames(inectexp)<-inectgene[inectgene$Ensembl.Gene.ID%in%rownames(inectexp),2]
inectexp<-as.data.frame(t(inectexp))
module<-modulepair[modulepair$ClassGene=='Cluster9',]#change different module
modulepair<-as.data.frame(t(pair[rownames(pair)%in%module$X,]))

sumpair<-apply(modulepair, 1, sum)
modulepair<-modulepair[-(which(sumpair<0.05)),]
inectexp<-inectexp[rownames(inectexp)%in%rownames(modulepair),]

modulecca<-cca(modulepair,inectexp)
plot(modulecca)
permutest(modulecca,permu=999)
ef<-envfit(modulecca,inectexp,permu=999)

you need to record data into tables to plot
#heatmap
library(ComplexHeatmap)
data1<-read.csv("./cca.csv",header = T,stringsAsFactors = F,row.names = 1)
col = c( "significant" = "#33A02C", "single-significant" = "#E31A1C")
alter_fun = function(x, y, w, h, v) {
  n=sum(v)
  h=h*0.9
  grid.rect(x, y, w-unit(0.5, "mm"), h-unit(0.5, "mm"), gp = gpar(fill = "#CCCCCC", col = NA))
  if(v["single-significant"]) grid.rect(x, y - h*0.5 + 0.95:n/n*h, w*1, 1/n*h, gp = gpar(fill =
col[names(which(v))], col = NA), just = "top")
  if(v["significant"]) grid.rect(x, y - h*0.5 + 0.95:n/n*h, w*1, 1/n*h, gp = gpar(fill =
col[names(which(v))], col = NA), just = "top")
}
oncoPrint(data1, get_type = function(x) strsplit(x, ";")[[1]],
  alter_fun = alter_fun, col = col, row_order = NULL,
  column_order = colnames(data1), show_column_names = TRUE,
  heatmap_legend_param = list(title = "Alternations",
at = c("significant", "single-significant"),

```

labels = c("significant", "single-significant"))

