Integrative cross-omics and cross-context analysis elucidates molecular links underlying genetic effects on complex traits

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Introduction

This vignette provides an introduction to the X.ING package. The R package X.ING implements the X-ING method for cross-omics and cross-context analysis.

To install the development version of the X.ING package, please load the devtools package first. Note that X.ING requires the CCA, RGCCA, Rcpp, and RcppArmadillo packages.

To install this package, run the following command in R

```
library(devtools)
install_github("ylustat/X.ING")
```

Load the package using the following command:

```
library("X.ING")
```

```
##
## Attaching package: 'X.ING'
## The following objects are masked from 'package:base':
##
## scale, unique
```

This vignette depends on R packages MendelianRandomization, ggplot2 and tidyverse, you can load these packages using the following command:

```
suppressMessages(library("tidyverse"))
suppressMessages(library("pROC"))
suppressMessages(library("MASS"))
library("ggplot2")
library("MendelianRandomization")
library("RGCCA")
```

Fit XING using simulated data

The main function of the X.ING package is XING(). In this section, we fit XING function using simulated data (provided in the package) as an example to illustrate the basic usage of XING. The example data is a list with two elements, where z is a list of z-scores and gamma is a list of true underlying association status. 1 in gamma indicates the presence of non-zero effect and 0 indicates the absence of effect. We have three sets of data available in the example data (L=3). The number of contexts are 10, 8, and 6, respectively.

```
data("example_data")
length(example$z)
```

```
## [1] 3
```

```
sapply(example$z, dim)
## [,1] [,2] [,3]
## [1,] 6000 6000 6000
```

The proportion of true associations are ~ 0.04 .

```
sapply(example$gamma, mean)
```

```
## [1] 0.04265000 0.03893750 0.03922222
```

The X-ING takes the z-score list as input. It automatically use CCA (when L=2) for GCCA (when L>2) for the multi-view learning. Afterwards, it applies PCA to extract shared patterns across contexts for each omics. A simplest application where two sets of z-scores are analyzed:

```
z <- example$z
res <- XING(z_list = z[1:2])</pre>
```

The output includes posterior probabilities (alpha) and posterior means (mu):

```
names(res)
```

[2,]

10

```
## [1] "mu" "alpha"
lengths(res)
```

```
## mu alpha
## 2 2
```

AUC can be calculated based on the posterior probability and true association status

```
gamma <- example$gamma
mapply(function(g,pp) auc(c(g),c(pp), quiet = T), gamma[1:2], res$alpha)</pre>
```

```
## [1] 0.8548316 0.8457119
```

It switches to GCCA (implemented in RGCCA package) if L>2 (note that the computation speed may substantially slowed):

```
z <- example$z
res <- XING(z_list = z)</pre>
```

Similar to the previous example, the output also includes posterior probabilities (alpha) and posterior means (mu):

```
names(res)
```

```
## [1] "mu" "alpha" lengths(res)
```

```
## mu alpha
## 3 3
```

And the corresponding AUCs:

```
mapply(function(g,pp) auc(c(g),c(pp), quiet = T), gamma, res$alpha)
```

```
## [1] 0.8953075 0.8766113 0.8168692
```

Users may specify the sample overlap matrix \hat{R} through the Lambda_list parameter. The number of canonical coefficients to keep, and the number of PCs can be specified by CC and PC_list. For example,

```
res <- XING(z_list = z, CC = 4, PC_list = c(4,3,2), Lambda_list = lapply(sapply(z,ncol),diag)) mapply(function(g,pp) auc(c(g),c(pp), quiet = T), gamma, resalpha)
```

[1] 0.8249574 0.8323304 0.8175283

To reduce computation time, user may specify the iteration time via iterT. Also, the tolerance parameter controls the criterion for convergence of the GCCA algorithm. A smaller tolerance may lead to substantially longer computation time.