Package ‘AIMS’

March 22, 2017

Type Package

Title AIMS : Absolute Assignment of Breast Cancer Intrinsic Molecular Subtype

Version 1.6.0

Date 2014-06-25

Description This package contains the AIMS implementation. It contains necessary functions to assign the five intrinsic molecular subtypes (Luminal A, Luminal B, Her2-enriched, Basal-like, Normal-like). Assignments could be done on individual samples as well as on dataset of gene expression data.

Author Eric R. Paquet, Michael T. Hallett

Maintainer Eric R Paquet <eric.r.paquet@gmail.com>

biocViews Classification, RNASeq, Microarray, Software, GeneExpression

Depends R (>= 2.10), e1071, Biobase

Suggests breastCancerVDX, hgu133a.db, RUnit, BiocGenerics

License Artistic-2.0

URL http://www.bci.mcgill.ca/AIMS

LazyData yes

Installable Everywhere yes

NeedsCompilation no

R topics documented:

AIMSmodel ................................................................. 2
applyAIMS ................................................................. 3
mcgillExample ............................................................ 5

Index 6
The AIMS model

Description

This is the model definition for AIMS. It contains the naive bayes classifier composed of the 100 rules described in Paquet et al. "Absolute assignment of breast cancer intrinsic molecular subtype" (under review at JNCI).

Usage

AIMSmodel

Details

This is the AIMS model define using 100 simple rules of the form gene A < gene B and combine within a naive bayes classifier within e1071. (Paquet et al. under review JNCI)

Briefly, using a suitably large training dataset (~5000 gene breast cancer gene expression profiles), the approach identifies a small set of simple binary rules (~20) that examine the raw expression measurements for pairs of genes from a single breast cancer patient, and only that patient. The binary rules are of the form "if the expression of gene x is greater than gene y, then tend to assign subtype z for that patient". Subtypes could be: Basal, Her2, LumA, LumB, or Normal. The collection of binary rules is combined for a single estimation of a patient subtype via a single probabilistic model using naiveBayes in e1071. In this way, since only expression levels of genes with a single patient is considered, the method represents a promising approach to ablate the instability caused by relativistic approaches (Paquet et al. in review at JNCI).

Value

all.pairs  The 100 rules in AIMS in the form EntrezID gene A < EntrezID gene B
k          The selected number of optimal rules. For AIMS we have shown it is 20.
one.vs.all.tsp The Naive bayes classifier used in combination with the 100 rules
selected.pairs.list The list of rules sorted from the best discriminating rule to the least discriminating rules subdivided by subtype.

Author(s)

Eric R. Paquet (eric.r.paquet@gmail.com)

See Also

applyAIMS, mcgillExample, naiveBayes

Examples

```r
## Load a sample of the McGill dataset used in the paper
data(AIMSmodel)

## list the top-scoring rules for the individual subtypes
AIMSmodel$selected.pairs.list
```
applyAIMS

## List the posterior probability tables for the 100-rules
AIMSmodel$one.vs.all.tsp[[20]]$table

---

**applyAIMS**

*Function that will assign the AIMS subtype to a gene expression matrix*

**Description**

Given a gene expression matrix \( D \) where rows correspond to genes and columns to samples and a list of Entrez gene ids, this function will assign the breast cancer molecular subtype using Absolute PAM50 (AIMS) (Paquet et al. under review at JNCI).

**Usage**

```
applyAIMS (eset, EntrezID)
```

**Arguments**

- **eset**
  - An ExpressionSet (Biobase) or a matrix \( D \) of gene expression. Rows are genes and columns are samples (tumor from breast cancer patients).

- **EntrezID**
  - A character vector corresponding to the Entrez Ids of the genes of matrix \( D \).

**Details**

We defined Absolute assignment of breast cancer Molecular Intrinsic Subtypes (AIMS) to stabilize the actual PAM50. The idea of the approach is to use simple rules of the form "if gene A expression < gene B expression assigned sample to subtype X". By using those simple rules we are able to assign subtype using only the expression values of one patient. We have shown AIMS recapitulates PAM50 subtype assignments and preserved the prognostic values of the subtypes. This function will return the subtype assignment as well as the posterior probabilities for all the subtypes.

**Value**

- **cl**
  - Subtypes identified by the AIMS. It could be either "Basal", "Her2", "LumA", "LumB" or "Normal".

- **prob**
  - A vector corresponding to the posterior probabilities of the subtypes in cl.

- **all.probs**
  - A matrix of all the posterior probabilities for all the samples and all the subtypes

- **rules.matrix**
  - The matrix of 100 rules used by AIMS to assign the breast cancer subtypes. Rows correspond to rules and columns to samples. This is a 0,1 matrix in which 1 represents the rules is true and 0 the rule is false

- **data.used**
  - The expression values used to evaluate the simple rules.

- **EntrezID.used**
  - The list of Entrez ids used by AIMS

**Author(s)**

Eric R. Paquet (eric.r.paquet@gmail.com)
applyAIMS

References


See Also

AIMSmodel, mcgillExample

Examples

## Load the McGill dataset used in the paper
data(mcgillExample)

## Convert the expression matrix to an ExpressionSet.
## could also send directly an expression matrix
mcgillExample$D <- ExpressionSet(assayData=mcgillExample$D)

## Assign AIMS on McGill dataset
mcgill.AIMS.subtypes.batch <- applyAIMS (mcgillExample$D, mcgillExample$EntrezID)

## Print a summary of all the subtype in the dataset
table(mcgill.AIMS.subtypes.batch$cl)

## We can do the samething using only one sample
mcgill.AIMS.subtypes.first <- applyAIMS (mcgillExample$D[,1,drop=FALSE], mcgillExample$EntrezID)
table(mcgill.AIMS.subtypes.first$cl)

if (mcgill.AIMS.subtypes.batch$cl[1] == mcgill.AIMS.subtypes.first$cl[1]){
  message("Identical assignment batch and first sample")
} else{
  message("Different assignment batch and first sample")
}

## We can do the samething for the first 20 samples
mcgill.AIMS.subtypes.first20 <- applyAIMS (mcgillExample$D[,1:20,drop=FALSE], mcgillExample$EntrezID)
table(mcgill.AIMS.subtypes.first20$cl)

if (all(mcgill.AIMS.subtypes.batch$cl[1:20] == mcgill.AIMS.subtypes.first20$cl)){
  message("Identical assignment batch and first 20")
} else{
  message("Different assignment batch and first 20")
}

## We can do the samething using 50 randomly selected samples, no
## set.seed needed
```
mcgillExample $D = \text{sample}(1: \text{ncol}(\text{mcgillExample}$D$), 50)$

mcgill.AIMS.subtypes.rand50 <- applyAIMS (mcgillExample$D[, sel.rand.50, drop=FALSE],
  mcgillExample$EntrezID)

\text{table(mcgill.AIMS.subtypes.rand50$cl)}$

if (\text{all(mcgill.AIMS.subtypes.batch$cl[sel.rand.50] == mcgill.AIMS.subtypes.rand50$cl})){
  \text{message("Identical assignment batch and random 50")}
}else{
  \text{message("Different assignment batch and random 50")}
}
```

---

**mcgillExample**

*Sample of a breast cancer gene expression dataset generated at McGill University*

---

**Description**

This is a breast cancer gene expression dataset generated at McGill University using the Affymetrix Gene ST platform. This is the validation dataset used in AIMS' paper and in use in the examples of AIMS.

**Usage**

mcgillExample

**Details**

A sample of a breast cancer gene expression dataset currently generated at McGill University on the Affymetrix gene ST platform.

**Value**

- **D**: A matrix of gene gene expression, rows are genes and columns are samples
- **EntrezID**: A vector of Entrez ids corresponding to genes (rows) of D

**Author(s)**

Eric R. Paquet (eric.r.paquet@gmail.com)

**See Also**

AIMSmodel, applyAIMS

**Examples**

```r
## Load a sample of the McGill dataset used in the paper
data(mcgillExample)

## Dimensions of the gene expression matrix
dim(mcgillExample$D)

## Number of EntrezID
length(mcgillExample$EntrezID)
```
Index

*Topic classif
  applyAIMS, 3

*Topic datasets
  mcgillExample, 5

*Topic models
  AIMSmodel, 2

AIMSmodel, 2, 4, 5
applyAIMS, 2, 3, 5
mcgillExample, 2, 4, 5
naiveBayes, 2