# Package 'BayRepulsive'

October 7, 2018

Title BayRepulsive: A Bayesian Repulsive Deconvolution Model				
<ul> <li>Date 2018-10-07</li> <li>Depends mvtnorm, alabama, psych, optimx</li> <li>Author Yuliang Li and Yanxun Xu</li> <li>Maintainer Yuliang Li <yli193@jhu.edu></yli193@jhu.edu></li> <li>Description A Non-negative Matrix Factorization (NMF) with repulsiveness introduced by determinantal point process (DPP).</li> </ul>				
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BayRepulsive_unknown				
BayRepulsive_known is a deconvolution function designed for in- ferring tumor heterogeneity, used when the number of subclones is				
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# Description

Takes in the observed data matrix, the number of subclones, the number of features and samples, gives the estiamted NMF results.

# Usage

#### **Arguments**

Datause The observed data matrix. Each row is a sample.

K The number of subclones

Nobs The number of samples, i.e., the number of rows of the Datause

Nfeature The number of features, i.e., the number of columns of the Datause

Niter The number of maximum iterations

epsilon break if the L2 distance of the two estiamted proportion matrix in row is less

than epsilon

tau The hyperparameter for DPP, a large number is prefered, default value is 100

seed The random seed, default as 1

#### **Details**

Given an observed matrix, whose rows are mixed samples of unknown number of subclones, returns the results of NMF.

This function will create a bunch of globel variables, named Datause, Nobs, Nfeature, sigma0, mu0, K, Theta, W.star, Z.star, W\_temp, sigma.square, data.now, Z.now, i. Thus, users should avoid these variable names when using BayRepulsive\_unknown, if they don't want the variables to be overwritten. Especially i, which is commonly used in loops.

#### Value

W the estiamted signature matrix.

Z the estiamted proportion matrix.

C sum of estimated square error used as measure of performance for deconvolution.

### Source

BayRepulsive: A Bayesian Repulsive Deconvolution Model for Inferring Tumor Heterogeneity

#### **Examples**

```
rm(list=ls())
library(BayRepulsive)
data(CCLE)
set.seed(1)
Nobs
      <- dim(CCLE$DATA)[1]
Nfeature <- dim(CCLE$DATA)[2]
        <- matrix(rnorm(Nobs * Nfeature, mean = 0, sd = 0.1), nrow = Nobs)</pre>
DATA
        <- CCLE$DATA + error
DATA
         <- pmax(DATA,0)
result1 <- BayRepulsive_known(Datause = DATA, K = 3, Nobs = Nobs,
                              Nfeature = Nfeature)
cor(as.vector(result1$W), as.vector(CCLE$W))
#-----
rm(list=ls())
library(BayRepulsive)
data(Inhouse)
     <- dim(Inhouse$DATA)[1]
Nfeature <- dim(Inhouse$DATA)[2]
```

BayRepulsive\_unknown

BayRepulsive\_unknown is a deconvolution function designed for inferring tumor heterogeneity, used when the number of subclones is unknown.

## **Description**

Takes in the observed data matrix, the range of number of subclones, the number of features and samples, gives the estimation of the NMF, including the estimated number of subclones.

## Usage

#### **Arguments**

Datause	The observed data matrix.	Each row is a sample.

K\_min The minimum number of subclonesK\_max The Maximum number of subclones

Nobs The number of samples, i.e., the number of rows of the Datause

Nfeature The number of features, i.e., the number of columns of the Datause

Niter The number of maximum iterations

epsilon Break if the L2 distance of the two estiamted proportion matrix in row is less

than epsilon

tau The hyperparameter for DPP, a large number is prefered, default value is 100

seed The random seed, default as 1

# **Details**

Given an observed matrix, whose rows are mixed samples of unknown number of subclones, we give an estimation of number of subclones along with NMF results.

This function will create a bunch of globel variables, named Datause, Nobs, Nfeature, sigma0, mu0, K, Theta, W.star, Z.star, W\_temp, sigma.square, data.now, Z.now, i. Thus, users should avoid these variable names when using BayRepulsive\_unknown, if they don't want the variables to be overwritten. Especially i, which is commonly used in loops.

#### Value

W the estiamted signature matrix.

Z and the estiamted number of subclones.

K the estimated number of subclones.

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#### **Source**

BayRepulsive: A Bayesian Repulsive Deconvolution Model for Inferring Tumor Heterogeneity

### **Examples**

```
rm(list=ls())
library(BayRepulsive)
data(CCLE)
set.seed(1)
Nobs
         <- dim(CCLE$DATA)[1]
Nfeature <- dim(CCLE$DATA)[2]
         <- matrix(rnorm(Nobs * Nfeature, mean = 0, sd = 0.1), nrow = Nobs)</pre>
error
DATA
         <- CCLE$DATA + error
DATA
         <- pmax(DATA,0)
result1 <- BayRepulsive_unknown(Datause = DATA, K_min = 2, K_max = 6, Nobs = Nobs,
                                  Nfeature = Nfeature)
cor(as.vector(result1$W), as.vector(CCLE$W))
rm(list=ls())
library(BayRepulsive)
data(Inhouse)
Nobs
          <- dim(Inhouse$DATA)[1]
Nfeature <- dim(Inhouse$DATA)[2]</pre>
result1 <- BayRepulsive_unknown(Datause = Inhouse$DATA, K_min = 2, K_max = 6, Nobs = Nobs,
                                 Nfeature = Nfeature, seed = 12)
# handle the label swithing issue
          <- result1$W
W_{est[,1]} \leftarrow result1$W[,2]
W_est[,2] <- result1$W[,1]</pre>
cor(as.vector(W_est), as.vector(Inhouse$W))
```

CCLE CCLE

# Description

This dataset was generated from the pure cell line expression data from CCLE.

# Usage

data(CCLE)

#### **Format**

This is a data frame with three components: (a) the pure cell line expression, (b) the sample proportion, (c) the mixed data. Z: the pure cell line expression of three cancer cell lines – NCIH524\_LUNG, NCIH209\_LUNG, SBC5\_LUNG. Each line is for one cancer cell line. We selected top 100 differentially expressed gene. W: the sample proportion. Each row is the proportion of one sample. We used this sample porportion to mix 24 mixed samples. DATA: the mixed data. Each line is the gene expression for one sample. DATA = WZ.

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#### **Examples**

```
# import the data
data(CCLE)
# get the mixed data
CCLE$DATA
# get the gene expression level of pure cell line
CCLE$Z
# get the proportion
CCLE$W
```

Inhouse

Inhouse

## **Description**

This was generated the pure cell line expression data from one prostate cancer patient at Johns Hopkins hospital.

## Usage

data(Inhouse)

#### **Format**

This is a data frame with three components: (a) the pure cell line expression, (b) the sample proportion, (c) the mixed data. Z: the pure cell line expression of three cancer cell lines – naive CD4+ T cell, naive CD8+ T cell, and activated CD4+ T cell in tumor sample. Each line is for one cancer cell line. We selected top 100 differentially expressed gene. W: the sample proportion. Each row is the proportion of one sample. We used this sample porportion to mix 10 mixed samples. DATA: the mixed data with noise. Each line is the gene expression for one sample.

## **Examples**

```
# import the data
data(Inhouse)
# get the mixed data
Inhouse$DATA
# get the gene expression level of pure cell line
Inhouse$Z
# get the proportion
Inhouse$W
# get the added noise
Inhouse$DATA - Inhouse$W%*%Inhouse$Z
```

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