

# **Improved prediction of breast cancer outcome by identifying heterogeneous biomarkers**

## **User Manual (version 1.0)**

March 4, 2018

## Index

1. Installation .....	3
2. Quick start .....	3
1) Run .....	3
2) Example.....	4
3. Description of class CPR.....	5
1) class CPR .....	5
2) method CPR.fit() .....	6
3) method CPR.get_biomarkers() .....	6
4) method CPR.get_subnetwork() .....	6
4. Contact.....	7
5. Reference.....	7

# 1. Installation

This library 'CPR' requires:

Python 3,  
Numpy, ( $\geq 1.6.1$ )  
Scipy, ( $\geq 0.9$ )  
Scikit-learn ( $\geq 0.18$ )

To download or update those libraries, use 'pip' in command line.

*pip install numpy scipy scikit-learn*

Download the CPR module from <https://github.com/mathcom/CPR>.

If successfully downloaded, user can find the following 4 files:

*CPR.py*  
*ex\_EXPRESSION.txt*  
*ex\_NETWORK.txt*  
*ex\_CLINICAL.txt*

## 2. Quick start

### 1) Run

\$ **python CPR.py** [-h] [-m NUMCLUSTERS] [-d DAMPINGFACTOR]  
[-n NUMBIOMARKERS] [-c CONDITIONHUBGENE]  
**EXPRESSION\_FILE NETWORK\_FILE CLINICAL\_FILE**  
**RESULT\_FILE**

Option	Name	Description																
	<i>EXPRESSION_FILE</i> (positional argument)	Tab-delimited file for gene expression profiles as below: <table><tr><td>PATIENT</td><td>TCGA-AR-A24H</td><td>TCGA-AR-A24L</td><td>TCGA-AR-A24M</td></tr><tr><td>A1CF</td><td>-0.436158</td><td>-0.276784</td><td>-0.309453</td></tr><tr><td>A2M</td><td>1.90128</td><td>2.72735</td><td>4.03939</td></tr><tr><td>A4GALT</td><td>-0.408337</td><td>-0.247608</td><td>-0.260444</td></tr></table>	PATIENT	TCGA-AR-A24H	TCGA-AR-A24L	TCGA-AR-A24M	A1CF	-0.436158	-0.276784	-0.309453	A2M	1.90128	2.72735	4.03939	A4GALT	-0.408337	-0.247608	-0.260444
PATIENT	TCGA-AR-A24H	TCGA-AR-A24L	TCGA-AR-A24M															
A1CF	-0.436158	-0.276784	-0.309453															
A2M	1.90128	2.72735	4.03939															
A4GALT	-0.408337	-0.247608	-0.260444															
	<i>NETWORK_FILE</i> (positional argument)	Tab-delimited file for gene interaction network as below: <table><tr><td>GENE1</td><td>GENE2</td></tr><tr><td>RPL37A</td><td>RPS27A</td></tr><tr><td>MRPL1</td><td>MRPS36</td></tr><tr><td>RFC3</td><td>SPRTN</td></tr></table>	GENE1	GENE2	RPL37A	RPS27A	MRPL1	MRPS36	RFC3	SPRTN								
GENE1	GENE2																	
RPL37A	RPS27A																	
MRPL1	MRPS36																	
RFC3	SPRTN																	
	<i>CLINICAL_FILE</i> (positional argument)	Tab-delimited file for patient’s clinical data as below. LABEL=1:poor prognosis and 0:good prognosis: <table><tr><td>PATIENT</td><td>LABEL</td></tr><tr><td>TCGA-AR-A24H</td><td>0</td></tr><tr><td>TCGA-AR-A24L</td><td>0</td></tr><tr><td>TCGA-AR-A2LH</td><td>1</td></tr></table>	PATIENT	LABEL	TCGA-AR-A24H	0	TCGA-AR-A24L	0	TCGA-AR-A2LH	1								
PATIENT	LABEL																	
TCGA-AR-A24H	0																	
TCGA-AR-A24L	0																	
TCGA-AR-A2LH	1																	
	<i>RESULT_FILE</i> (positional argument)	The results of CPR are saved with the following names: <div>1) <i>RESULT_FILE_biomarker.txt</i> 2) <i>RESULT_FILE_subnetwork.txt</i> 3) <i>RESULT_FILE_accuracy.txt</i></div>																

Option	Name	Description
<b>-h</b>	<i>Help message (optional argument)</i>	Show this help message and exit
<b>-m</b>	<i>Number of sample clusters (optional argument)</i>	A parameter of <i>K-means clustering</i> algorithm. This parameter decides number of sample clusters to handle the heterogeneity of patients. The default value makes the number of clusters be determined by using the <i>silhouette score</i> . If a specific number is given, the K-means is carried out with the number. (default=0)
<b>-d</b>	<i>Damping factor (optional argument)</i>	A parameter of <i>PageRank</i> algorithm. This parameter decides an influence of network information on prediction. The value must be between 0 and 1. (default=0.7)
<b>-n</b>	<i>Number of markers (optional argument)</i>	This parameter decides number of biomarkers to use in prediction. (default=70)
<b>-c</b>	<i>Condition of hub-gene (optional argument)</i>	This parameter is used to identify a hub-gene. When $c$ is a given parameter and $x$ is the total of genes, we define top $cx$ genes with high degree as hub-genes. The value must be between 0 and 1. (default=0.02)

## 2) Example

\$ **python CPR.py** ex\_EXPRESSION.txt ex\_NETWORK.txt ex\_CLINICAL.txt ex\_RESULT.txt

### (1) Log in command line

```
>>> 0. Arguments
Namespace(CLINICAL_FILE='ex_CLINICAL.txt',      EXPRESSION_FILE='ex_EXPRESSION.txt',
NETWORK_FILE='ex_NETWORK.txt',      RESULT_FILE='ex_RESULT',      conditionHubgene=0.02,
dampingFactor=0.7, numBiomarkers=70, numClusters=0)
>>> 1. Load data
>>> 2. Preprocess data
   n_samples: 189
   n_genes  : 8819
   n_edges  : 150168
>>> 3. Conduct CPR
   K-means clustering
   -> n_clusters: 3
       In cluster[0], n_samples:76, n_goods:44, n_poor:32
       In cluster[1], n_samples:107, n_goods:52, n_poor:55
       In cluster[2], n_samples:6, n_goods:3, n_poor:3
   Modified PageRank
>>> 4. 10-fold Cross validation
   10% complete!
   20% complete!
   30% complete!
   40% complete!
   50% complete!
   60% complete!
   70% complete!
   80% complete!
   90% complete!
   100% complete!
   AUC-ROC : 0.710
   Accuracy: 0.640
```

(2) ex\_RESULT\_biomarker.txt

- PRscore is the mean of scores computed in each sample cluster.

GeneSymbol	PRscore
PIK3R2	1.684137
HSPA8	1.645991
CALM1	1.555840
POLR2L	1.516215
HDAC2	1.481628
H3F3A	1.480993
GSK3B	1.422521
EGFR	1.395188
POLR2C	1.383287

(3) ex\_RESULT\_subnetwork.txt

- A resulting subnetwork is a subnetwork of the given whole network.
- Each edge has at least one biomarker gene.

source	target
MAPK13	NFATC3
CREBBP	EP400
GMNN	PSMC2
RNPS1	RPSA
EGFR	TLR2

(4) ex\_RESULT\_accuracy.txt

AUC-ROC : 0.710
Accuracy: 0.640

## 3. Description of class CPR

A user can import and utilize CPR.py in python. CPR.py provides one class and its three functions.

### 1) class CPR

Clustering and PageRank-based gene selection method. For more detail, please refer to “Improved prediction for breast cancer outcome by identifying heterogeneous biomarkers”.

Parameters	<p><b>dampingFactor:</b> float, optional (default=0.7) A parameter of <i>PageRank</i> algorithm. This parameter decides an influence of network information on prediction Range = 0.0 ~ 1.0</p> <p><b>n_biomarkers:</b> int, optional (default=70) User can control the number of prognostic biomarkers.</p> <p><b>n_clusters:</b> int, optional (default=0) A parameter of <i>K-means clustering</i> algorithm.</p>
------------	--

	<p>This parameter decides number of sample clusters to handle the heterogeneity of patients. The default value makes the number of clusters be determined by using the <i>silhouette score</i>. If a number is given, the K-means is carried out with the number.</p> <p><b>c_hubgene:</b> float, optional (default=0.02)  This parameter is used to identify a hub-gene. When <i>c</i> is a given parameter and <i>x</i> is the total of genes, we define top <i>cx</i> genes with high degree as hub-genes.</p> <p><b>logshow:</b> bool, optional (default=False)  If <i>logshow</i> is <i>False</i>, any log in class CPR is not shown in command line.</p>
--	--

## 2) [method](#) CPR.fit()

Build a CPR gene selection model.

<b>Parameters</b>	<p><b>expr:</b> numpy.array(shape=[n_samples,n_genes], dtype=numpy.float32)  A gene expression dataset without any header.  The order of sample must be equal to one of <i>labels</i>.  The order of gene must be equal to one of <i>genes</i>.</p> <p><b>labels:</b> numpy.array(shape=[n_samples], dtype=numpy.int32)  A value of vector represents a label of sample.  (0: <i>good</i> prognosis and 1: <i>poor</i> prognosis)  The order of label must be equal to one of sample in <i>expr</i>.</p> <p><b>genes:</b> list  A list of genes in expression data.  The order of gene must be equal to one of gene in <i>expr</i>.</p> <p><b>edges:</b> list  A list of edges, and all edges have <i>tuple</i> type.</p> <p><b>random_state:</b> int or None, optional (default=1)  This parameter is used in scikit-learn functions.  If <i>int</i>, the results are always same. If <i>None</i>, a result can be different each time.</p>
-------------------	--

## 3) [method](#) CPR.get\_biomarkers()

A list of biomarkers identified by built model is provided as list type.

<b>Returns</b>	<p><b>biomarkers:</b> list  An element of the result has <i>tuple</i> type and two values.  (0:gene symbol and 1:PRscore)  The list is sorted by the descending order of PRscore.</p>
----------------	---

## 4) [method](#) CPR.get\_subnetwork()

A list of edges containing at least one biomarker is provided as list type.

<b>Returns</b>	<p><b>edges:</b> list  An element of the result has <i>tuple</i> type.  Each tuple consists of two gene symbols.</p>
----------------	--

## 4. Contact

Bug reporting, questions or any suggestions are highly appreciated.

Jonghwan Choi ([mathcom@inu.ac.kr](mailto:mathcom@inu.ac.kr))

Jaegyeon Ahn ([jgahn@inu.ac.kr](mailto:jgahn@inu.ac.kr))

## 5. Reference

Choi, Jonghwan, et al. "Improved prediction of breast cancer outcome by identifying heterogeneous biomarkers." *Bioinformatics* **33.22** (2017): 3619-3626.