

To run the CODE&ACAS algorithms, please refer to the ACAS_release directory.

Input argument is a cell and its arguments are described as follow:

dim — integer, the embedded dimensionality

```
model_optimisation — Cell, options for optimisation  
Default: {'optimisation', 'grid', 'obj_option',  
'quant_knn', 'obj_para', {'knn', 10,10}};
```

obj_option — String, choosing options for computing the optimisation score. Options:

quant_knn score,these two parameters are used to compute the $K(R)$ and $K(Q)$

parameter_range — Cell, parameter searching range for the proposed algorithm. Including the eta_1, eta_2,

alpha (correspond to ξ in the paper), beta (correspond to γ in the paper). The

following parameter range is used for all data:

```

{'eta1', [-1, 10], 'eta2', [-1,
10], 'alpha', [0, 3], 'beta', [0, 3]}

```

X — $m \times k$ matrix, embeddings for the row objects in R

Y — $n \times k$ matrix, embeddings for the column objects in R

model -- Structure, contains the 4 parameter values (eta1, eta2, alpha, beta) for the computed co-embedding

Synthetical Data

At first load the associate synthetical data to the workspace from the Data folder, for example:

```
*****code*****
rng(213);
% load the associate data file
load('Data/dot.mat');
*****code*****
```

The **Plotcluster.m** function is used to plot data in 2&3 dimensions. To use this function, run the following code to see the plot:

```
*****code*****
figure
Plotcluster(X', lx, Y', ly);
*****code*****
```

Thus, we can be able to visualising the X and Y data points with their coloured labeling information corresponding to label vector l_x and l_y , respectively. The synthetical relational matrix **R** is computed by the **CalRelationXY.m** function:

```
*****code*****
R      = CalRelationXY(X,Y,'gaussian','mean');
*****code*****
```

At last , we obtain the row and column co-embeddings of LEDF as well as plot it by running script of the following:

```

*****code*****
opt = {'optimisation', 'ga', 'obj_option', 'quant_knn', 'obj_para',
{'knn', 10,10}};

para_range = {'eta1',[-1, 10],'eta2', [-1, 10],'alpha',[0, 3], 'beta',
[0,3]};

I = {'R', R, 'method', 'LEDeig','dim',2, 'model_optimisation', opt,
'parameter_range',para_range };

led = LEDF(I);
led = training(led);

figure;
Plotcluster(led.X',lx,led.Y',ly);

*****code*****

```

Small Clinical Trial Text Data

This is a smaller database we extracted from the Clinical Trial data which only include 4 topics domain, and the resulting co-embedding we used in the paper are only 2 dimensionality, the knn parameters k_r , k_c we used to produce the results of the paper are 5&5 respectively:

```
*****code*****
rng(213);
```

```

% load the associate data file
load('Data/Small_CliTri.mat');

opt = {'optimisation', 'ga', 'obj_option', 'quant_knn', 'obj_para',
{'knn', 5,5}};

para_range = {'eta1',[-1, 10],'eta2', [-1, 10],'alpha',[0, 3], 'beta',
[0,3]};

I = {'R', R, 'method', 'LEDeig','dim',2, 'model_optimisation', opt,
'parameter_range',para_range };

led = LEDF(I);
led = training(led);

figure;
Plotcluster(led.X',lx,led.Y',ly);

*****code*****

```

Clinical Trial Text Data

We searched the results from 2 to 20 dimensions and the knn score parameters setting are 10&15.

```

*****code*****

rng(213);
% load the associate data file
load('Data/ClinicalTrial.mat');

led = cell(20,1);

for dim = 2 : 20
    opt = {'optimisation', 'ga', 'obj_option', 'quant_knn', 'obj_para',
{'knn', 10,15}};

    para_range = {'eta1',[-1, 10],'eta2', [-1, 10],'alpha',[0, 3], 'beta',
[0,3]};

    I = {'R', R, 'method', 'LEDeig','dim',dim, 'model_optimisation',
opt, 'parameter_range',para_range };

    led{dim-1} = LEDF(I);
    led{dim-1} = training(led{dim-1});
end

*****code*****

```

Reuters Text Data

We searched the results from 2 to 20 dimensions and the knn score parameters setting are 5&5. The co-embedding for the proposed method was produced through the following code:

```

*****code*****

rng(213);
% load the associate data file

```

```

load('Data/Reuters.mat');

led = cell(20,1);

for dim = 2 : 20
    opt = {'optimisation', 'ga', 'obj_option', 'quant_knn', 'obj_para',
{'knn', 5,5}};

    para_range = {'eta1',[-1, 10],'eta2', [-1, 10],'alpha',[0, 3], 'beta',
[0,3]};

    I = {'R', R, 'method', 'LEDeig','dim',dim, 'model_optimisation',
opt, 'parameter_range',para_range };

    led{dim-1} = LEDF(I);
    led{dim-1} = training(led{dim-1});
end

*****code*****

```

Cora Citation Data

The code for obtaining the Cora data embedding in 2 to 10 dimensional space is as follow:

```

*****code*****

rng(213);

% load the associate data file
load('Data/processed_cora.mat');

led = cell(10,1);

for dim = 2 : 10
    opt = {'optimisation', 'ga', 'obj_option', 'quant_knn', 'obj_para',
{'knn', 5,5}};

    para_range = {'eta1',[-1, 10],'eta2', [-1, 10],'alpha',[0, 3], 'beta',
[0,3]};

    I = {'R', R, 'method', 'LEDeig','dim',dim, 'model_optimisation',
opt, 'parameter_range',para_range };

    led{dim-1} = LEDF(I);
    led{dim-1} = training(led{dim-1});
end

*****code*****

```

Citeseer Citation Data

The code for obtaining the Citesser data embedding in 2 to 10 dimensional space is as follow:

```

*****code*****

rng(213);
% load the associate data file

```

```

load('Data/processed_citeseer.mat');

led = cell(10,1);

for dim = 2 : 10
    opt = {'optimisation', 'ga', 'obj_option', 'quant_knn', 'obj_para',
{'knn', 5,5}};

    para_range = {'etal',[-1, 10],'eta2', [-1, 10],'alpha',[0, 3], 'beta',
[0,3]};

    I = {'R', R, 'method', 'LEDeig','dim',dim, 'model_optimisation',
opt, 'parameter_range',para_range };

    led{dim-1} = LEDF(I);
    led{dim-1} = training(led{dim-1});
end
*****code*****

```

Industry Company Sector Data

This dataset results are also produce under settings the knn score parameters as 5&5:

```

*****code*****

rng(213);

% load the associate data file
load('Data/processed_industry.mat');

led = cell(10,1);

for dim = 2 : 10
    opt = {'optimisation', 'ga', 'obj_option', 'quant_knn', 'obj_para',
{'knn', 5,5}};

    para_range = {'etal',[-1, 10],'eta2', [-1, 10],'alpha',[0, 3], 'beta',
[0,3]};

    I = {'R', R, 'method', 'LEDeig','dim',dim, 'model_optimisation',
opt, 'parameter_range',para_range };

    led{dim-1} = LEDF(I);
    led{dim-1} = training(led{dim-1});
end

save('industry_LEDF.mat');

*****code*****

```