Article

# Network matrix based methods for between-network comparison

## WenJun Zhang

School of Life Sciences, Sun Yat-sen University, Guangzhou 510275, China; International Academy of Ecology and Environmental Sciences, Hong Kong E-mail: zhwj@mail.sysu.edu.cn, wjzhang@iaees.org

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# Abstract

In present article, I introduced some network matrix based methods for comparing and testing between-network difference/similarity, including the methods for interval weights based network matrix, including between-network similarity, randomization test of between-network difference, and statistic test of between-network difference, and the method for Boolean weights based network matrix. In addition, degree change index, weight change index, and eigenvector matrix change index were presented also. Matlab codes of the methods were provided.

Keywords network matrix; network comparison; difference; similarity; algorithms; Matlab.

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## **1** Introduction

The structure of network covers node degree, network connectance, aggregation strength, etc (Dormann,2011; Zhang and Zhan, 2011). Various statistics may be used in the difference comparison of network structure (Solow, 1993; Manly, 1997; Zhang, 2012a, 2018). For example, a Java algorithm has been presented to statistically compare between-network structure difference (Zhang, 2011a). In addition, the Java algorithm was also developed to statistically compare between-community structure difference (Zhang, 2011b). In present article, I introduce some network matrix based methods to compare and test between-network difference/similarity.

## 2 Methods

Suppose we compare the two networks A and B, based on their network matrices  $A=(a_{ij})$  and  $B=(b_{ij})$ , i, j=1, 2, ..., m, where m is the number of nodes,  $a_{ij}$  and  $b_{ij}$  are the weights between two nodes i and j in two networks respectively.

## 2.1 Interval weights based network matrix

Assume the weights  $a_{ij}$  and  $b_{ij}$  are all interval values. Several methods below can be used to compare two networks. In these methods, network matrices *A* and *B* are firstly transformed to vectors  $A_1$  and  $B_1$  respectively,

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i.e.,  $(A_1, B_1) = \{(a_{ij}, b_{ij}) \mid i, j=1, 2, ..., m\}$ , and the corresponding methods are thus used.

2.1.1 Between-network similarity

By using Pearson correlation, etc (Zhang, 2012a, 2012c, 2015, 2018; Zhang and Li, 2015a, 2015b), we can achieve the test results for network similarity/correlation. Generalized Matlab codes for Pearson correlation between two network matrices, *A* and *B*, are as follows (Zhang, 2012a, 2018)

AA=input('Input the excel file name of network matrix A: ','s'); A=xlsread(AA); BB=input('Input the excel file name of network matrix B: ','s'); B=xlsread(BB); m=size(A,1); mm=size(A,2); n=size(B,1); nn=size(B,2); if ~((m==mm) & (n==nn) & (m==n)) error('Network matrices A and B should be the square matrices of the same size.'); end % Matrices A and B are the square matrices of the same size. sig=input('Input significance level(e.g., 0.01): '); A1=reshape(A, m\*m, 1); %Transform matrix A to vector A1 B1=reshape(B, m\*m, 1); % Transform matrix B to vector B1 r=corr(A1,B1); fprintf(['Pearson correlation r=' num2str(r) '\n']); tvalue= $abs(r)/sqrt((1-r^2)/(m*m-2));$ p=(1-tcdf(tvalue,m\*m-2))\*2;sigma=p<sig;

 $if (sigma == 1) fprintf(['Pearson correlation is statistically significant (p=' num2str(p) ')\n']); end$ 

 $if (sigma == 0) \ fprintf(['Pearson \ correlation \ is \ not \ statistically \ significant \ (p=' \ num 2 str(p) \ ')\n']); \ end \ (p=' \ num 2 str(p) \ ')\n') \ (p=' \ num 2 str(p) \ ')\n'); \ end \ (p=' \ num 2 str(p) \ ')\n') \ (p=' \ num 2 str(p) \ ')\n'); \ end \ (p=' \ num 2 str(p) \ ')\n') \ (p=' \ num 2 str(p) \ ')\ (p=' \ num 2 str(p) \ ')\$ 

#### 2.1.2 Randomization test of between-network difference

By using randomization test for between-network difference based on Euclidean distance, Manhattan distance, Chebyshov distance, and Pearson correlation-based distance (Zhang, 2012a, 2015, 2018; Manly, 1997; Schoenly and Zhang, 1999; Solow, 1993; Zhang and Schoenly, 1999), we can achieve the test results for network difference/similarity. Generalized Matlab codes for randomization test between two network matrices, *A* and *B*, are as follows (Zhang, 2012a, 2018)

AA=input('Input the excel file name of network matrix A: ','s');

A=xlsread(AA);

BB=input('Input the excel file name of network matrix B: ','s');

B=xlsread(BB);

m=size(A,1); mm=size(A,2); n=size(B,1); nn=size(B,2);

if ~((m==mm) & (n==nn) & (m==n))

error('Network matrices A and B should be the square matrices of the same size.');

end

% Matrices A and B are the square matrices of the same size.

sim=input('Input the maximum number of simulations (e.g., 100): ');

sel=input('Choose distance measure (1: Euclidean distance; 2: Manhattan distance; 3: Chebyshov distance; 4: Pearson correlation): ');

sig=input('Input the significance level (e.g., 0.05): ');

A1=reshape(A, m\*m, 1); %Transform matrix A to vector A1

B1=reshape(B, m\*m, 1); % Transform matrix B to vector B1

pvalue=randTest(A1,B1,sim,sel);

```
if (pvalue<sig) fprintf(['Difference is statistically significant (p=' num2str(pvalue) ')\n']); end
```

if (pvalue>=sig) fprintf(['Difference is not statistically significant (p=' num2str(pvalue) ')\n']); end

The functions, randTest.m, euclideandis.m, manhattandis.m, chebyshovdis.m, and correcoeffdiff.m, used in the algorithm above are as follows (Zhang, 2018)

```
functionpvalue=randTest(x,y,sim,sel)
                                              %pvalue: calculated p value.
if ((\min(size(x)) \sim = 1) | (\min(size(y)) \sim = 1))
                                              % sim: times of randomizations.
error('Both x and y are vectors');
                                             %x and y: two vectors to be tested. x and y are row vectors.
end
m=max(size(x));
if (max(size(y))~=m)
error('Vector sizes do not match.');
end
if (sim < =1)
error('No. randomizations are too less.');
end
dum=min(min(x),min(y));
if (dum<0)
x=x-dum;
y=y-dum;
end
ma=-1e10;
for j=1:2
for i=1:m
in=1;
if (j==1) dum=x(i);
else dum=y(i);
end
while (m \sim = 0)
if ((abs(dum-floor(dum))<1) & (~(abs(dum-floor(dum))<=1e-10)))
in=in*10;
dum=dum*10;
if ((floor(dum+1e-10))~=(floor(dum))) break; end
else break; end
end
if (in>ma) ma=in; end
end; end
x=x.*ma.*1.0;
y=y.*ma.*1.0;
switch sel
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case 1
dxy=euclideandis(x,y);
case 2
dxy=manhattandis(x,y);
case 3
dxy=chebyshovdis(x,y);
case 4
dxy=correcoeffdiff(x,y);
end
nrx=sum(x);
nrxy=sum(sum(x+y));
fr=0;
for sm=1:sim
ar=floor(x+y);
col=sum(ar);
br(1)=ar(1);
for i=2:m
br(i)=br(i-1)+ar(i);
end
cols=randperm(nrxy);
p1(1:m)=0;
for j=1:m
if (ar(j)==0) continue; end
if (j==1) temp=0;
else temp=br(j-1);
end
for i=1:nrx
if ((cols(i) > temp) & (cols(i) < =br(j))) p1(j)=p1(j)+1; end
end; end
p2=ar'-p1;
switch sel
case 1
dum=euclideandis(p1,p2);
case 2
dum=manhattandis(p1,p2);
case 3
dum=chebyshovdis(p1,p2);
case 4
dum=correcoeffdiff(p1,p2);
end
if (abs(dum)>=abs(dxy))
fr=fr+1;
end
end
pvalue=fr/sim;
```

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function distance = euclideandis(x,y)
                                           %x and y: two vectors to be tested.
if (max(size(x))~=max(size(y)))
error('Array sizes do not match.');
end
if ((\min(size(x)) \sim = 1) | (\min(size(y)) \sim = 1))
error('Both x and y are vectors');
end
distance =sqrt(sum((x-y).^2))/max(size(x));
function distance = manhattandis(x,y)
                                            %x and y: two vectors to be tested.
if (max(size(x))~=max(size(y)))
error('Array sizes do not match.');
end
if ((\min(size(x)) \sim = 1) | (\min(size(y)) \sim = 1))
error('Both x and y are vectors');
end
distance =sum(abs(x-y))/max(size(x));
function distance = chebyshovdis(x,y)
                                            %x and y: two vectors to be tested.
if (max(size(x))~=max(size(y)))
error('Array sizes do not match.');
end
if ((\min(size(x)) \sim = 1) | (\min(size(y)) \sim = 1))
error('Both x and y are vectors');
end
distance =max(abs(x-y));
function diff = correcoeffdiff(x,y)
                                         %x and y: two vectors to be tested.
m=max(size(x));
if (m~=max(size(y)))
error('Array sizes do not match.');
end
if ((\min(size(x)) \sim = 1) | (\min(size(y)) \sim = 1))
error('Both x and y are vectors');
end
xbar=mean(x);
ybar=mean(y);
aa = sum(x.*y) - ybar*sum(x) - xbar*sum(y) + m*xbar*ybar;
bb=sum(x.^2)-2*xbar*sum(x)+m*xbar^2;
cc=sum(y.^2)-2*ybar*sum(y)+m*ybar^2;
diff =1-aa/sqrt(bb*cc);
```

#### 2.1.3 Statistic test of between-network difference

Known the weight pairs  $(a_{ij}, b_{ij})$ , i, j=1, 2, ..., m, i.e.,  $(A_1, B_1)$ . Assume the difference  $A_1$ - $B_1$  between two vectors  $A_1$  and  $B_1$  follows the normal distribution. If there is not statistic difference between network matrices A and B, the mean of  $A_1$ - $B_1$  should be zero. The *t*-test can be used to test the statistic difference (Zhang, 2018). The Matlab codes are as follows

AA=input('Input the excel file name of network matrix A: ','s'); A=xlsread(AA); BB=input('Input the excel file name of network matrix B: ','s'); B=xlsread(BB); m=size(A,1); mm=size(A,2); n=size(B,1); nn=size(B,2); if ~((m==mm) & (n==nn) & (m==n)) error('Network matrices A and B should be the square matrices of the same size.'); end % Matrices A and B are the square matrices of the same size. alpha=input('Input significance level(e.g., 0.01)'); A1=reshape(A, m\*m, 1); % Transform matrix A to vector A1 B1=reshape(B, m\*m, 1); % Transform matrix B to vector B1 [h,p,ci,stats]=ttest(A1-B1,0,alpha,0); if (h==1) fprintf(['Difference is statistically significant (p=' num2str(p) ')\n']); end

## 2.2 Boolean weights based network matrix

If network matrices *A* and *B* are Boolean type, i.e., the weights  $a_{ij}$  and  $b_{ij}$  take 0 or 1, point correlation can be used to calculate between-network similarity/correlation (Zhang, 2015, 2017, 2018). In this method, network matrices *A* and *B* are firstly transformed to vectors  $A_1$  and  $B_1$  respectively, i.e.,  $(A_1, B_1) = \{(a_{ij}, b_{ij}) | i, j=1, 2, ..., m\}$ , and point correlation is thus calculated and tested.

Generalized Matlab codes for Point correlation between two network matrices, *A* and *B*, are as follows (Zhang, 2015, 2017, 2018)

AA=input('Input the excel file name of network matrix A: ','s'); A=xlsread(AA); BB=input('Input the excel file name of network matrix B: ','s'); B=xlsread(BB); m=size(A,1); mm=size(A,2); n=size(B,1); nn=size(B,2); if ~((m==mm) & (n==nn) & (m==n)) error('Network matrices A and B should be the square matrices of the same size.'); end % Matrices A and B are the square matrices of the same size. sig=input('Input the significance level (e.g., 0.05): '); x=reshape(A, m\*m, 1); % Transform matrix A to vector x y=reshape(B, m\*m, 1); % Transform matrix B to vector y n=size(x,1); aa=sum((x==0) & (y==0)); bb=sum((x==0) & (y==0)); cc=sum((x~=0) & (y==0));

dd=sum((x~=0) & (y~=0));

pointcorr=(aa\*dd-bb\*cc)/sqrt((aa+bb)\*(cc+dd)\*(aa+cc)\*(bb+dd));

 $chi2 = n*(aa*dd-bb*cc)^2/((aa+bb)*(cc+dd)*(aa+cc)*(bb+dd));$ 

chi2test=chi2>chi2inv(1-sig,1);

%chi2=10.8 for sig=0.001; chi2=6.635 for sig=0.01; chi2=7.87 for sig=0.005

fprintf(['Point correlation=' num2str(pointcorr) '\n']);

 $if (chi2test == 1) fprintf(['Point correlation is statistically significant (p=' num2str(sig) ')\n']); end$ 

if (chi2test==0) fprintf(['Point correlation is not statistically significant (p=' num2str(sig) ')\n']); end

## 2.3 Other generalized methods

2.3.1 Degree change index

Degree change index, proposed by Zhang (2012b), can be popularized to compare two networks

$$D = \sum_{j=1}^{m} (|O_{Aj} - O_{Bj}| + |I_{Aj} - I_{Bj}|)$$

where *D*: value of degree change index; *m*: total number of nodes in the network;  $O_{Aj}$ ,  $O_{Bj}$ : out-degree of node for networks *A* and *B* respectively;  $I_{Aj}$ ,  $I_{Bj}$ : in-degree of node *j* for networks *A* and *B* respectively. 2.3.2 Weight change index

Weight change index (*S*), popularized from parameter robustness (Zhang, 2016), can be used to compare two networks

$$S = \sum_{i} \sum_{j} |a_{ij} - b_{ij}|$$

2.3.3 Eigenvector matrix change index

Suppose the eigenvector matrices of the network matrices A and B are  $V_a$  and  $V_b$  respectively. The eigenvector matrix change index is defined as

$$V = \sum_{i} \sum_{j} |V_{aij} - V_{bij}|$$

where  $V_{aij}$  and  $V_{bij}$  are elements of the eigenvector matrices  $V_a$  and  $V_b$  respectively, i, j=1, 2, ..., m.

AA=input('Input the excel file name of network matrix A: ','s');

A=xlsread(AA);

BB=input('Input the excel file name of network matrix B: ','s');

B=xlsread(BB);

m=size(A,1); mm=size(A,2); n=size(B,1); nn=size(B,2);

if ~((m==mm) & (n==nn) & (m==n))

error('Network matrices A and B should be the square matrices of the same size.');

end

% Matrices A and B are the square matrices of the same size.

S=sum(sum(abs(A-B)));

[VA,D] = eig(A,'nobalance');

[VB,D] = eig(B,'nobalance');

V=sum(sum(abs(VA-VB))); fprintf(['Weight change index S=' num2str(S) '\n']); fprintf(['Eigenvector matrix change index V=' num2str(V) '\n']);

## **3 Discussion**

In practical uses, several methods above can be jointly used to analyze various differences or similarities between two networks. It should be noted that all methods above can be improved, in particular the degree change index, weight change index, and eigenvector matrix change index. For instance, these absolute indices can be standardized as some forms of proportions.

Finally, I assume that the two network matrices are known, which should be defined and specified by researchers.

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