

# Algorithms for network positions

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## Outline

### Measuring structural equivalence

- Distance-based similarity measure
- Correlation measure

### Algorithms for approximate structural equivalence

- CONCOR
- STRUCTURE
- Tabu Search, and other combinatorial optimisation approaches

### Some practical issues

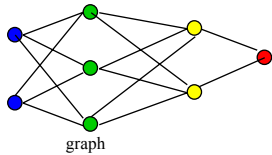
- Including the transpose?
- How many blocks?
- Which algorithm?

### Algorithms for approximate regular equivalence

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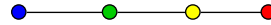
# Structural equivalence

Nodes  $a$  and  $b$  are *structurally equivalent* if, whenever  $(a,x)$  is an edge in  $G$  then so is  $(b,x)$ , and conversely



```
00111000
00111000
11000110
11000110
11000110
00111001
00111001
00000110
```

adjacency matrix for graph



```
0100
1010
0101
0010
```

adjacency matrix for blockmodel

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# Measuring structural equivalence

Two nodes are *structurally equivalent* if and only if the corresponding rows (or columns) of the adjacency matrix are identical

We can therefore measure the *degree* to which two nodes are structurally equivalent by measuring the degree to which their columns are identical:

## A distance measure

e.g., square root of sum of squared discrepancies between corresponding column entries (Euclidean distance): Burt's **STRUCTURE** program

## A proximity measure

e.g., correlation coefficient between columns (Pearson or phi coefficient): Breiger, Boorman & Arabie's (1975) **CONCOR** algorithm

## Directed graphs:

the **transpose** of the adjacency matrix is usually stacked below the matrix  
*to measure similarity across both rows and columns of the original matrix*

## Multiple graphs:

the matrices are usually stacked one below another  
*to measure similarity across all graphs*

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## Early blockmodelling algorithms

Clustering algorithms can then be used to determine blocks of approximately structurally equivalent nodes

Algorithms may be:

*Hierarchical* or *non-hierarchical*

*Divisive* or *agglomerative*

Early algorithms:

**STRUCTURE** (hierarchical, agglomerative)

**CONCOR** (hierarchical, divisive)

## More recent algorithms

Define a criterion of goodness of fit of a partition to an equivalence form

Use a combinatorial optimisation algorithm to find a partition with optimal fit

Examples:

**UCINET: Tabu search**

**Pajek: local optimisation**

start with some partition, and replace the current partition with any partition in its “neighbourhood” that has a more optimal fit; repeat until no further improvement is possible

## Doreian data in DL format

```
DL n=14 format=fullmatrix
labels
"Executive"
"Auditor"
"Sheriff"
"Council 1"
"Council 2"
"Council 3"
"Council 4"
"President"
"Council 5"
"Council 6"
"Fr Council"
"Fr Pres"
"City Mayor"
"Prosecutor"
data:
0 0 1 1 0 0 1 0 0 0 0 0 0 1 0
0 0 0 0 0 0 0 1 1 1 0 1 0 0
1 0 0 1 0 1 1 0 0 0 0 0 0 1
1 0 1 0 1 1 1 0 0 0 0 1 0 0
0 0 0 1 0 1 0 0 0 0 0 0 0 0
0 0 1 1 1 0 0 0 0 0 0 1 0 1
1 0 1 1 0 0 0 0 0 0 0 0 0 0
0 1 0 0 0 0 0 0 1 1 0 1 1 0
0 1 0 0 0 0 0 1 0 1 0 0 1 0
0 1 0 0 0 0 0 1 1 0 0 0 1 0
0 0 0 0 0 0 0 0 0 0 0 0 1 0 0
0 1 0 1 0 1 0 1 0 0 1 0 1 0
1 0 0 0 0 0 0 1 1 1 0 1 0 0
0 0 1 0 0 1 0 0 0 0 0 0 0 0 0
```

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## UCINET 6: CONCOR

In UCINET 6, read in data in DL format and then select:

Network

Roles & Positions

Structural

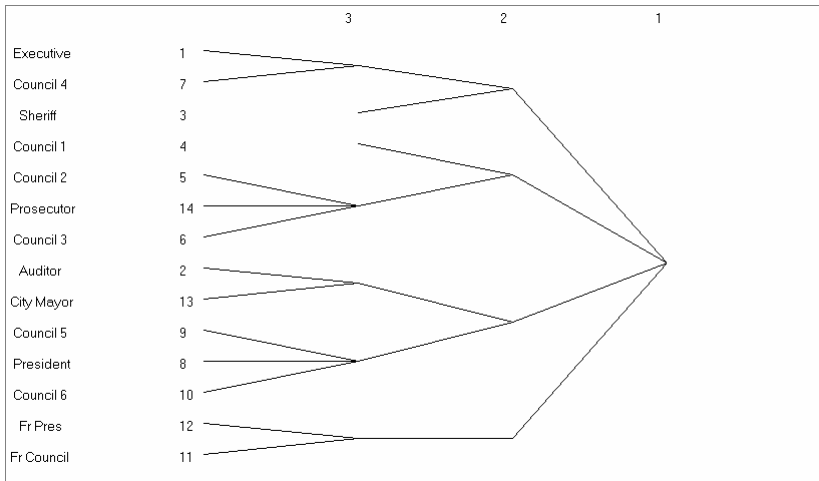
CONCOR

[choose max depth of splits = 3,  
defaults, otherwise]

*Use "ConcorCCPart" as attribute for colouring nodes*

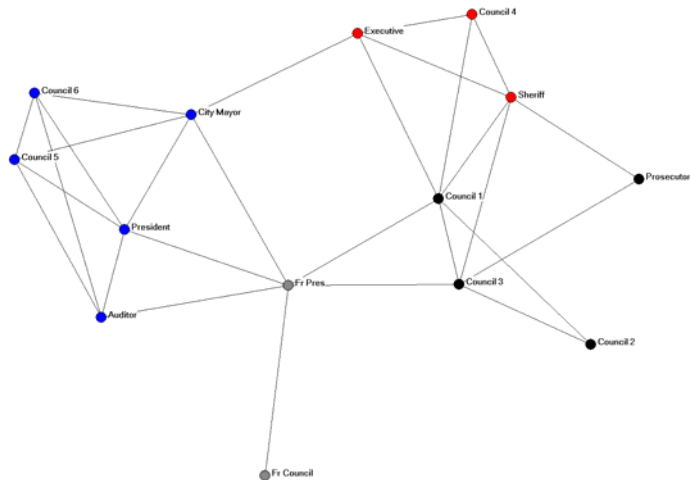
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# CONCOR partition



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# 4-block CONCOR solution



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# Selected output

## Blocked Matrix

	1	1	1	1 1
	1 3 7	5 4 6 4	2 9 0 8 3	2 1
	E S C	C P C C	A C C P C	F F
1 Executive	1 1	1	1	
3 Sheriff	1 1	1 1 1		
7 Council 4	1 1	1		
5 Council 2		1 1		
14 Prosecutor	1	1		
6 Council 3	1	1 1 1	1	1
4 Council 1	1 1 1	1 1		1
2 Auditor			1 1 1	1
9 Council 5			1 1 1 1	
10 Council 6			1 1 1 1	
8 President			1 1 1 1 1	1
13 City Mayor	1		1 1 1	1
12 Fr Pres		1 1	1 1 1	1
11 Fr Council				1

## Density Matrix

	1	2	3	4
1	1.000	0.417	0.067	0.000
2	0.417	0.667	0.000	0.250
3	0.067	0.000	0.900	0.300
4	0.000	0.250	0.300	1.000

R-squared = 0.500

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# A 4-block solution using “optimisation”

## Results

Number of errors: 24  
R-square = 0.520

## Errors per block

1	2 3 4
2	1 0 3
3	1 6 2 1
4	0 2 0 1
5	3 1 1 0

## Block Assignments:

- 1: Auditor President Council 5 Council 6 City Mayor
- 2: Executive Council 2 Council 3 Council 4 Prosecutor
- 3: Sheriff Council 1
- 4: Fr Council Fr Pres

## Blocked Adjacency Matrix

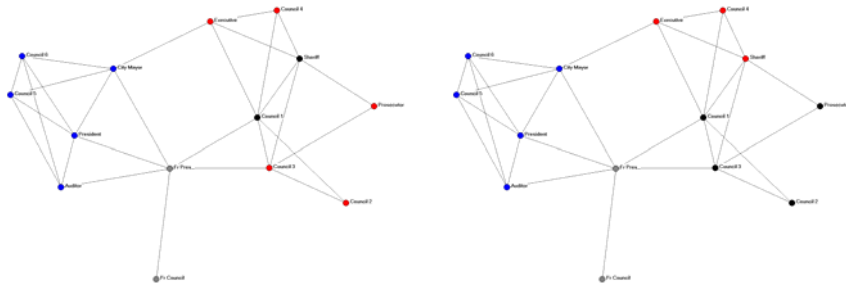
	1	1	1	1 1
	8 2 0 3 9	7 5 6 1 4	4 3	2 1
	P A C C C	C C C E P	C S	F F
8 President	1 1 1 1			1 1
2 Auditor	1 1 1			1 1
10 Council 6	1 1 1 1			1 1
13 City Mayor	1 1 1	1		1 1
9 Council 5	1 1 1 1			1 1
7 Council 4		1	1 1	1 1
5 Council 2		1	1	1 1
6 Council 3		1	1 1 1 1	1 1
1 Executive	1	1	1 1	1 1
14 Prosecutor		1	1	1 1
4 Council 1		1 1 1 1	1	1 1
3 Sheriff		1 1 1 1 1		1 1
12 Fr Pres	1 1 1	1	1	1 1
11 Fr Council				1 1

## Density Table

	1	2	3	4
1	0.90	0.04	0.00	0.30
2	0.04	0.30	0.80	0.10
3	0.00	0.80	1.00	0.25
4	0.30	0.10	0.25	1.00

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## Comparing CONCOR (left) and optimisation (right) solutions



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## Some practical issues

### Include the transpose for directed graphs?

Probably, provided that there are no missing respondents

### How many blocks?

Tricky question

Purpose is important:

- many blocks can be good for producing visualisations

- few blocks can be more useful if major variations in position of primary interest

Heuristics from cluster analysis

Embed within a statistical framework for a more principled approach

### Which algorithm?

CONCOR works remarkably well as a quick, exploratory tool

Optimisation is valuable in less exploratory contexts

A statistical framework can be exceptionally helpful

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# Regular equivalence on UCINET

Select:

Network

Roles & Positions

Maximal Regular

Optimization

And then specify Input Dataset and Number of blocks (3, here)

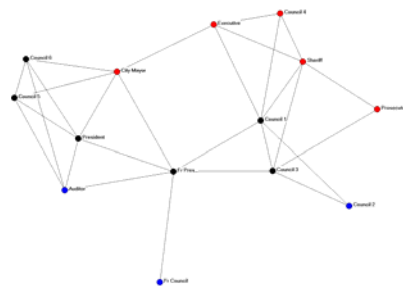
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# Regular Equivalence via Optimization

```

* REGULAR BLOCKMODELS VIA TABU SEARCH
* -----
*
* RESULTS:
* No. of errors: 0.000
*
* Blocked Adjacency Matrix
*
*      1 1 1 1 1
*      1 4 3 7 3 9 4 2 0 8 6 2 5 1
*      E P S C C C C F C P C A C F
* -----
* 1 Executive | 1 1 1 | 1 | | |
* 14 Prosecutor | 1 | | | 1 | |
* 3 Sheriff | 1 1 1 | 1 | 1 | |
* 7 Council 4 | 1 1 | | 1 | | |
* 13 City Mayor | 1 | | 1 1 1 1 | |
* -----
* 9 Council 5 | 1 | | 1 1 | 1 | |
* 4 Council 1 | 1 1 1 | | 1 | 1 | 1 |
* 12 Fr Pres | 1 | 1 | 1 1 | 1 1 | |
* 10 Council 6 | 1 | 1 | 1 | 1 | | |
* 8 President | 1 | 1 1 1 | | 1 | |
* 6 Council 3 | 1 1 | | 1 1 | | 1 |
* -----
* 2 Auditor | | 1 1 1 1 | | | |
* 5 Council 2 | | 1 | 1 | 1 | | |
* 11 Fr Council | | 1 | 1 | | | | |
* -----

```



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## References

- Borgatti, S., & Everett, M. E. (1989). The class of all regular equivalences: algebraic structure and computation. *Social Networks, 11*, 65-88.
- Breiger, R. L., Boorman, S. A., & Arabie, P. (1975). An algorithm for clustering relational data with applications to social network analysis and comparison with multidimensional scaling. *Journal of Mathematical Psychology, 12*, 328-382.
- Burt, R. S. (1976). Positions in networks. *Social Forces, 55*, 93-122.
- Everett, M. E., & Borgatti, S. (1996). Exact colorations of graphs and digraphs. *Social Networks, 18*, 319-331.
- Doreian, P., Batagelj, V., & Ferligoj, A. (2004). *Generalized blockmodeling*. Cambridge: Cambridge University Press.
- Doreian, P., Batagelj, V., & Ferligoj, A. (2005). Positional analyses of sociometric data. Pp. 77-97 in P. Carrington, J. Scott & S. Wasserman (Eds.) *Models and Methods in Social Network Analysis*. New York: Cambridge University Press.