An Introduction to Matrix Visualization & corrplot Package

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Content

Outline

Matrix Visualization

Matrix visualization is to convert a digital matrix to a graph.

- Presentation
  1. Glyph
  2. Color
  3. Other details
- **Model**
  1. Seriation (reordering) model
  2. Optimization algorithms
  3. Partition algorithms
- Goal
  1. Display data vividly
  2. Find the hidden pattern in data (clustering?)

Function in corrplot Package

Function:

- corrplot()
- corrplot.circle()
- corrplot.ellipse()
- corrplot.number()
- corrplot.pie()
- corrplot.shade()
- corrplot.square()
- corrplot.shade()
- corrplot.mtest()

Rforge: `http://r-forge.r-project.org/projects/corrplot/`

Blog: `http://taiyun.cos.name/wp-content/uploads/2009/10/corrplot.zip`

R Graph Gallery: `http://addictedtor.free.fr/graphiques/graphcode.php?graph=152`

PCA Order

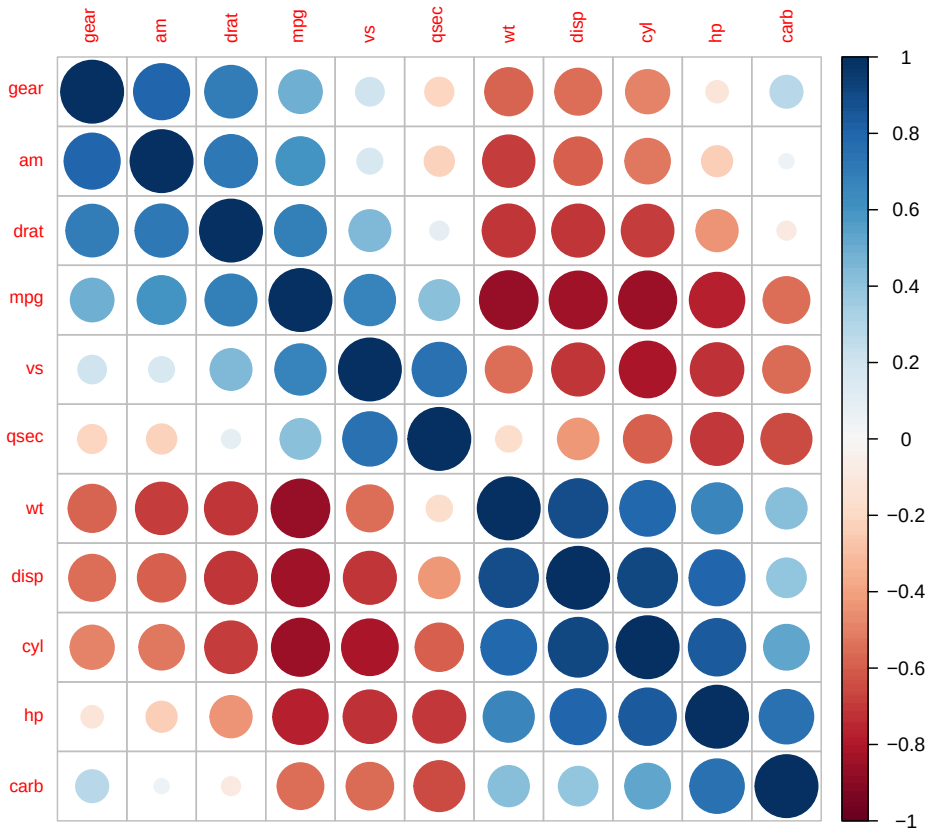


Figure: circle graph

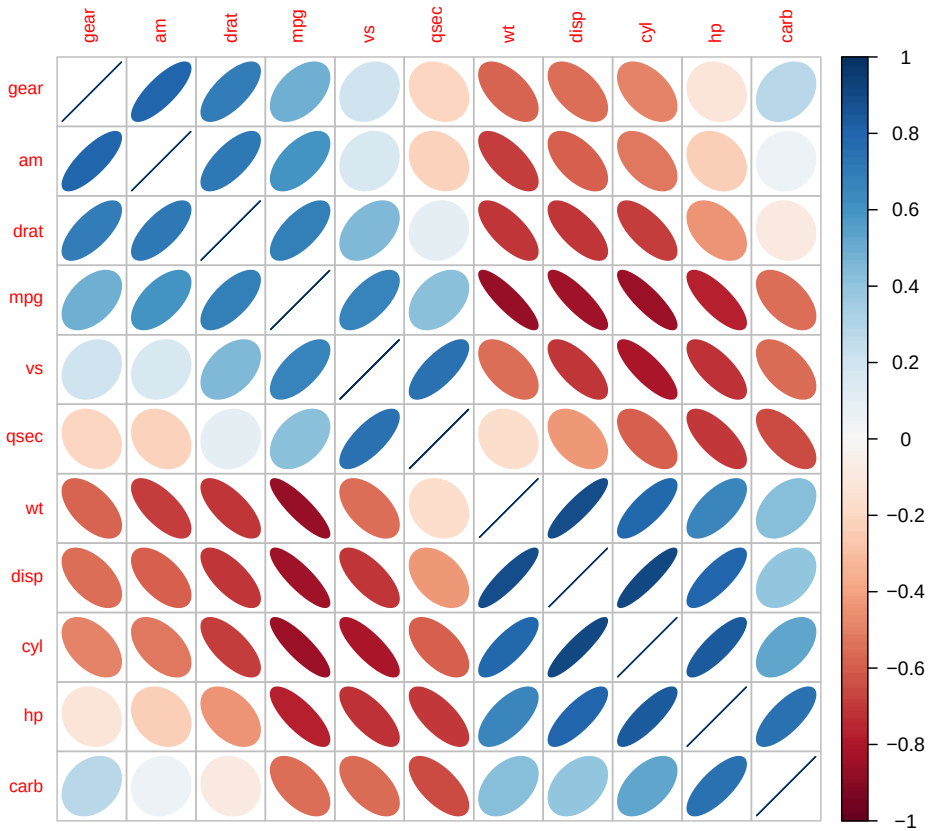


Figure: ellipse graph

HC Order (complete)

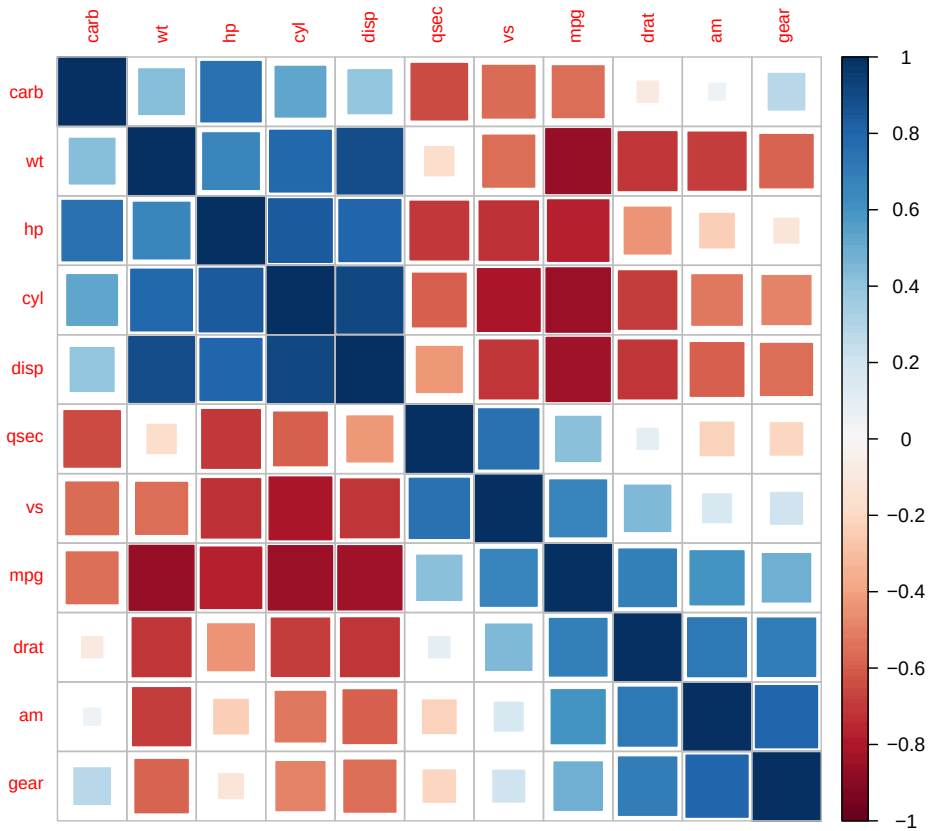


Figure: square graph

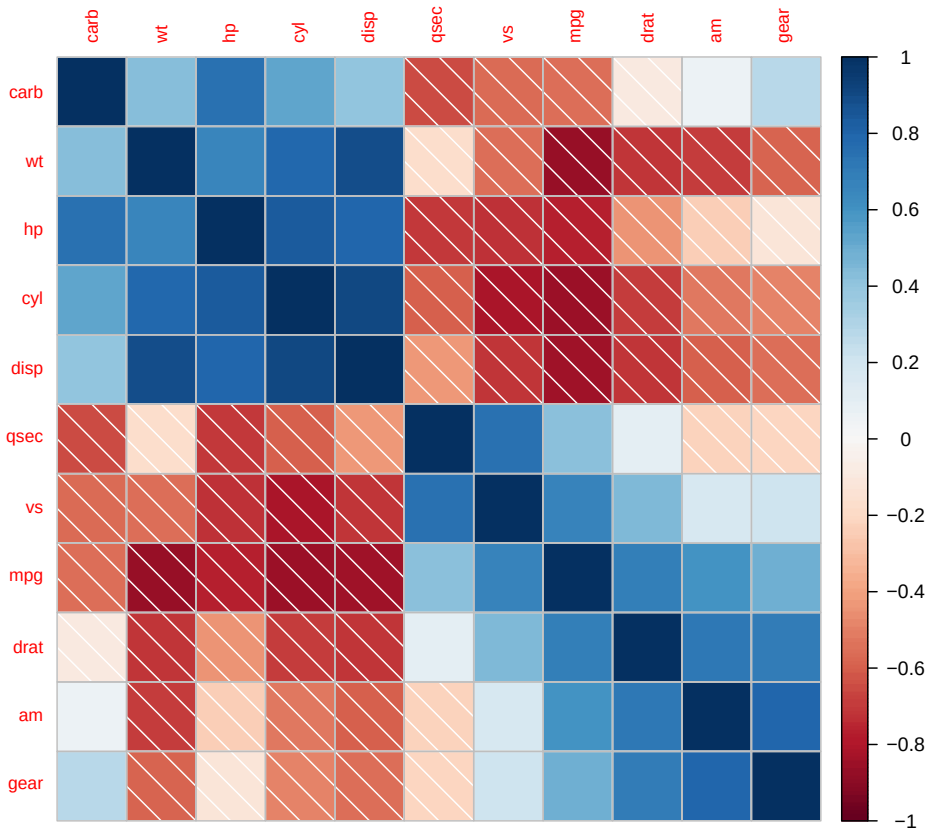


Figure: shade graph

Original Order

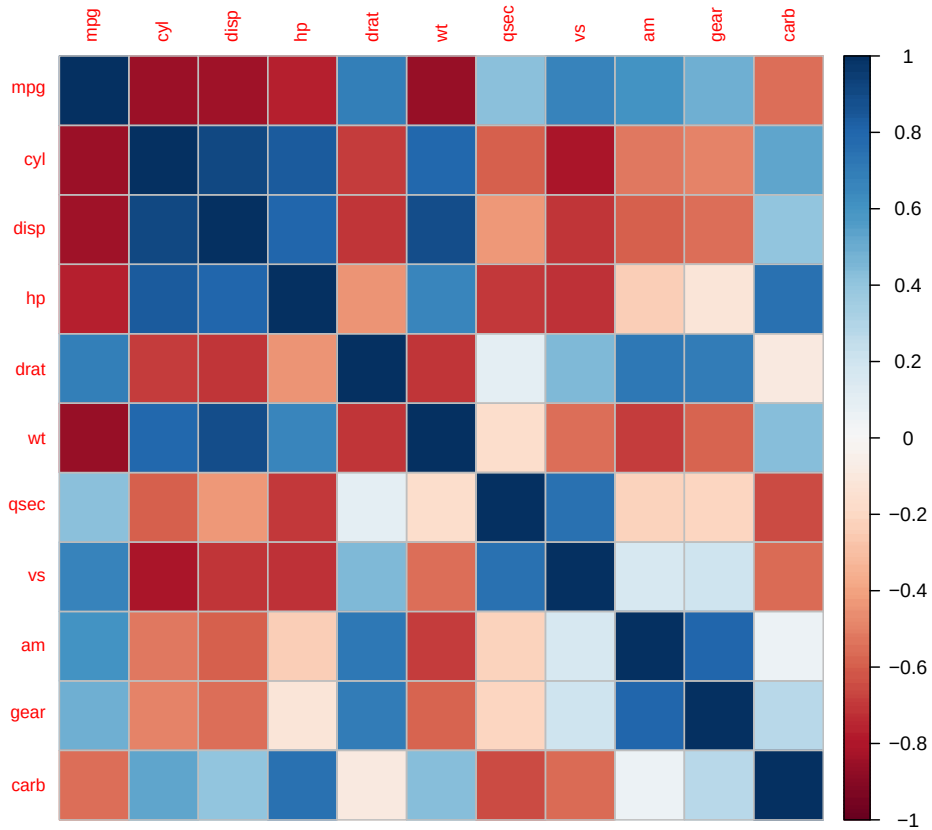


Figure: image graph

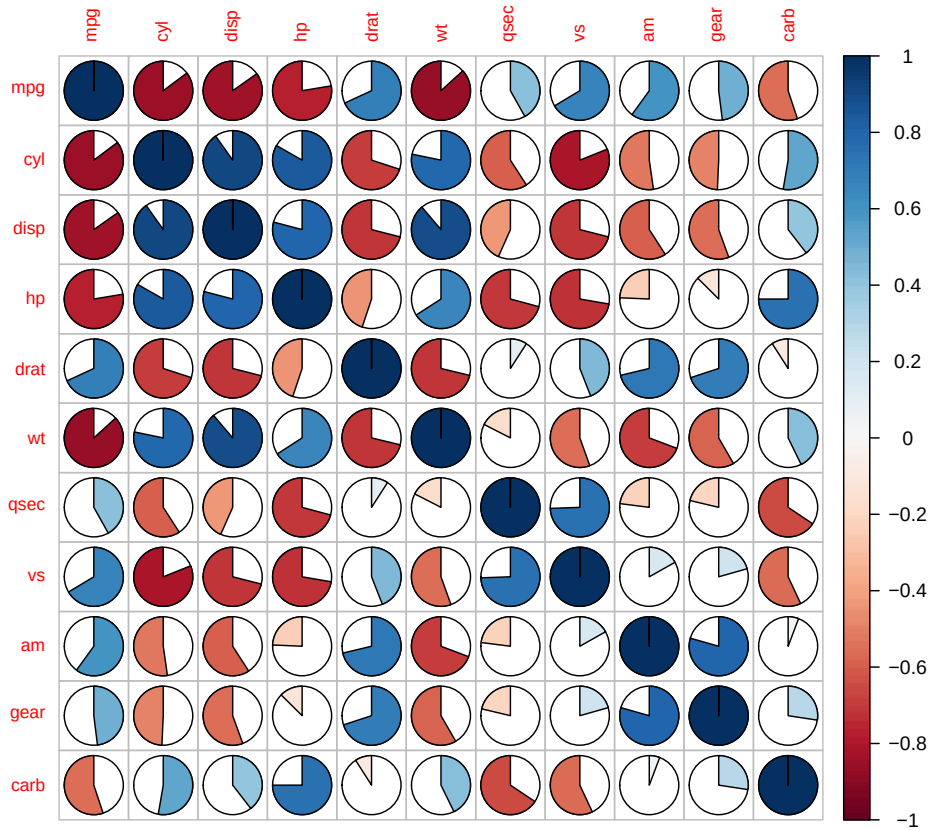


Figure: pie graph

Digital Matrix

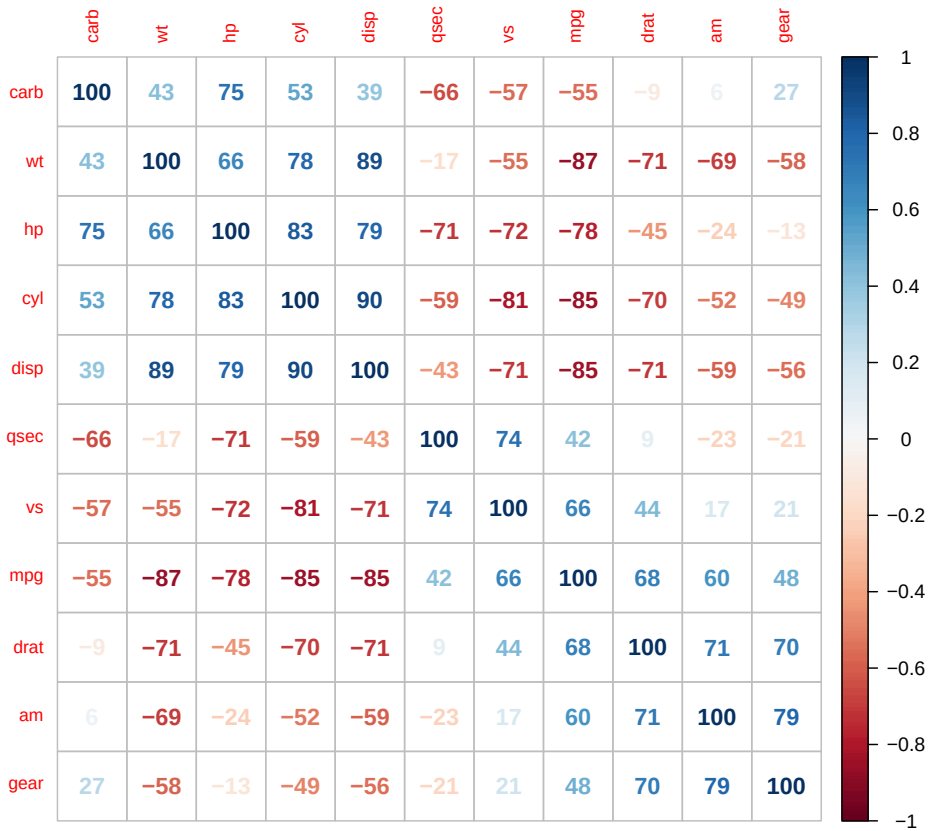


Figure: colored-digits graph

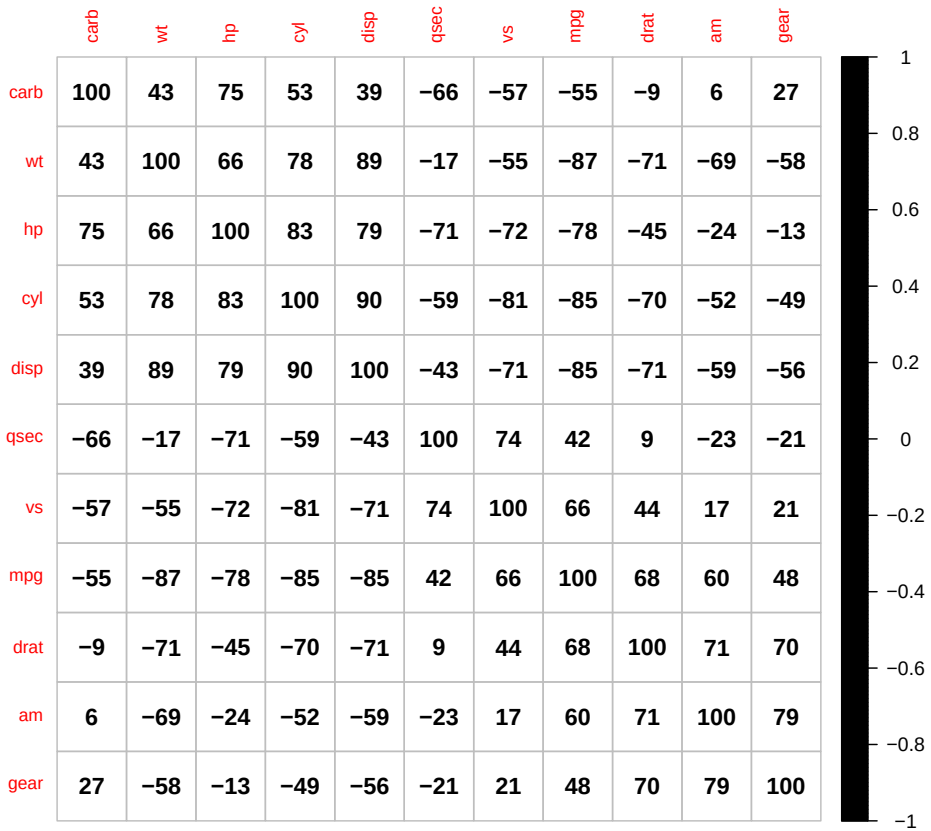


Figure: black-digits graph

Print in Black and White

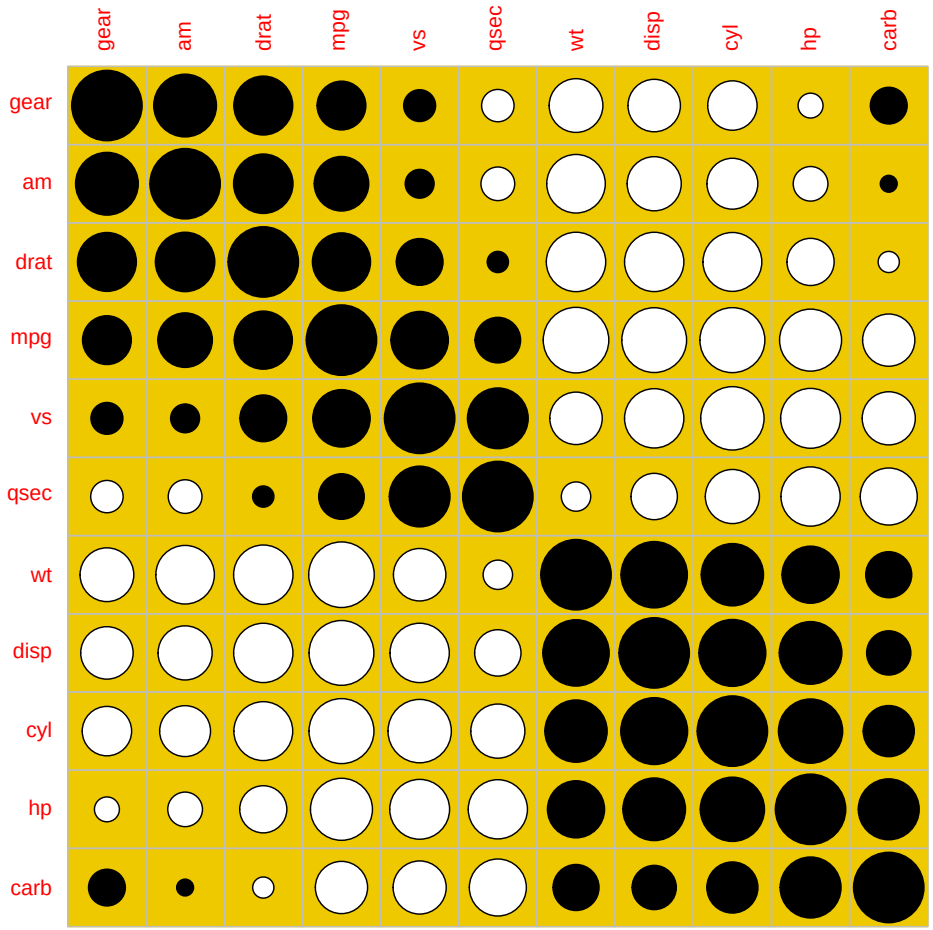


Figure: weiqi graph

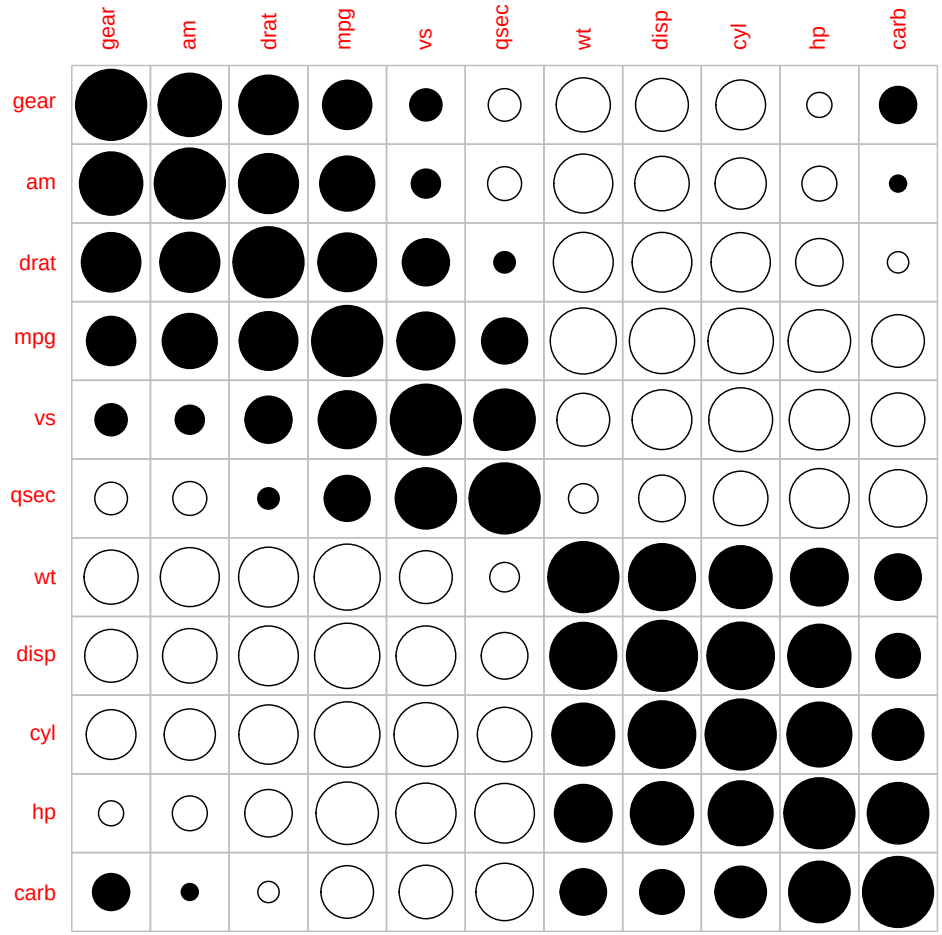


Figure: black-white graph

Test for Association/Correlation(α=0.05)

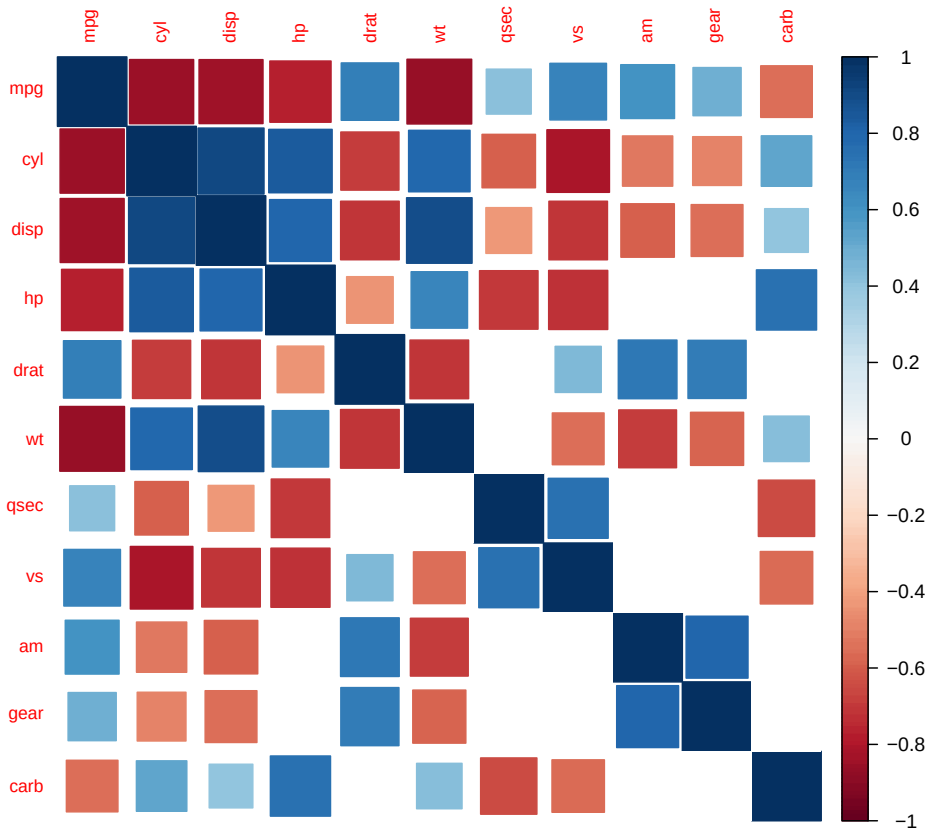


Figure: multi-correlation test (blank method)

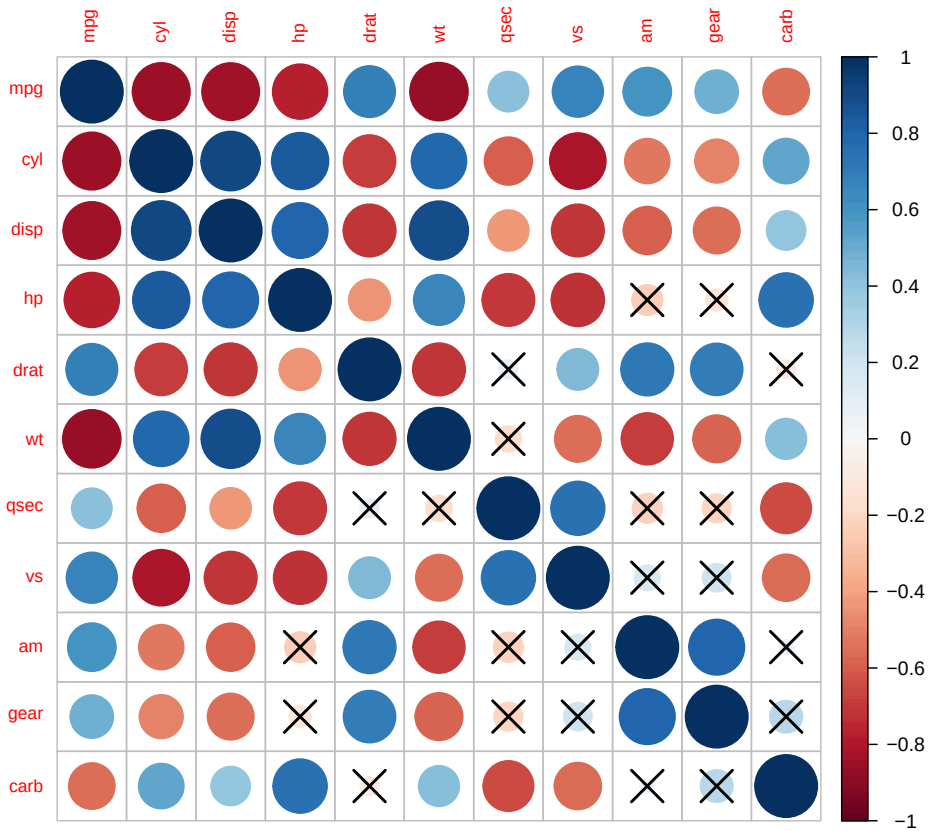


Figure: multi-correlation test (cross method)

Confidence Interval(95%)

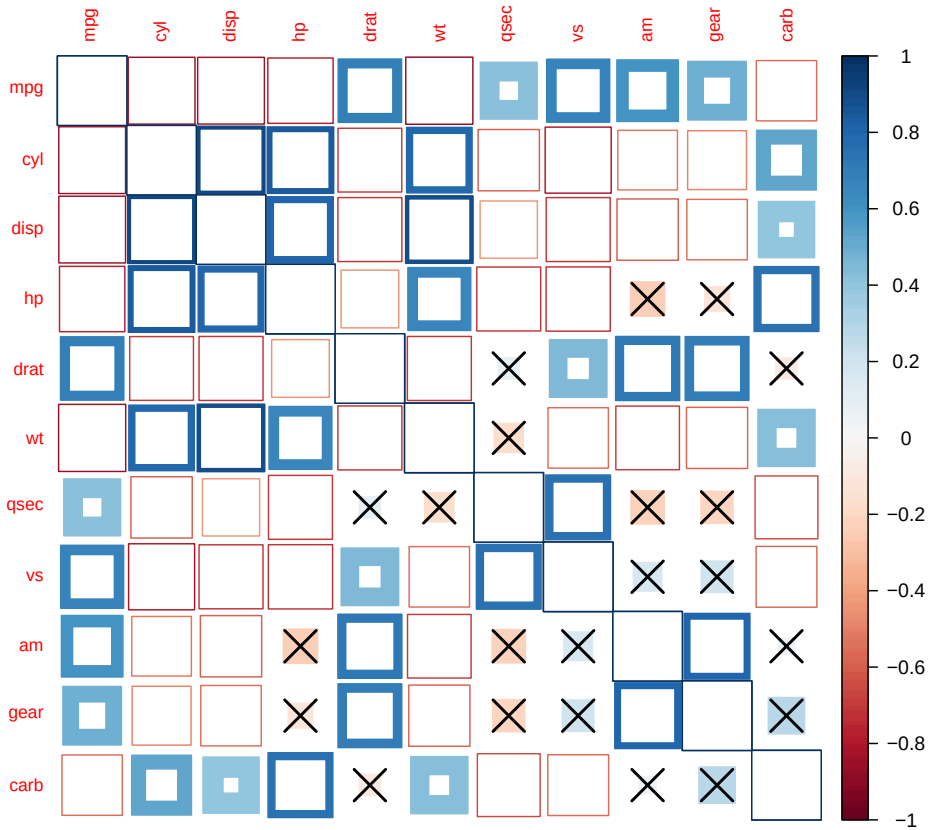


Figure: duo-square graph

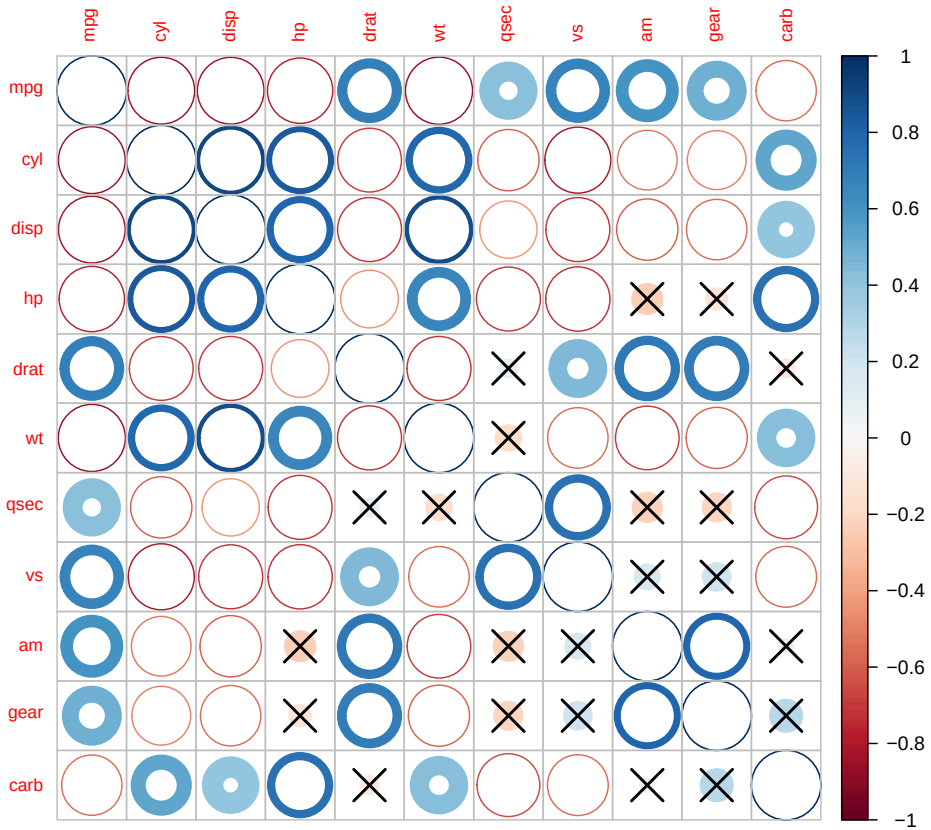


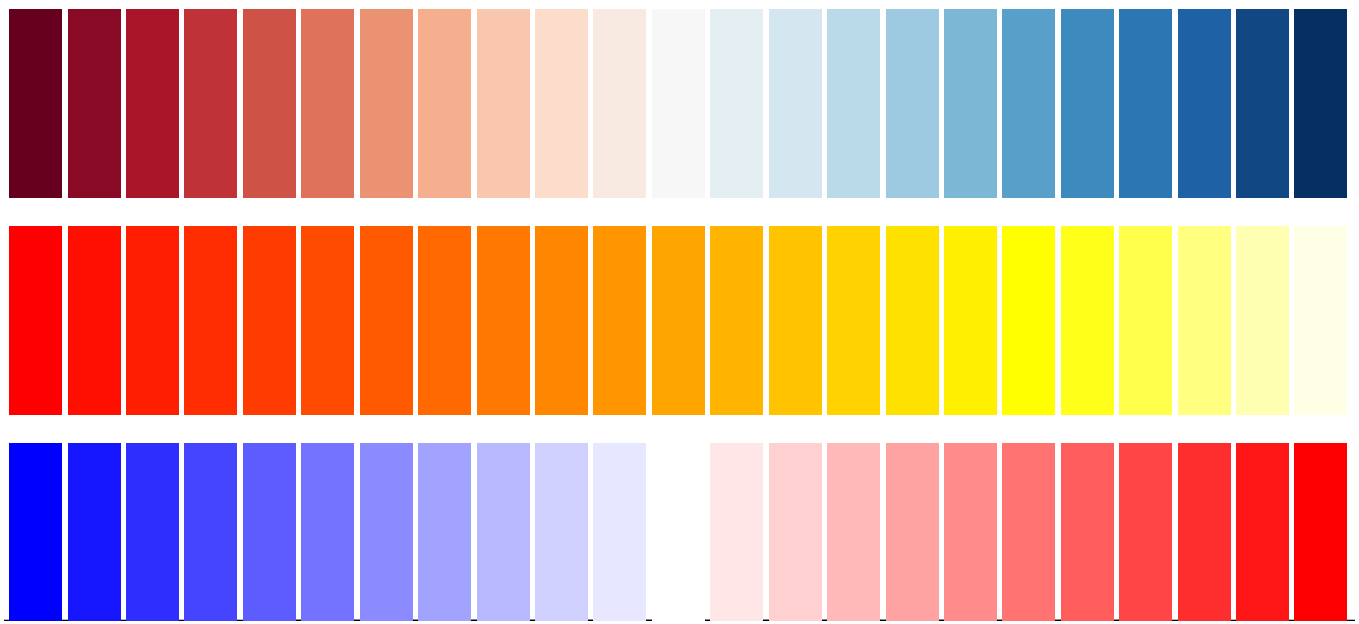
Figure: duo-circle graph

Choose Proper Color

- interpolate a set of given colors to create new color palettes

```
colorRamp(colors, bias = 1, space = c("rgb", "Lab"), ...)
colorRampPalette(colors, ...)
```

- Examples



Upper or Lower

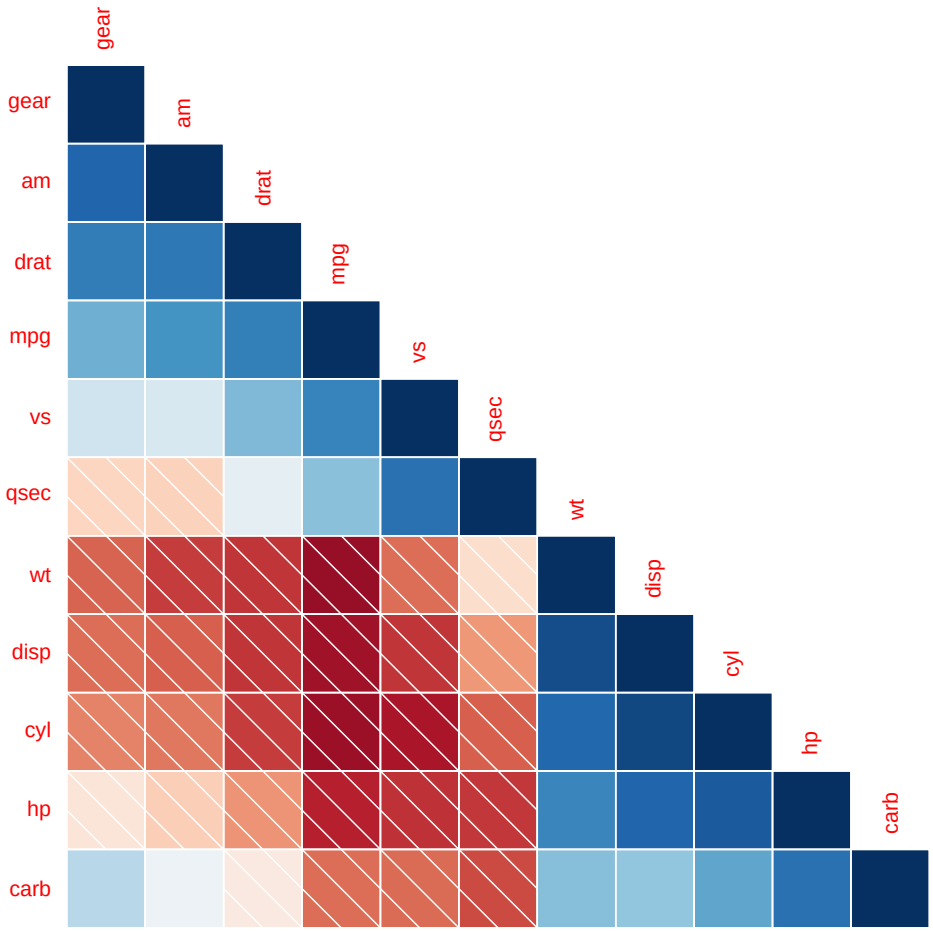


Figure: lower

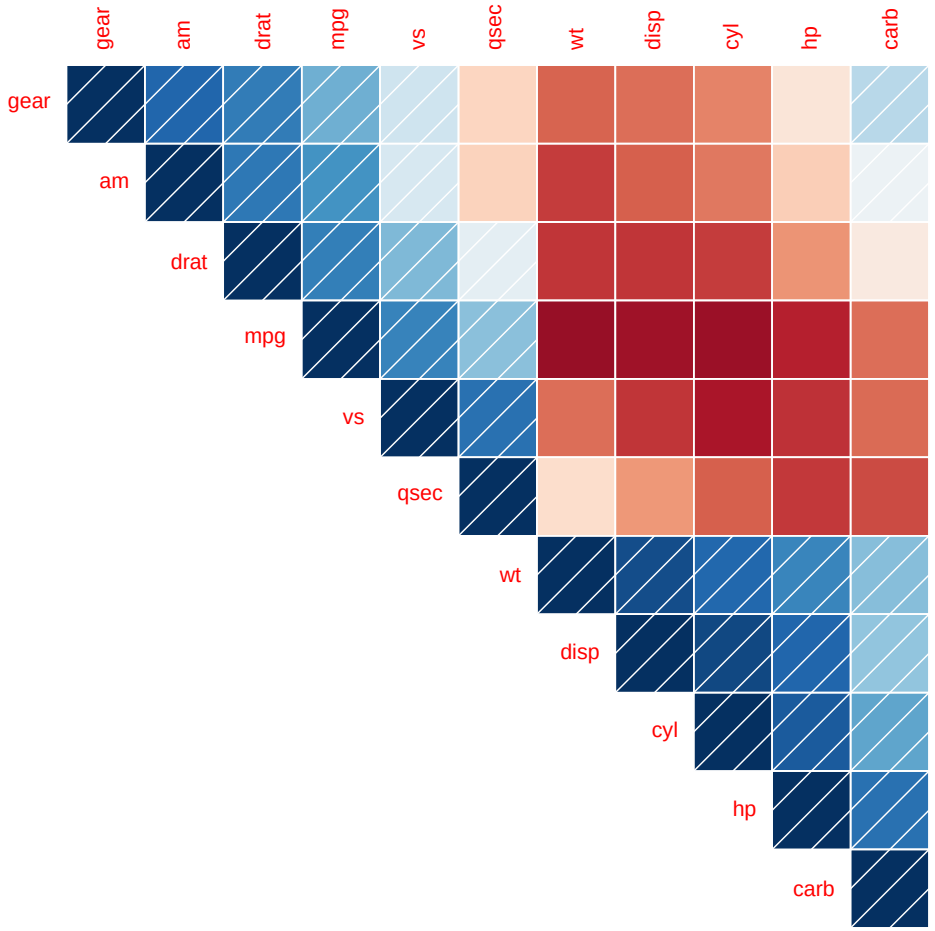


Figure: upper

Outline, colorkey, grid, text label, etc

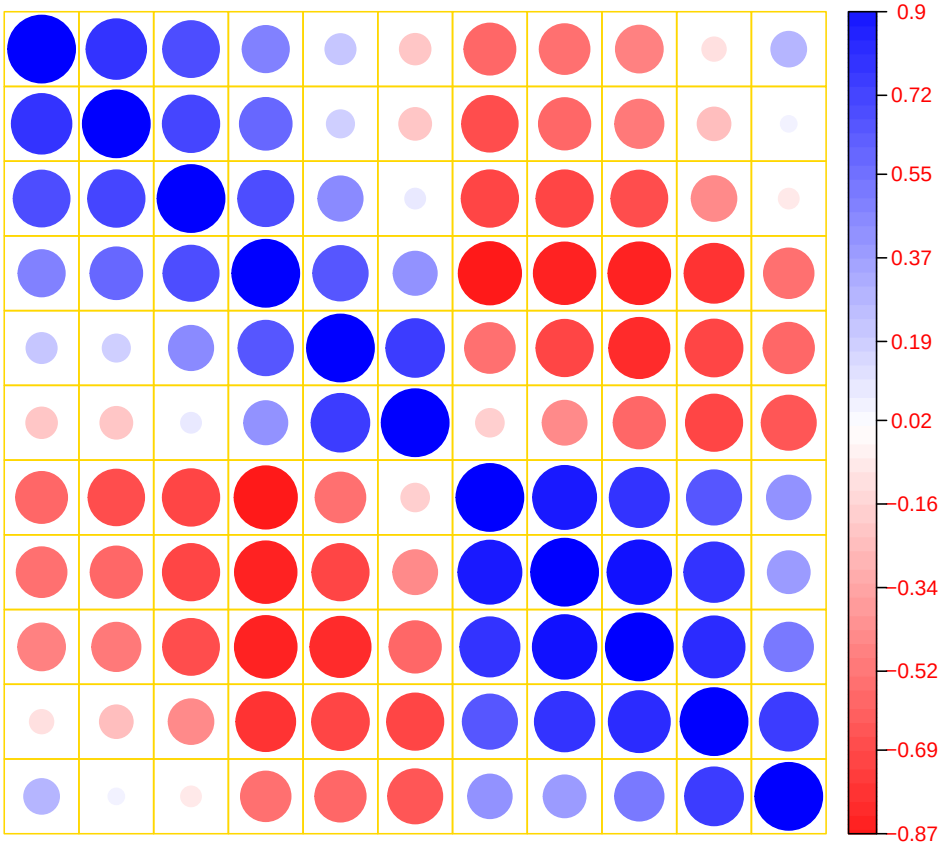


Figure: outline-0

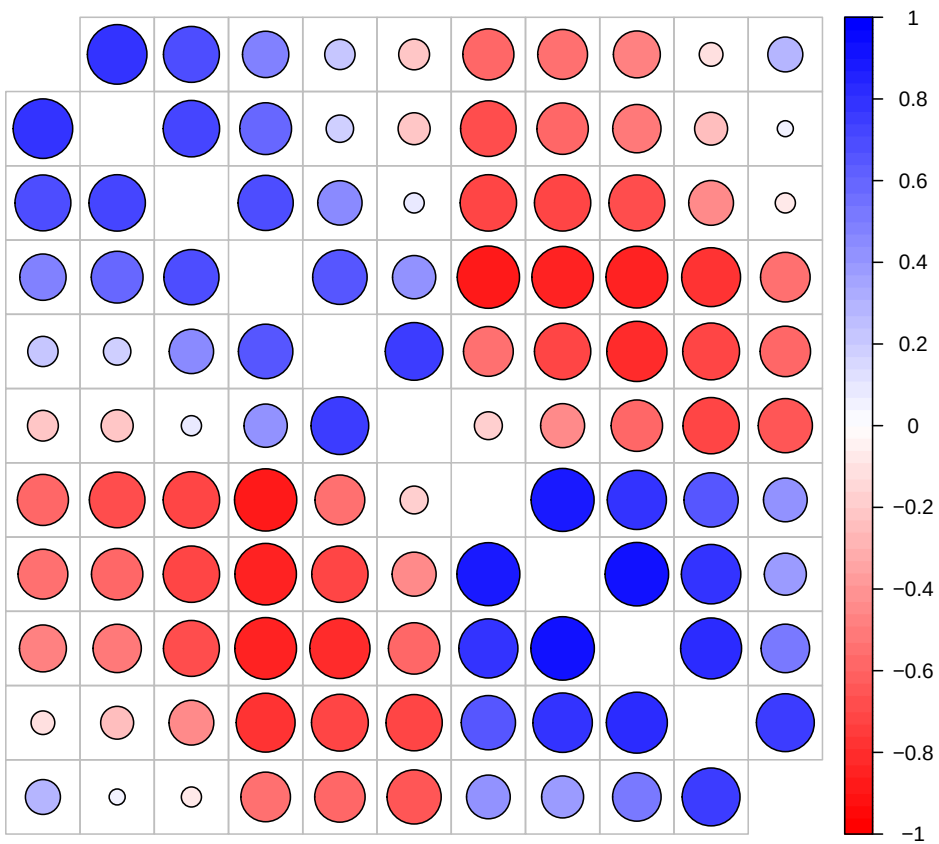


Figure: outline-1

Who cares *corrplot*?

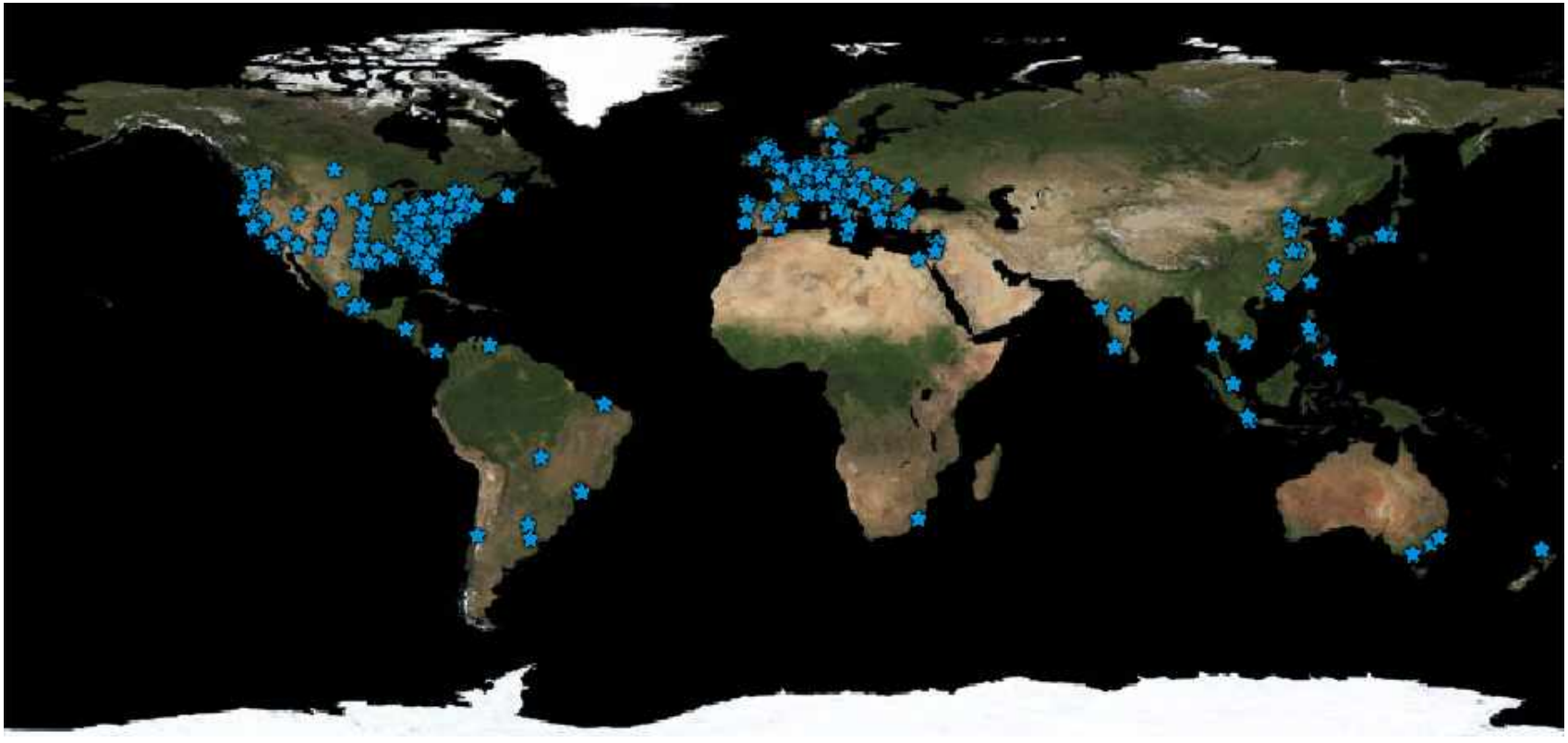


Figure: Visitor Map

Summary

- What can *corrplot* do?
  1. Basic seriation: HC, PCA, alphabet
  2. Display methods: circle, ellipse, square, etc
  3. Details: color, grid, colorkey, text-label, etc
- Advantages
  1. Creates nice and helpful pictures
  2. Flexible and good at details
  3. Easy and convenience: merely one function (about 400 lines)
- Disadvantages
  1. Lack seriation method
  2. Slow and sucks when handle large matrix
- How to get *corrplot*:
  1. From R-forge
  2. Ask me to send

Outline

Why need?

Get the hidden Structure and Pattern:

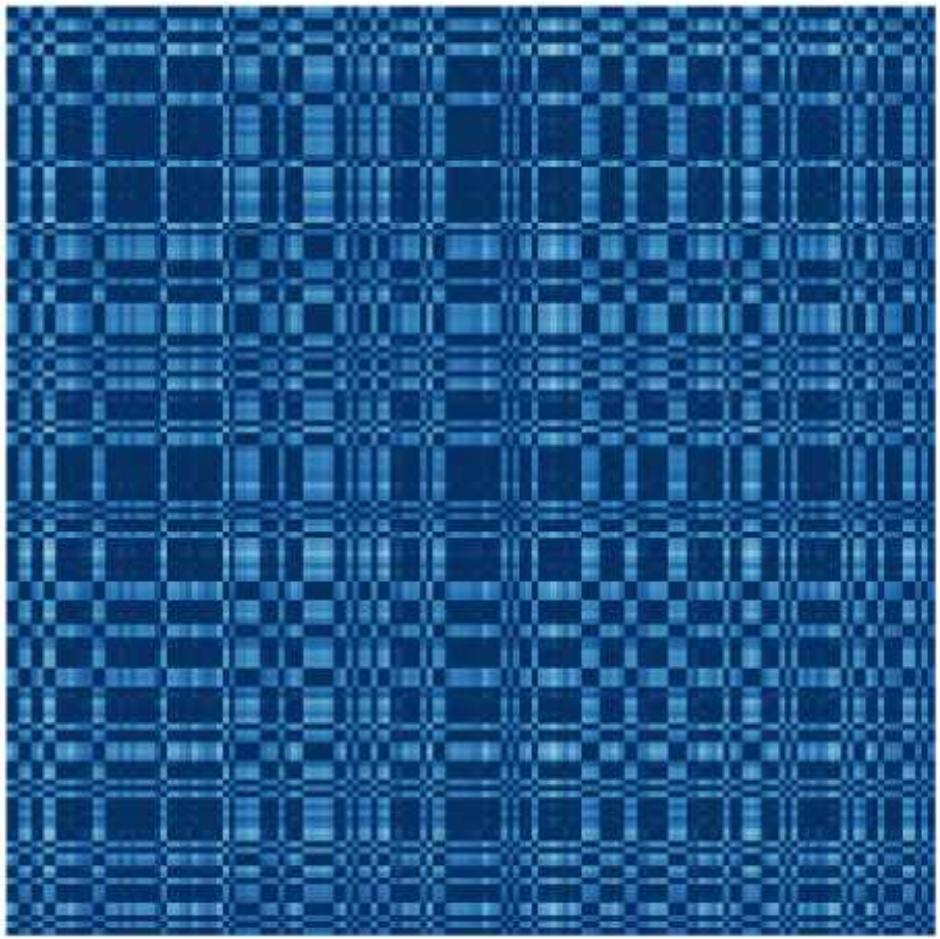


Figure: random

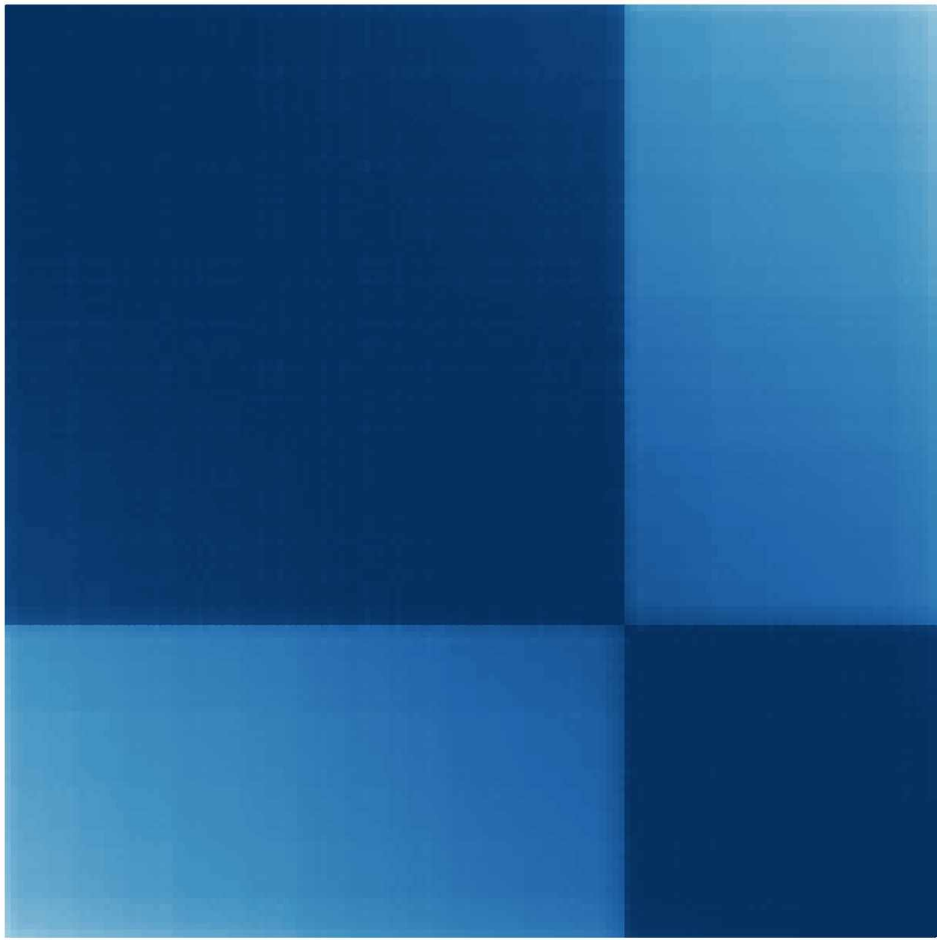


Figure: ordered

How to measure ?

Robinson Matrix and Anti-Robinson Matrix

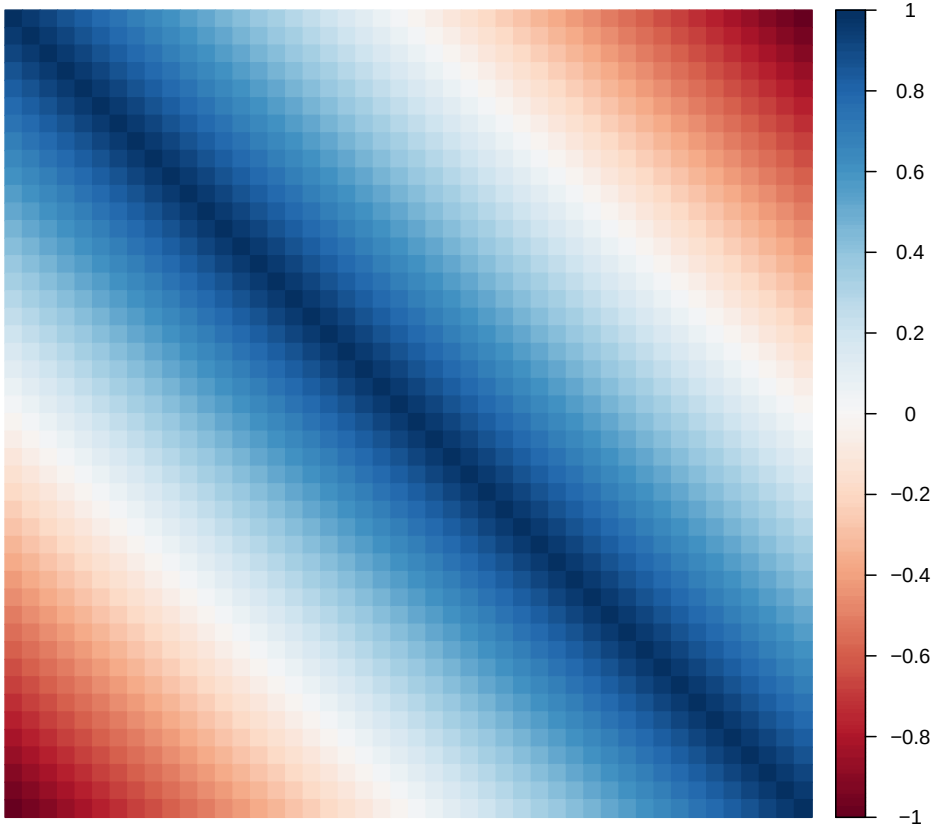


Figure: Robinson Matrix

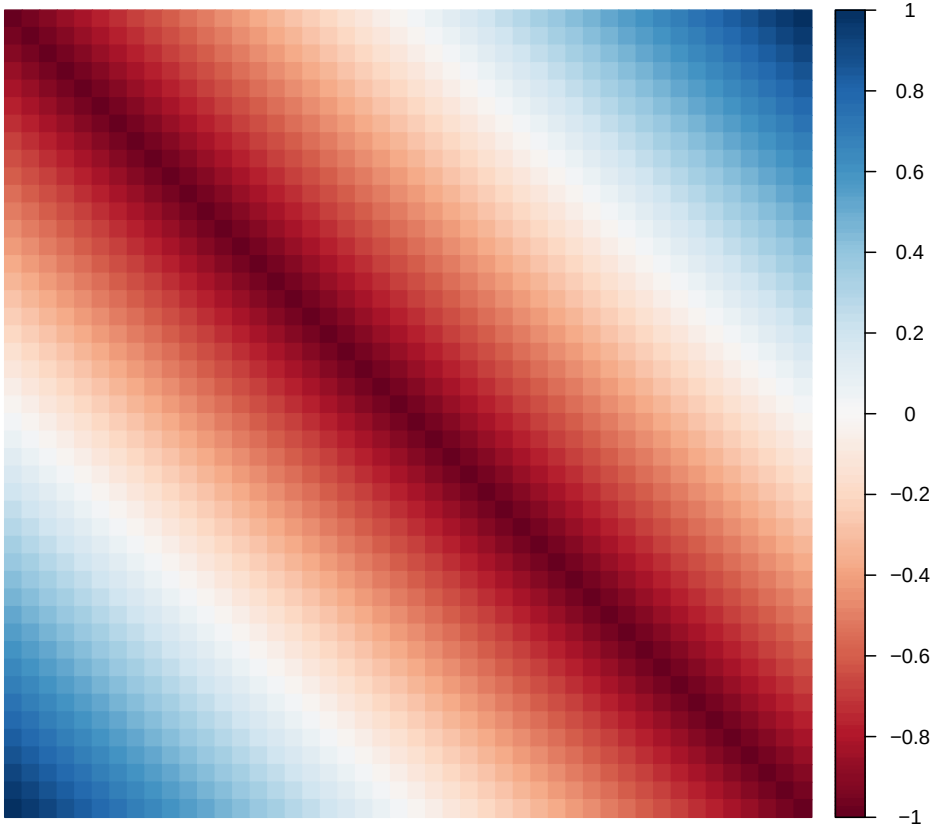


Figure: Anti Robinson Matrix

How to measure ?

Robinson Matrix and Pre-Robinson Matrix

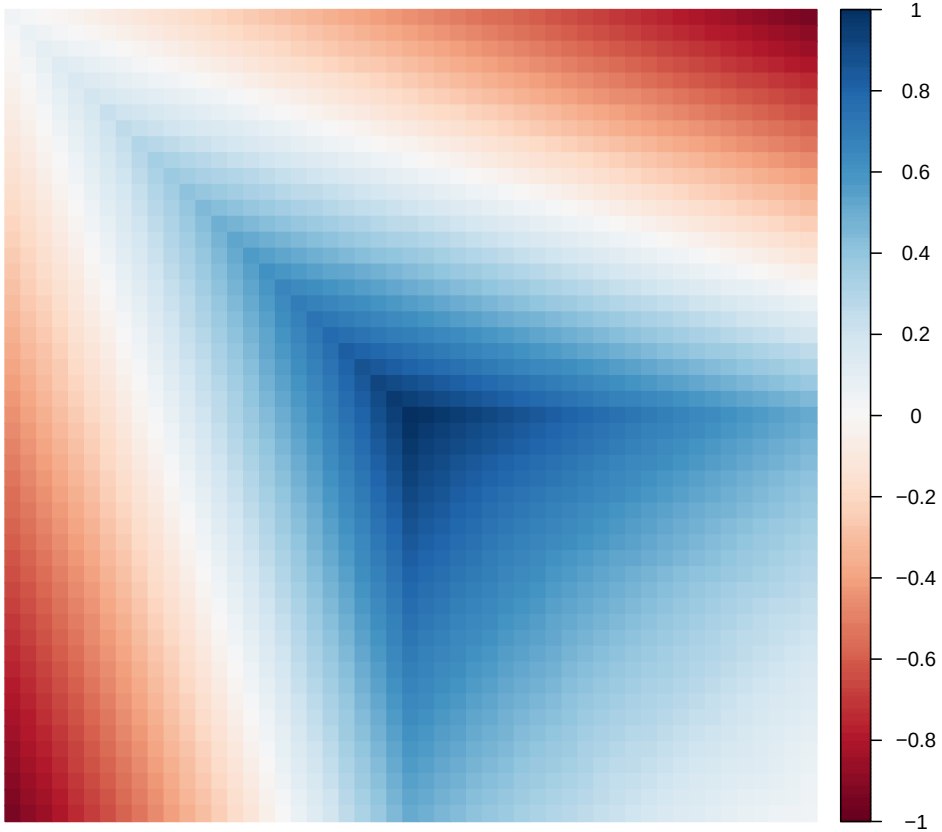


Figure: Robinson Matrix

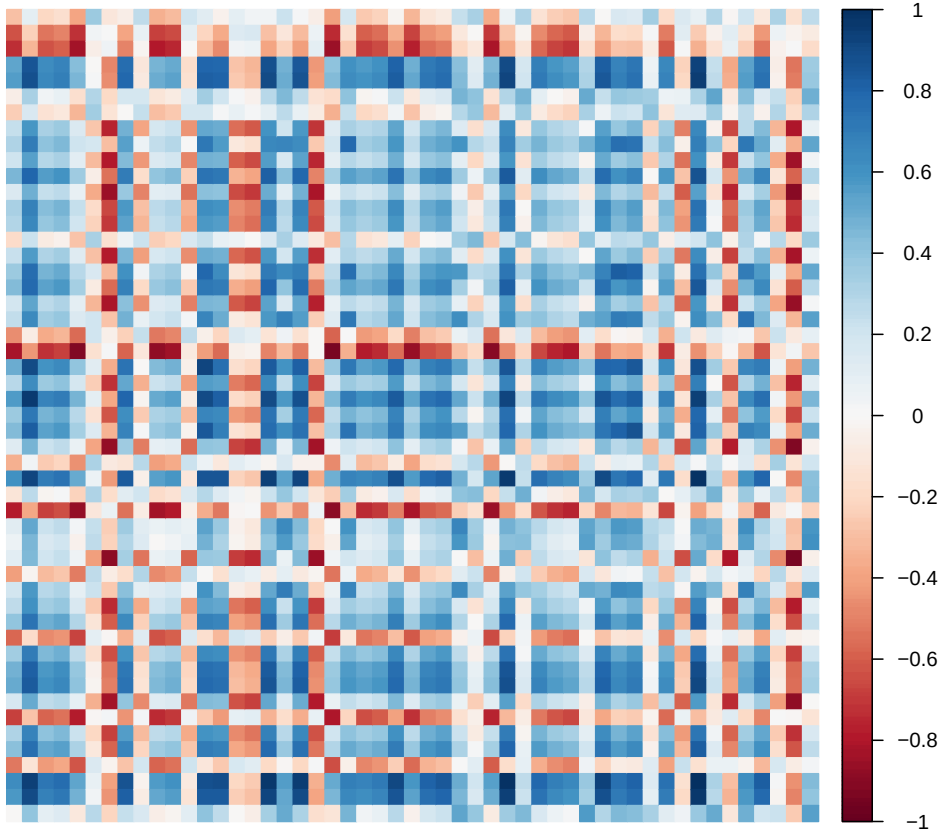


Figure: Pre Robinson Matrix

Combinatorial Optimization Model

- Anti-Robinson

$$L(\mathbf{D}) = \sum_{j<k<i} I(d_{ij} < d_{ik}) + \sum_{i<j<k} I(d_{ij} > d_{ik}) \tag{2.1}$$

- Hamiltonian path length

$$L(\mathbf{D}) = \sum_{i=1}^{n-1} d_{i,i+1} \tag{2.2}$$

- Inertia criterion

$$M(\mathbf{D}) = \sum_{i=1}^{n} \sum_{j=1}^{n} d_{ij} |i-j|^2 \tag{2.3}$$

- Least squares criterion

$$L(\mathbf{D}) = \sum_{i=1}^{n} \sum_{j=1}^{n} (d_{ij} - |i-j|)^2 \tag{2.4}$$

- Measure of effectives

$$M(\mathbf{X}) = \frac{1}{2}\sum_{i=1}^{n}\sum_{j=1}^{m} x_{ij}\left[x_{i,j+1} + x_{i,j-1} + x_{i+1,j} + x_{i-1,j}\right] \quad (2.5)$$

- Stress:

$$L(\mathbf{X}) = \sum_{i=1}^{n}\sum_{j=1}^{m} \sigma_{ij} \quad (2.6)$$

  - The Moore neighborhood:

$$\sigma_{ij} = \sum_{k=\max(1,i-1)}^{\min(n,i+1)} \sum_{l=\max(1,j-1)}^{\min(m,j+1)} (x_{ij} - x_{kl})^2 \quad (2.7)$$

  - The Neumann neighborhood :

$$\sigma_{ij} = \sum_{k=\max(1,i-1)}^{\min(n,i+1)} (x_{ij} - x_{kj})^2 + \sum_{l=\max(1,j-1)}^{\min(m,j+1)} (x_{ij} - x_{il})^2 \quad (2.8)$$

Reorder a matrix

- Five families of methods:
  1. Robinsonian：Ellipse seriation
  2. Dimension reduction：PCA, MDS
  3. Block modeling：Kmeans, Hierarchical clustering, etc
  4. Heuristics：SA, GA, PSO
  5. Graph methods：TSP
- Useful packages in R
  1. **seriation**
  2. **blockmodeling**
  3. **TSP**
  4. **Cairo**

seriation package

Table: Currently implemented methods in seriation package

Algorithm	`method`	Optimizes	Input data
Simulated annealing	`"ARSA"`	Gradient measure	`dist`
Branch-and-bound	`"BBURCG"`	Gradient measure	`dist`
Branch-and-bound	`"BBWRCG"`	Gradient measure (weighted)	`dist`
TSP solver	`"TSP"`	Hamiltonian path length	`dist`
Optimal leaf ordering	`"OLO"`	Hamiltonian path length	`dist`
Bond Energy Algorithm	`"BEA"`	Measure of effectiveness	`matrix`
TSP to optimize ME	`"BEA_TSP"`	Measure of effectiveness	`matrix`
Hierarchical clustering	`"HC"`	Other	`dist`
Gruvaeus and Wainer	`"GW"`	Other	`dist`
Rank-two ellipse seriation	`"Chen"`	Other	`dist`
MDS – first dimension	`"MDS"`	Other	`dist`
First principal component	`"PCA"`	Other	`matrix`

seriation package

Table: Implemented loss/merit functions in function `criterion`.

Name	`method`	merit/loss	Input data
Anti-Robinson events	`"AR_events"`	loss	`dist`
Anti-Robinson deviations	`"AR_deviations"`	loss	`dist`
Gradient measure	`"Gradient_raw"`	merit	`dist`
Gradient measure (weighted)	`"Gradient_weighted"`	merit	`dist`
Hamiltonian path length	`"Path_length"`	loss	`dist`
Inertia criterion	`"Inertia"`	merit	`dist`
Least squares criterion	`"Least_squares"`	loss	`dist`
Measure of effectiveness	`"ME"`	merit	`matrix`
Stress (Moore neighborhood)	`"Moore_stress"`	loss	`matrix`
Stress (Neumann neighborhood)	`"Neumann_stress"`	loss	`matrix`

Outline

《统计建模与R软件》Section 3.4

	FL	APP	AA	LA	SC
FL	1.00000000	0.2388057	0.044040889	0.306313037	0.092144656
APP	0.23880573	1.0000000	0.123419296	0.379614151	0.430769427
AA	0.04404089	0.1234193	1.000000000	0.001589766	0.001106763
LA	0.30631304	0.3796142	0.001589766	1.000000000	0.302439887
SC	0.09214466	0.4307694	0.001106763	0.302439887	1.000000000
LC	0.22843205	0.3712589	0.076824494	0.482774928	0.807545017
HON	-0.10674947	0.3536910	-0.030269601	0.645408595	0.410090809
SMS	0.27069919	0.4895490	0.054727421	0.361643880	0.799630538
EXP	0.54837963	0.1409249	0.265585352	0.140723415	0.015125832
DRV	0.34557633	0.3405493	0.093522030	0.393164148	0.704340067
AMB	0.28464484	0.5496359	0.044065981	0.346555034	0.842122228
GSP	0.33820196	0.5062987	0.197504552	0.502809305	0.721108973
POT	0.36745292	0.5073769	0.290032151	0.605507554	0.671821239
KJ	0.46720619	0.2840928	-0.323319352	0.685155768	0.482455962
SUIT	0.58591822	0.3842084	0.140017368	0.326957419	0.250283416

	LC	HON	SMS	EXP	DRV
FL	0.2284320	-0.106749472	0.27069919	0.54837963	0.34557633
APP	0.3712589	0.353690969	0.48954902	0.14092491	0.34054927
AA	0.0768245	-0.030269601	0.05472742	0.26558535	0.09352203
LA	0.4827749	0.645408595	0.36164388	0.14072342	0.39316415
SC	0.8075450	0.410090809	0.79963054	0.01512583	0.70434007
LC	1.0000000	0.355844464	0.81802080	0.14720197	0.69751518
HON	0.3558445	1.000000000	0.23990754	-0.15593849	0.28018499
SMS	0.8180208	0.239907539	1.00000000	0.25541758	0.81473421
EXP	0.1472020	-0.155938495	0.25541758	1.00000000	0.33722821

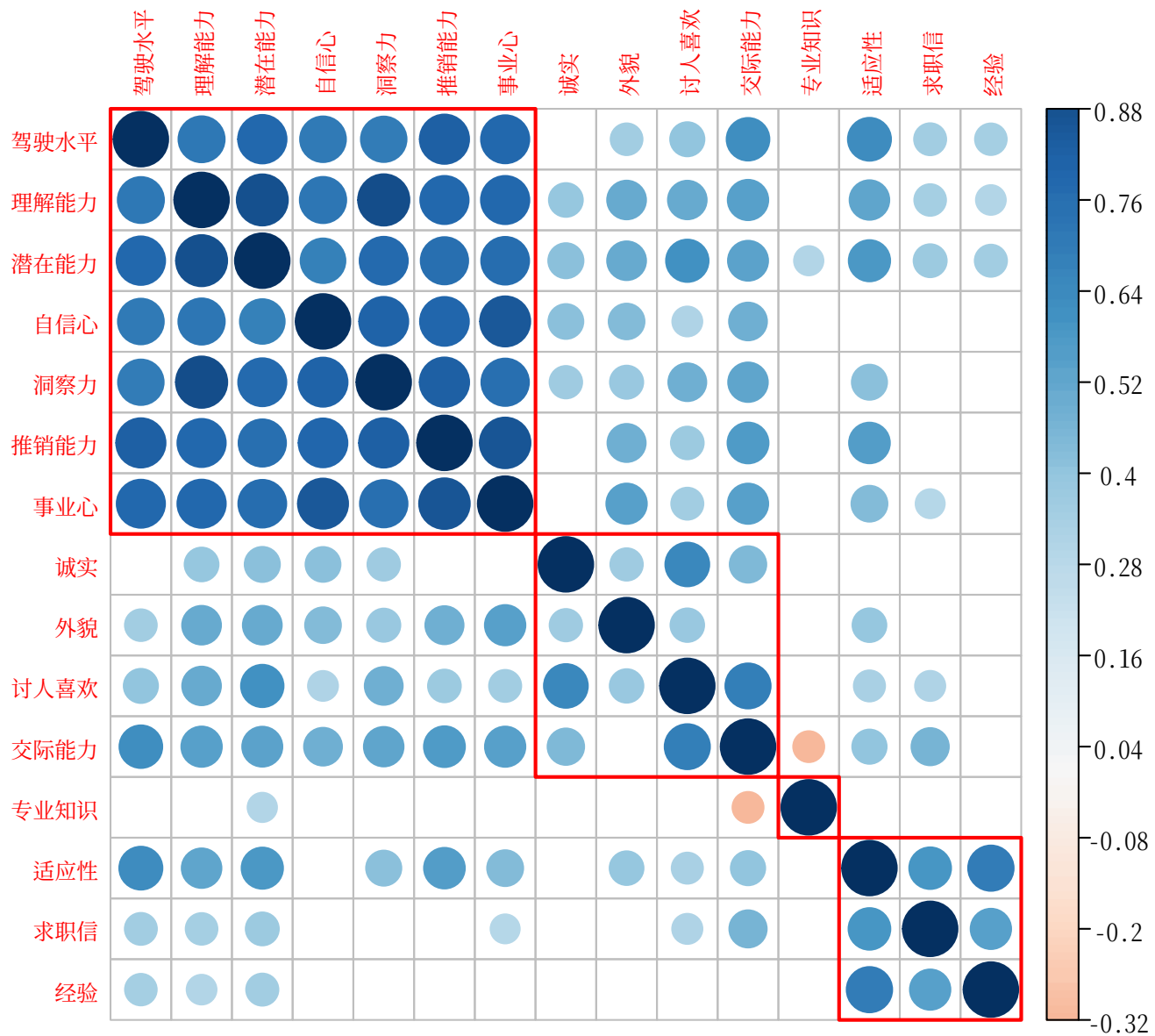
	AMB	GSP	POT	KJ	SUIT
DRV	0.6975152	0.280184989	0.81473421	0.33722821	1.00000000
AMB	0.7575421	0.214606359	0.85952656	0.19548192	0.78032317
GSP	0.8828486	0.385821758	0.78212322	0.29926823	0.71407319
POT	0.7773162	0.415657447	0.75360983	0.34833878	0.78840024
KJ	0.5268356	0.448245522	0.56328419	0.21495316	0.61280767
SUIT	0.4161447	0.002755617	0.55803585	0.69263617	0.62255406

	AMB	GSP	POT	KJ	SUIT
FL	0.28464484	0.3382020	0.3674529	0.4672062	0.585918216
APP	0.54963595	0.5062987	0.5073769	0.2840928	0.384208365
AA	0.04406598	0.1975046	0.2900322	-0.3233194	0.140017368
LA	0.34655503	0.5028093	0.6055076	0.6851558	0.326957419
SC	0.84212223	0.7211090	0.6718212	0.4824560	0.250283416
LC	0.75754208	0.8828486	0.7773162	0.5268356	0.416144671
HON	0.21460636	0.3858218	0.4156574	0.4482455	0.002755617
SMS	0.85952656	0.7821232	0.7536098	0.5632842	0.558035847
EXP	0.19548192	0.2992682	0.3483388	0.2149532	0.692636173
DRV	0.78032317	0.7140732	0.7884002	0.6128077	0.622554062
AMB	1.00000000	0.7838707	0.7688695	0.5471256	0.434768242
GSP	0.78387073	1.0000000	0.8758309	0.5494076	0.527816315
POT	0.76886954	0.8758309	1.0000000	0.5393968	0.573873154
KJ	0.54712558	0.5494076	0.5393968	1.0000000	0.395798842
SUIT	0.43476824	0.5278163	0.5738732	0.3957988	1.000000000

为了便于选择哪些变量是相关的,将上述相关矩阵中相关系数的绝对值 ≥ 0.5 的值画上下划线.

下面将变量分组，分组的原则是：同一组中变量之间的相关系数尽可能的高，而不同组间的相关系数尽可能的低. 从相关系数最大的变量开始，LC(洞察力) 与 GSP(理解能力) 的相关系数是 0.882, GSP 与 POT(潜在能力) 的相关系数

A picture is worth a thousand words!



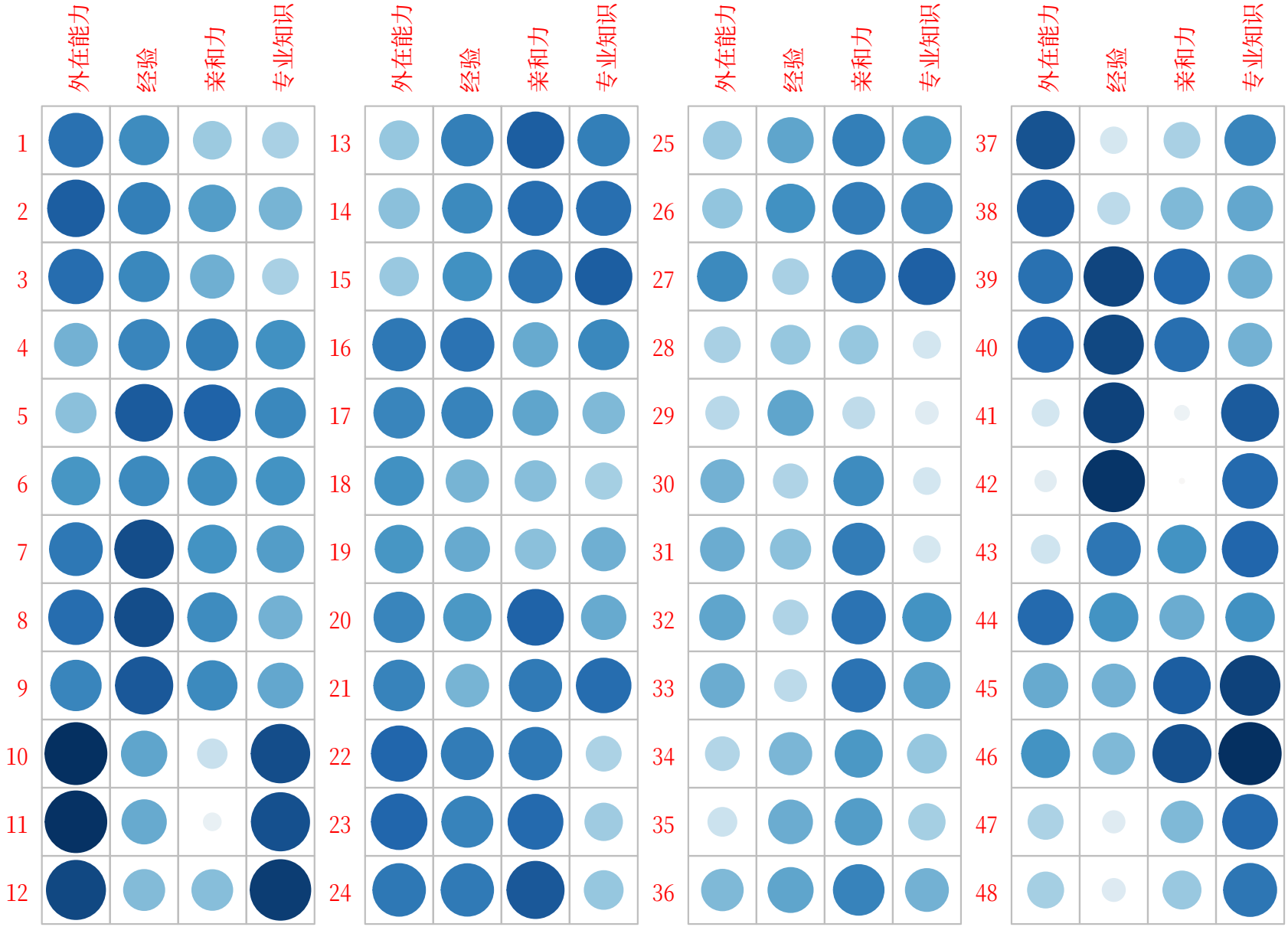
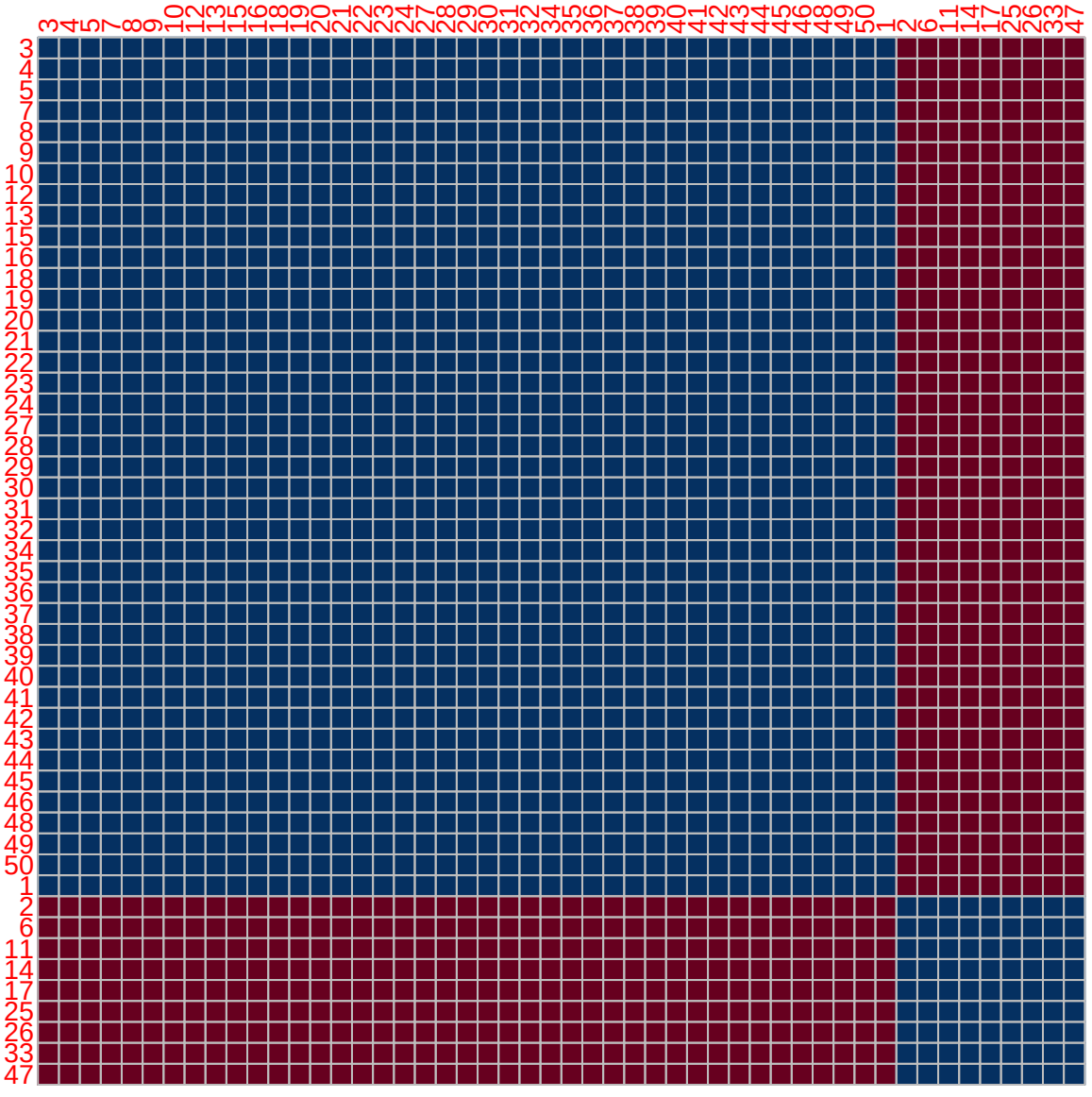


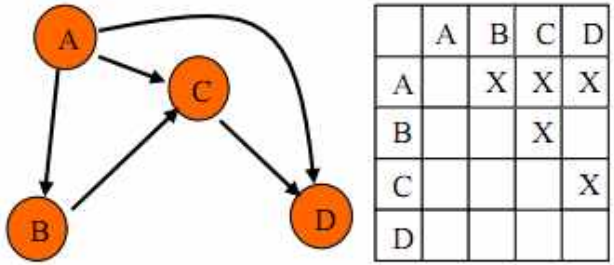
Figure: Factor Scores

Outlier Detection

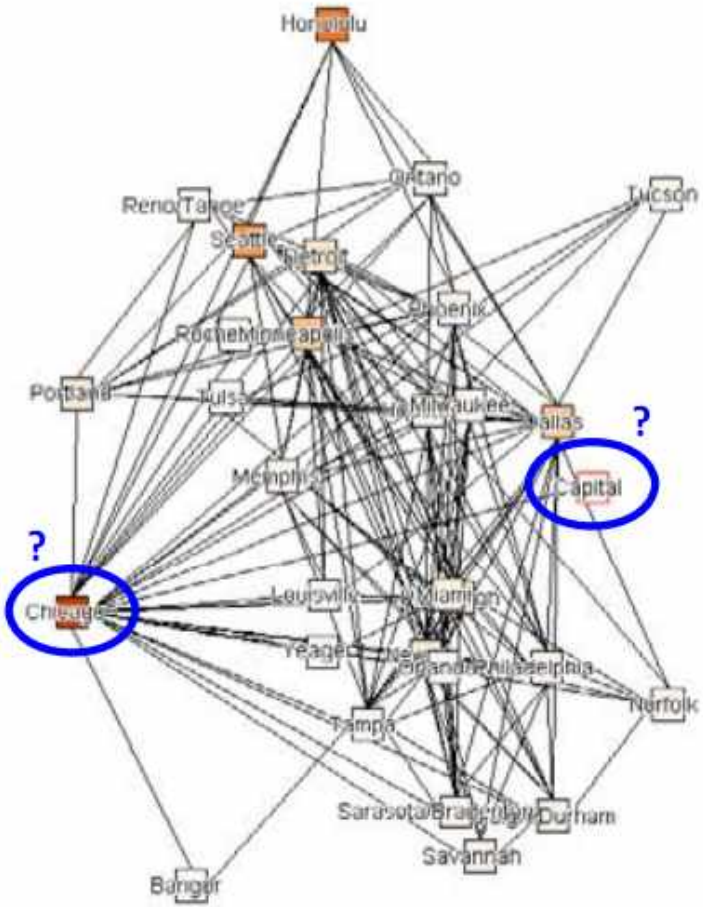


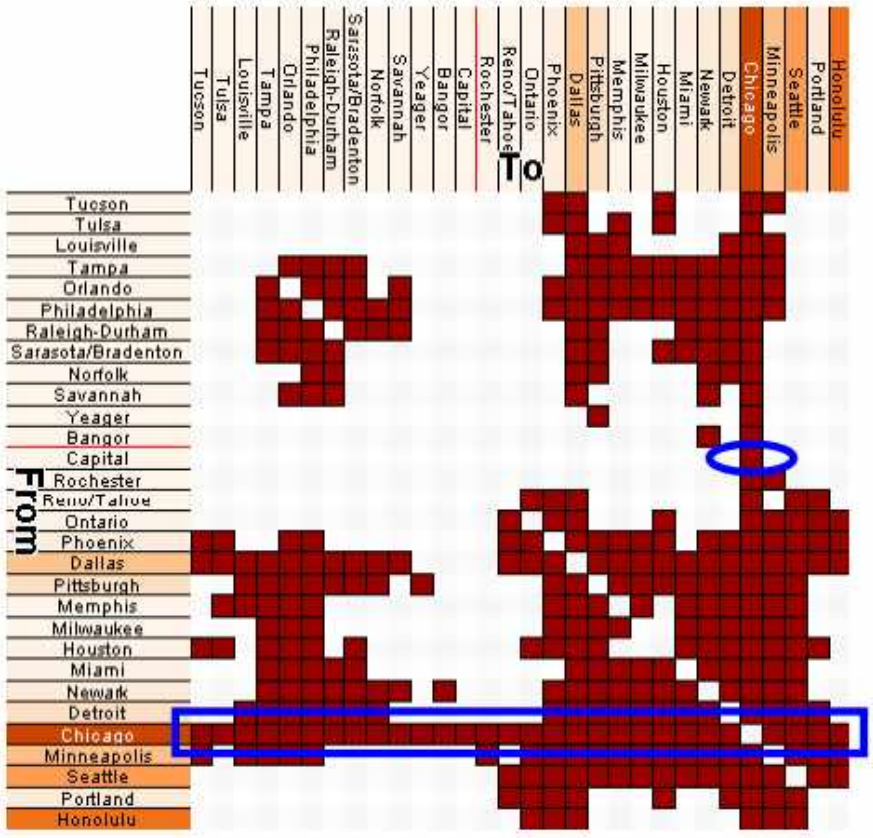
Social Networks Analysis

Matrix Visualization

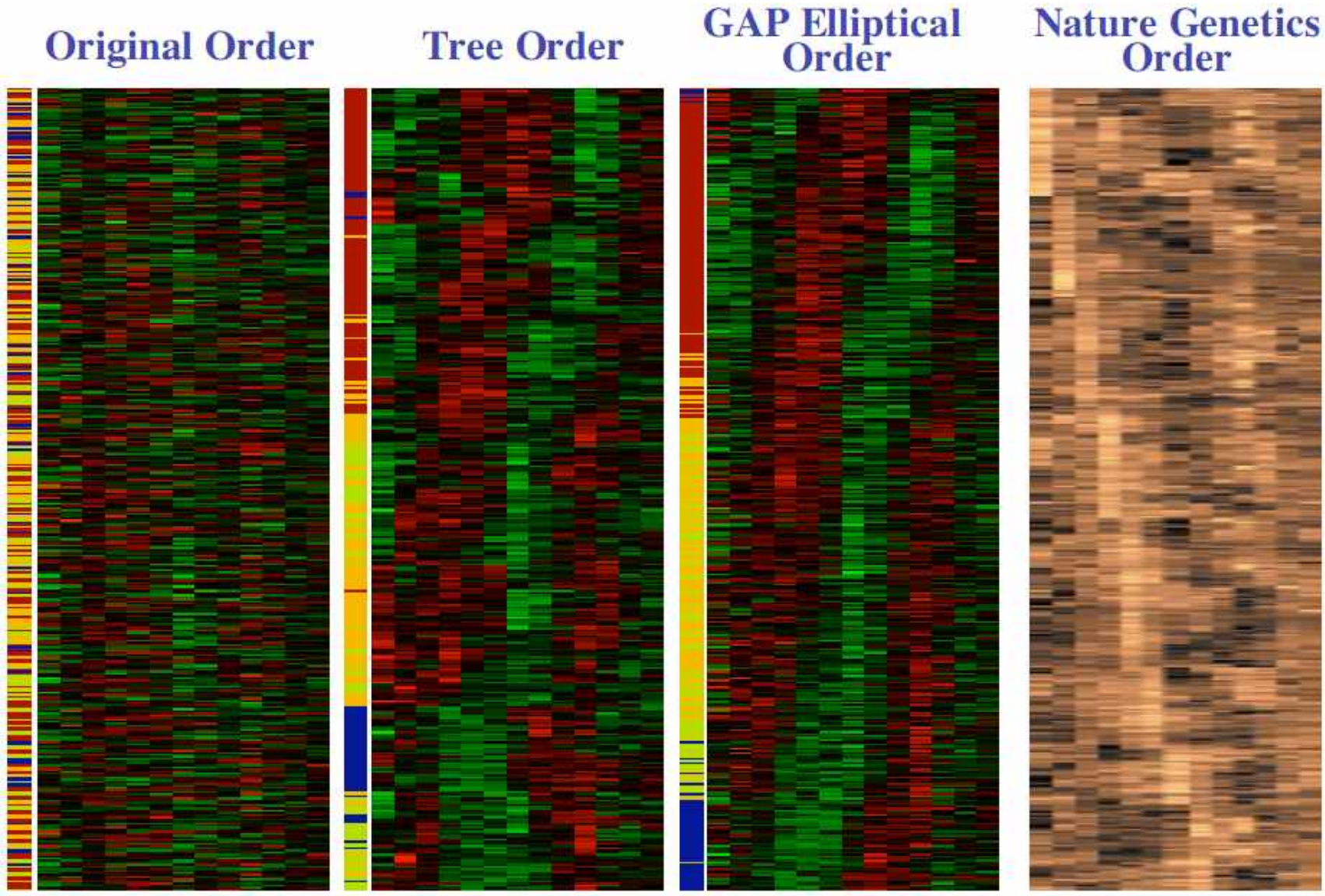


	A	B	C	D
A		X	X	X
B			X	
C				X
D				





cDNA Microarray Analysis



cDNA Microarray Analysis

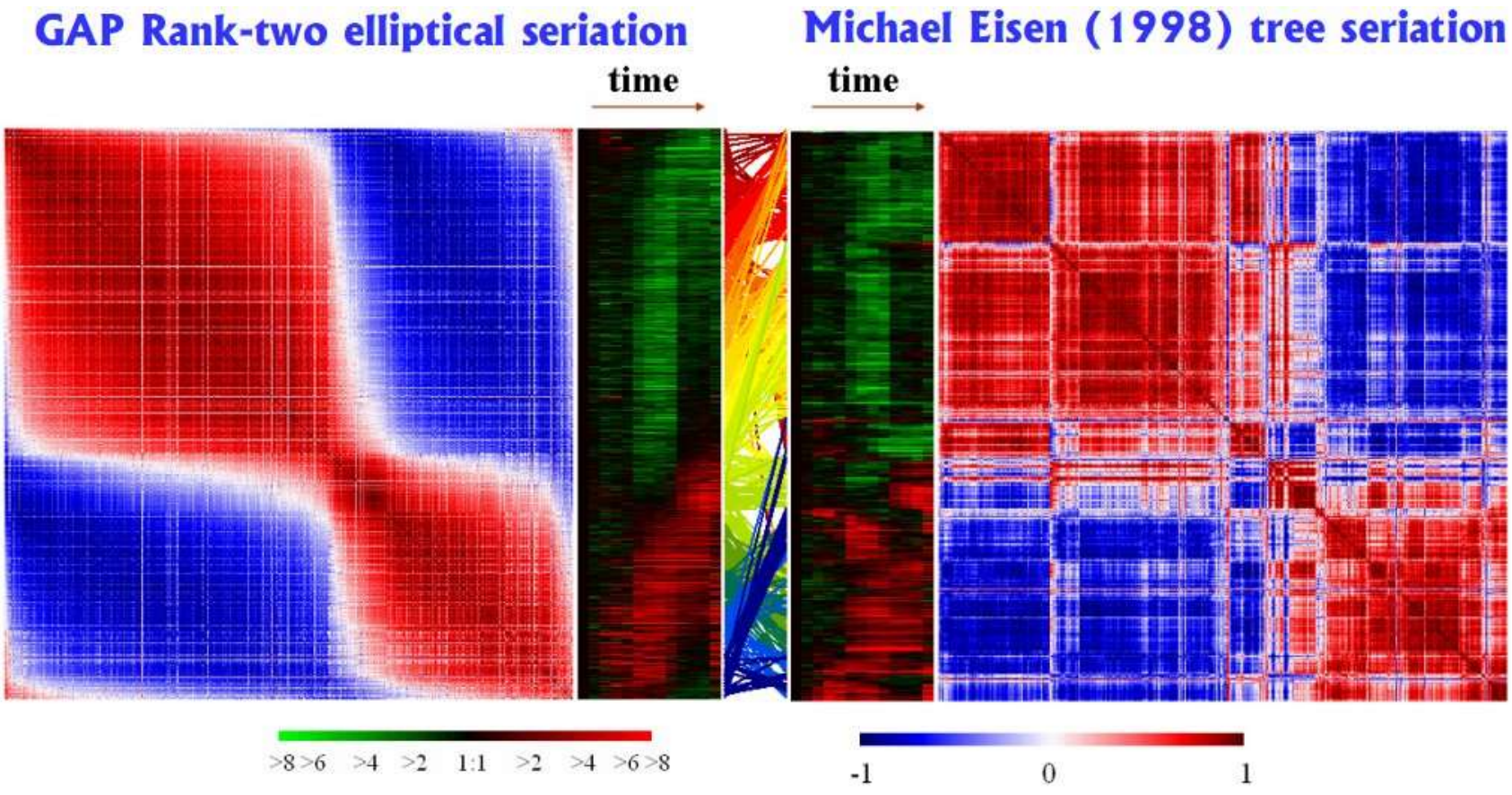
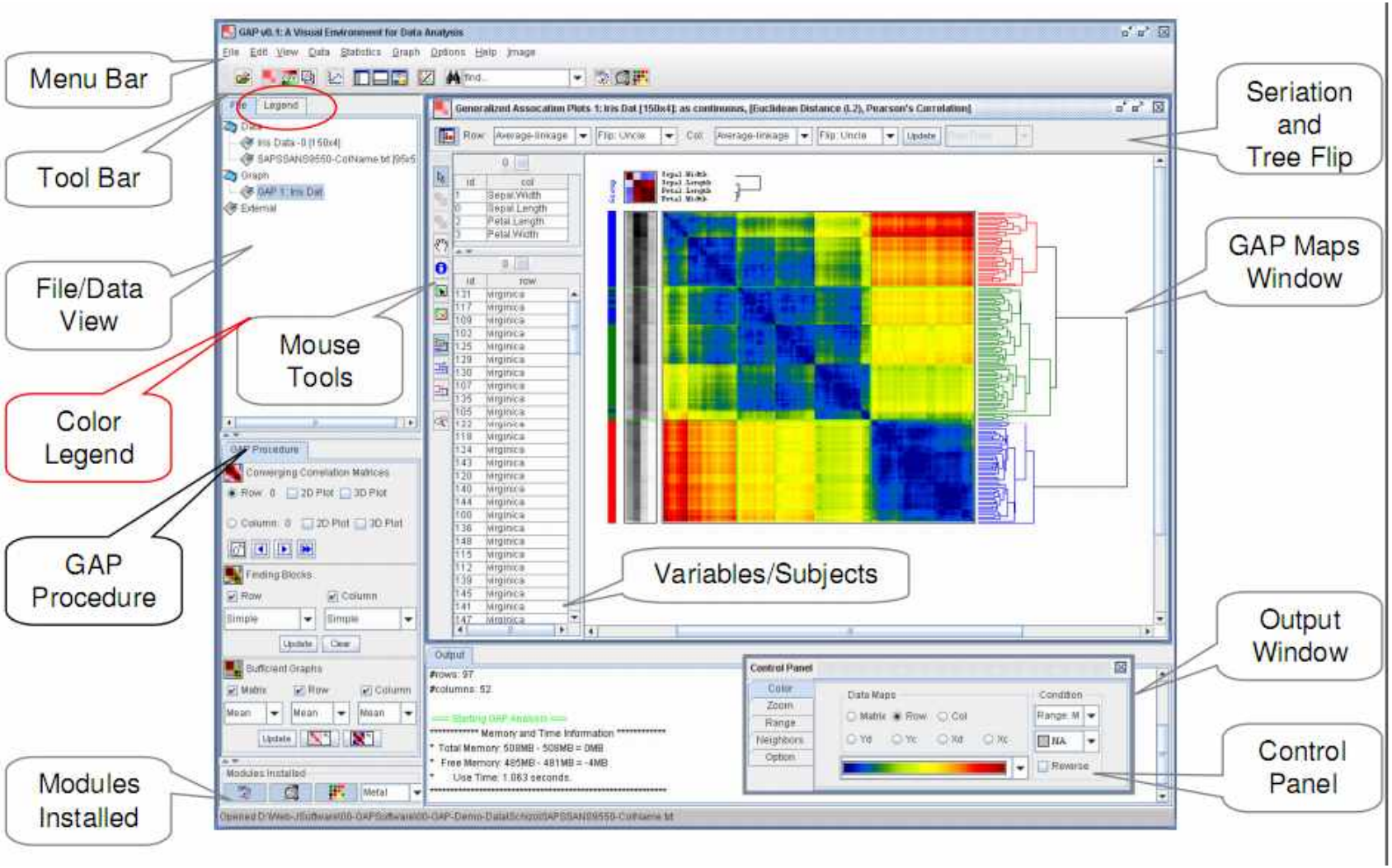


Image source: Dr. Chen Chun-houh's slide

Outline

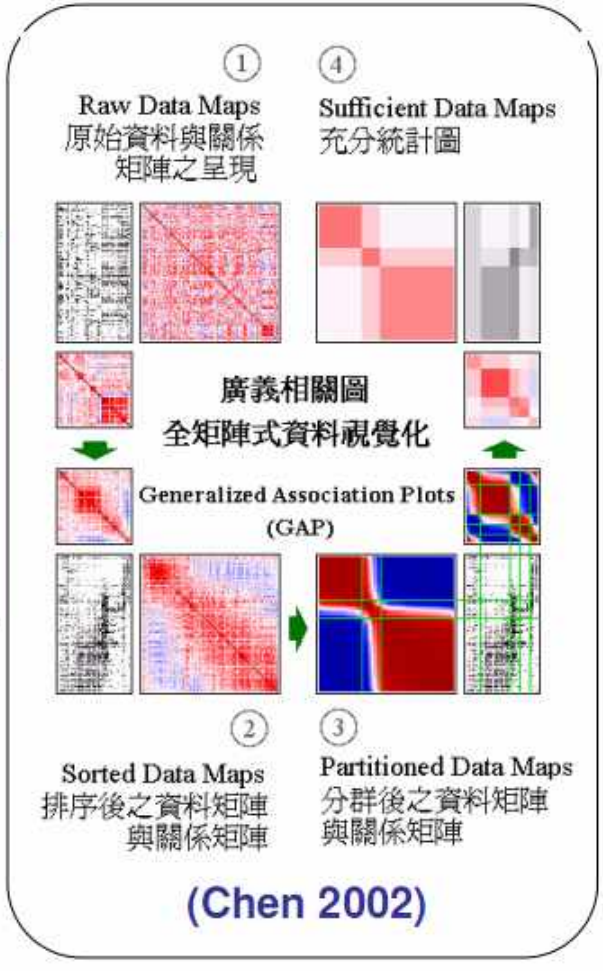
Main Window of Generalized Association Plots



Four Step of GAP

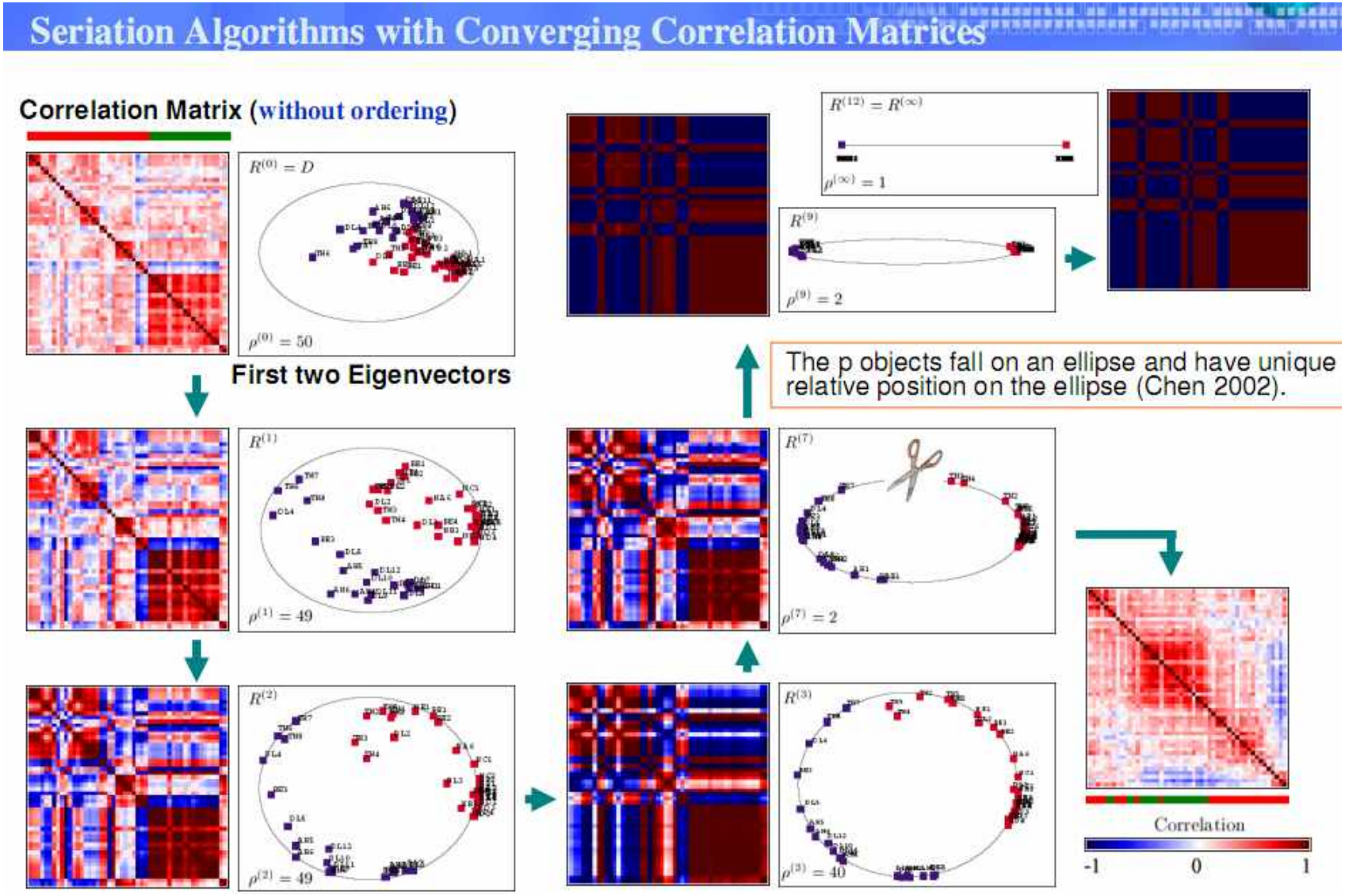
- Two Demo Datasets
- Four Steps of Generalized Association Plots (GAP)

- Generalization and Flexibility
- Modules/Software/Conclusion



NOTE: **Matrix Visualization (MV)**: reorderable matrix, the heatmap, color histogram, data image and matrix visualization.

Elliptical Seriation



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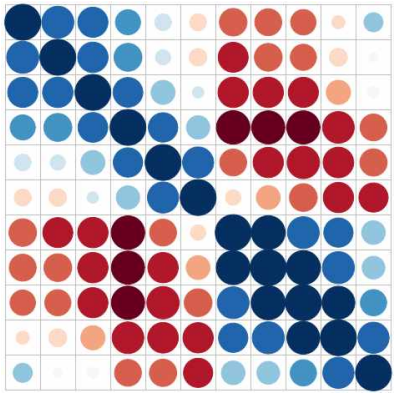
Best Wishes For You!

Thank You



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