

ORIGIN ≡ 1

CORRESPONDENCE ANALYSIS

Correspondence Analysis is primarily a graphical technique designed to show relationships between variables constituting rows and columns of a contingency table. For instance, in the example below, rows may represent different kinds of trees in a forest, columns may represent different forest localities, and the numbers represent counts of each tree observed in each locality. A typical result is a biplot graph showing points (here representing trees = rows) indicating relative similarity in the distribution of trees among localities. In addition often represented are vectors (here representing localities = columns) indicating relative similarity in occurrence of trees among localities. Close plotting of a vector (head) and point indicate greater association of tree with specific locality than expected at "random". As in PCA, a measure (here called "inertia") indicates relative quality of fit between the reduced dimensions of the plot versus the full dimensions of the problem. Data is from RA Johnson & DW Wichern *Applied Multivariate Statistical Analysis 4th Edition 1998*.

Read in Data:

Eight Tree species (rows) counted within plots of 10 localities (columns).

```
M := READPRN("c:\DATA\Multivariate\T12-10.DAT")
```

```
r := rows(M)      r = 8
```

```
c := cols(M)      c = 10
```

```
i := 1..r          j := 1..c
```

```
lri := 1          lcj := 1
```

$$T := \sum_i \sum_j M_{i,j} \quad T = 292 \quad \text{< Total number of observations in matrix M:}$$

	1	2	3	4	5	6	7	8	9	10
1	9	8	3	5	6	0	5	0	0	0
2	8	9	8	7	0	0	0	0	0	0
3	5	4	9	9	7	7	4	6	0	2
4	3	4	0	6	9	8	7	6	4	3
5	2	2	4	5	6	0	5	0	2	5
6	0	0	0	0	2	7	6	6	7	6
7	0	0	0	0	0	0	7	4	6	5
8	0	0	0	0	0	5	4	8	8	9

Matrix P of frequencies:

$$P := \frac{1}{T} M$$

^ Matrix P is called the "correspondence matrix"

$$P = \begin{pmatrix} 0.0308 & 0.0274 & 0.0103 & 0.0171 & 0.0205 & 0 & 0.0171 & 0 & 0 & 0 \\ 0.0274 & 0.0308 & 0.0274 & 0.024 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0.0171 & 0.0137 & 0.0308 & 0.0308 & 0.024 & 0.024 & 0.0137 & 0.0205 & 0 & 0.0068 \\ 0.0103 & 0.0137 & 0 & 0.0205 & 0.0308 & 0.0274 & 0.024 & 0.0205 & 0.0137 & 0.0103 \\ 0.0068 & 0.0068 & 0.0137 & 0.0171 & 0.0205 & 0 & 0.0171 & 0 & 0.0068 & 0.0171 \\ 0 & 0 & 0 & 0 & 0.0068 & 0.024 & 0.0205 & 0.0205 & 0.024 & 0.0205 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0.024 & 0.0137 & 0.0205 & 0.0171 \\ 0 & 0 & 0 & 0 & 0 & 0.0171 & 0.0137 & 0.0274 & 0.0274 & 0.0308 \end{pmatrix}$$

Matrix P_s of Centered & Scaled frequencies:

$$R := P \cdot l_c$$

Row (R) and Column (C) vectors of sums of matrix P:

$$C := P^T \cdot l_r$$

$$IsD_r := \text{diag} \left(\frac{1}{\sqrt{R}} \right)$$

< Inverse square-root matrices with values along the main diagonal

$$IsD_c := \text{diag} \left(\frac{1}{\sqrt{C}} \right)$$

$$R = \begin{pmatrix} 0.123 \\ 0.11 \\ 0.182 \\ 0.171 \\ 0.106 \\ 0.116 \\ 0.075 \\ 0.116 \end{pmatrix} \quad C = \begin{pmatrix} 0.09247 \\ 0.09247 \\ 0.08219 \\ 0.10959 \\ 0.10274 \\ 0.09247 \\ 0.13014 \\ 0.10274 \\ 0.09247 \\ 0.10274 \end{pmatrix}$$

$$P_s := \text{IsD}_r(P - R \cdot C^T) \cdot \text{IsD}_c \quad \text{< centered \& scaled correspondence matrix.}$$

$$P_s = \begin{pmatrix} 0.1819 & 0.1498 & 0.0014 & 0.0311 & 0.07 & -0.1068 & 0.0085 & -0.1125 & -0.1068 & -0.1125 \\ 0.1715 & 0.2055 & 0.1938 & 0.1092 & -0.1061 & -0.1007 & -0.1194 & -0.1061 & -0.1007 & -0.1061 \\ 0.0026 & -0.0238 & 0.1302 & 0.0775 & 0.039 & 0.0555 & -0.0646 & 0.0139 & -0.1295 & -0.0864 \\ -0.0442 & -0.017 & -0.1186 & 0.013 & 0.0997 & 0.0919 & 0.0113 & 0.0223 & -0.017 & -0.0552 \\ -0.0299 & -0.0299 & 0.0532 & 0.0509 & 0.0923 & -0.0991 & 0.0281 & -0.1044 & -0.0299 & 0.0595 \\ -0.1038 & -0.1038 & -0.0978 & -0.113 & -0.0468 & 0.1273 & 0.0438 & 0.0785 & 0.1273 & 0.0785 \\ -0.0835 & -0.0835 & -0.0787 & -0.0909 & -0.088 & -0.0835 & 0.1431 & 0.0677 & 0.1627 & 0.1066 \\ -0.1038 & -0.1038 & -0.0978 & -0.113 & -0.1094 & 0.0613 & -0.0118 & 0.1411 & 0.1603 & 0.1724 \end{pmatrix}$$

Singular Value Decomposition of matrix P_s :

$$\begin{aligned} \gamma_U &:= \text{reverse}(\text{sort}(\text{eigenvals}(P_s \cdot P_s^T))) \\ \gamma_V &:= \text{reverse}(\text{sort}(\text{eigenvals}(P_s^T \cdot P_s))) \\ U^{(j)} &:= \text{eigenvec}(P_s \cdot P_s^T, \gamma_{U_j}) \\ V^{(j)} &:= \text{eigenvec}(P_s^T \cdot P_s, \gamma_{V_j}) \\ \lambda &:= \sqrt{\gamma_U} \quad \text{< singular values} \end{aligned}$$

$$\gamma_U = \begin{pmatrix} 0.537 \\ 0.096 \\ 0.072 \\ 0.046 \\ 0.011 \\ 4.539 \times 10^{-3} \\ 3.882 \times 10^{-3} \\ 0 \end{pmatrix} \quad \gamma_V = \begin{pmatrix} 0.53673 \\ 0.09618 \\ 0.0721 \\ 0.04555 \\ 0.01107 \\ 0.00454 \\ 0.00388 \\ 0 \\ 0 \\ 0 \end{pmatrix} \quad \lambda = \begin{pmatrix} 0.7326 \\ 0.3101 \\ 0.2685 \\ 0.2134 \\ 0.1052 \\ 0.0674 \\ 0.0623 \\ 3.0039 \times 10^{-9} \end{pmatrix}$$

$$U = \begin{pmatrix} -0.39042 & 0.08311 & 0.47814 & -0.4562 & -0.03768 & -0.33691 & -0.40707 & 0.35112 \\ -0.53268 & 0.4985 & -0.40802 & -0.09251 & -0.07379 & 0.34196 & 0.24645 & 0.33104 \\ -0.19986 & -0.38891 & -0.40889 & 0.36216 & 0.43908 & -0.32169 & -0.18076 & 0.42604 \\ 0.06979 & -0.5382 & 0.17259 & -0.31809 & -0.05439 & 0.1596 & 0.61224 & 0.4138 \\ -0.08195 & 0.01512 & 0.42706 & 0.7086 & -0.41601 & 0.16851 & -0.03071 & 0.32583 \\ 0.40046 & -0.08307 & -0.14776 & -0.18661 & -0.00416 & 0.58951 & -0.55865 & 0.34123 \\ 0.36336 & 0.48503 & 0.32324 & 0.09371 & 0.62981 & -0.01639 & 0.21722 & 0.27449 \\ 0.46892 & 0.24757 & -0.315 & -0.07263 & -0.47713 & -0.51421 & 0.07627 & 0.34123 \end{pmatrix}$$

$$V = \begin{pmatrix} 0.38774 & -0.21076 & 0.06164 & -0.40292 & -0.0582 & -0.32693 & 0.42471 & 0.16413 & 0.16413 & 0.16417 \\ 0.38558 & -0.24276 & 0.01057 & -0.43452 & -0.19499 & 0.19683 & -0.26351 & 0.43824 & 0.43824 & 0.4382 \\ 0.34953 & -0.18206 & -0.40795 & 0.57182 & 0.23432 & 0.11673 & 0.32936 & 0.34774 & 0.34774 & 0.34774 \\ 0.30064 & 0.13553 & -0.05396 & 0.26465 & 0.00063 & 0.08255 & -0.66442 & 0.17881 & 0.17882 & 0.17884 \\ 0.11083 & 0.58167 & 0.48562 & 0.15977 & -0.23333 & -0.16067 & 0.07718 & 0.39287 & 0.39287 & 0.39288 \\ -0.20218 & 0.54005 & -0.46257 & -0.26868 & -0.09782 & 0.39427 & 0.26684 & 0.2543 & 0.25431 & 0.25431 \\ -0.18519 & -0.07562 & 0.50896 & 0.02909 & 0.60261 & 0.19554 & 0.15201 & 0.32748 & 0.32748 & 0.32747 \\ -0.31395 & 0.06439 & -0.33937 & -0.15671 & 0.33659 & -0.65731 & -0.25071 & 0.36671 & 0.36671 & 0.36671 \\ -0.42003 & -0.34836 & 0.0394 & -0.11651 & -0.06249 & 0.37722 & -0.14565 & 0.29052 & 0.29052 & 0.29056 \\ -0.35495 & -0.28969 & 0.0345 & 0.33932 & -0.59944 & -0.20021 & 0.12616 & 0.28805 & 0.28805 & 0.28804 \end{pmatrix}$$

Inertia:

$$\text{TI} := \sum_i \gamma U_i \quad \text{TI} = 0.77$$

$$\gamma U = \begin{pmatrix} 0.5367 \\ 0.0962 \\ 0.0721 \\ 0.0455 \\ 0.0111 \\ 0.0045 \\ 0.0039 \\ 0 \end{pmatrix} \quad \lambda = \begin{pmatrix} 0.7326 \\ 0.3101 \\ 0.2685 \\ 0.2134 \\ 0.1052 \\ 0.0674 \\ 0.0623 \\ 3.0039 \times 10^{-9} \end{pmatrix} \quad \frac{1}{\text{TI}} \cdot \lambda^2 = \begin{pmatrix} 0.697 \\ 0.1249 \\ 0.0936 \\ 0.0591 \\ 0.0144 \\ 0.0059 \\ 0.005 \\ 0 \end{pmatrix}$$

< Proportion of Inertia for each singular value squared $\lambda^2 = \gamma U$

Correspondence analysis coordinates:

$$0.697 + 0.1249 = 0.822$$

^ first two axes "account" for 82.2% of the fit

$$D_R := \text{diag}(R) \quad D_{S_R} := \text{diag}(\sqrt{R}) \quad \text{< diagonal R \& C matrices and diagonal square root matrices}$$

$$D_C := \text{diag}(C) \quad D_{S_C} := \text{diag}(\sqrt{C})$$

$$m := \text{rows}(P_S) \quad k := \text{cols}(P_S)$$

$$i := 1..k - m \quad k - m = 2 \quad Z_{m,i} := 0 \quad \text{< additional columns of zeros for use with matrix algebra}$$

$$\Lambda := \text{augment}(\text{diag}(\lambda), Z)$$

$$F := D_R^{-1} \cdot (D_{S_R} \cdot U) \cdot \Lambda$$

Matrix F represents vector positions for ROWS of M

Note: typically the first two (or three) columns of F and G are plotted together...

See plot below

$$F = \begin{pmatrix} -0.8146 & 0.0734 & 0.3656 & -0.2773 & -0.0113 & -0.0646 & -0.0722 & 3.0039 \times 10^{-9} & 0 & 0 \\ -1.1789 & 0.467 & -0.3309 & -0.0596 & -0.0234 & 0.0696 & 0.0464 & 3.0039 \times 10^{-9} & 0 & 0 \\ -0.3437 & -0.2831 & -0.2577 & 0.1814 & 0.1084 & -0.0509 & -0.0264 & 3.0039 \times 10^{-9} & 0 & 0 \\ 0.1236 & -0.4034 & 0.112 & -0.1641 & -0.0138 & 0.026 & 0.0922 & 3.0039 \times 10^{-9} & 0 & 0 \\ -0.1843 & 0.0144 & 0.3519 & 0.4641 & -0.1343 & 0.0348 & -0.0059 & 3.0039 \times 10^{-9} & 0 & 0 \\ 0.8598 & -0.0755 & -0.1163 & -0.1167 & -0.0013 & 0.1164 & -0.102 & 3.0039 \times 10^{-9} & 0 & 0 \\ 0.9698 & 0.548 & 0.3162 & 0.0729 & 0.2414 & -0.004 & 0.0493 & 3.0039 \times 10^{-9} & 0 & 0 \\ 1.0068 & 0.225 & -0.2479 & -0.0454 & -0.1471 & -0.1015 & 0.0139 & 3.0039 \times 10^{-9} & 0 & 0 \end{pmatrix}$$

$$G := D_C^{-1} \cdot (D_{S_C} \cdot V) \cdot \Lambda^T$$

Matrix G represents vector positions for COLUMNS of M

$$G = \begin{pmatrix} 0.9342 & -0.2149 & 0.0544 & -0.2828 & -0.0201 & -0.0724 & 0.087 & 1.6213 \times 10^{-9} \\ 0.929 & -0.2476 & 0.0093 & -0.305 & -0.0675 & 0.0436 & -0.054 & 4.3292 \times 10^{-9} \\ 0.8932 & -0.1969 & -0.3821 & 0.4257 & 0.086 & 0.0274 & 0.0716 & 3.6435 \times 10^{-9} \\ 0.6653 & 0.127 & -0.0438 & 0.1706 & 0.0002 & 0.0168 & -0.1251 & 1.6225 \times 10^{-9} \\ 0.2533 & 0.5628 & 0.4068 & 0.1064 & -0.0766 & -0.0338 & 0.015 & 3.6818 \times 10^{-9} \\ -0.4871 & 0.5508 & -0.4085 & -0.1886 & -0.0338 & 0.0874 & 0.0547 & 2.5121 \times 10^{-9} \\ -0.3761 & -0.065 & 0.3788 & 0.0172 & 0.1757 & 0.0365 & 0.0263 & 2.7269 \times 10^{-9} \\ -0.7176 & 0.0623 & -0.2843 & -0.1043 & 0.1105 & -0.1382 & -0.0487 & 3.4367 \times 10^{-9} \\ -1.012 & -0.3553 & 0.0348 & -0.0818 & -0.0216 & 0.0836 & -0.0298 & 2.8699 \times 10^{-9} \\ -0.8113 & -0.2803 & 0.0289 & 0.2259 & -0.1967 & -0.0421 & 0.0245 & 2.6995 \times 10^{-9} \end{pmatrix}$$

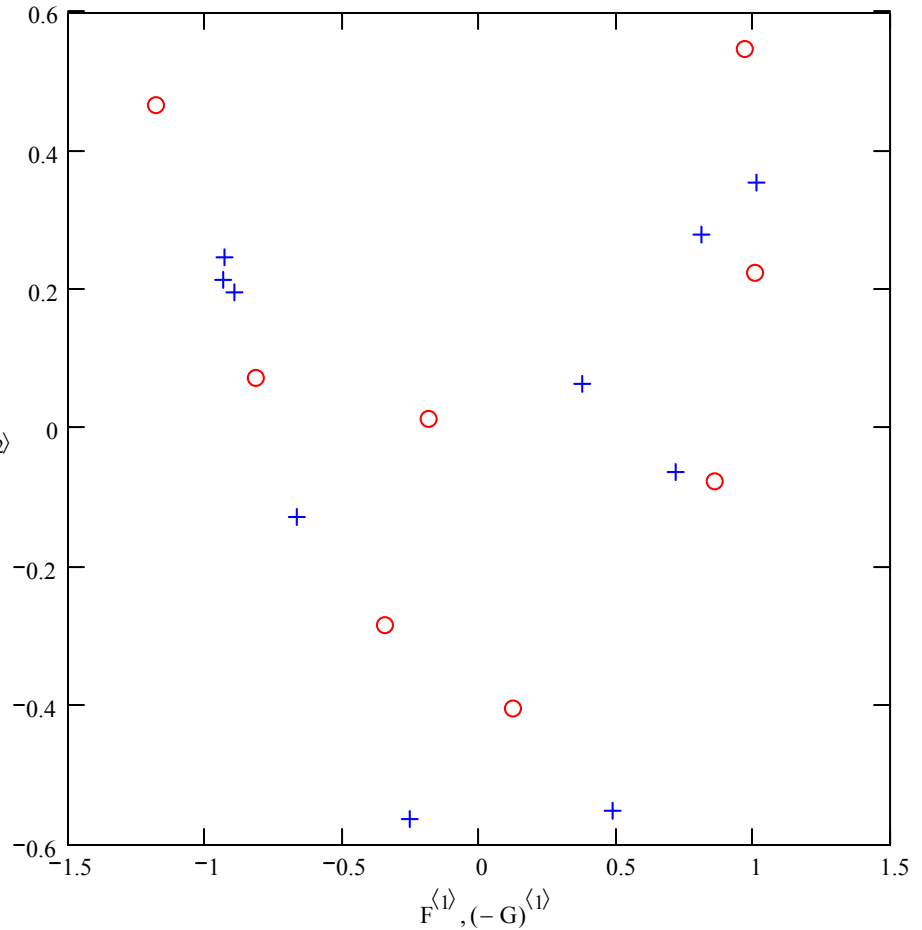
Biplot of Rows and Columns:

F (0) = ROWS of M = Trees
 G (+) = COLUMNS of M = Sites

Note: values of G are plotted as -G for comparison with R's output below.

As in PCA directions of eigenvectors are not unique.

F⁽²⁾
 (-G)⁽²⁾
 + +



Prototype in R:

```
#CORRESPONDENCE ANALYSIS
M=read.table("c:/MultivariateDATA/T1 2-10.DAT")
M
attach(M)
P=prop.table(M) # CREATES CORRESPONDENCE MATRIX P
P
```

```
> P
      v1      v2      v3      v4      v5      v6
1 0.030821918 0.027397260 0.01027397 0.01712329 0.020547945 0.000000000
2 0.027397260 0.030821918 0.02739726 0.02397260 0.000000000 0.000000000
3 0.017123288 0.013698630 0.03082192 0.03082192 0.023972603 0.02397260
4 0.010273973 0.013698630 0.00000000 0.02054795 0.030821918 0.02739726
5 0.006849315 0.006849315 0.01369863 0.01712329 0.020547945 0.000000000
6 0.000000000 0.000000000 0.00000000 0.00000000 0.006849315 0.02397260
7 0.000000000 0.000000000 0.00000000 0.00000000 0.000000000 0.000000000
8 0.000000000 0.000000000 0.00000000 0.00000000 0.000000000 0.01712329
      v7      v8      v9      v10
1 0.01712329 0.00000000 0.00000000 0.000000000
2 0.00000000 0.00000000 0.00000000 0.000000000
3 0.01369863 0.02054795 0.00000000 0.006849315
4 0.02397260 0.02054795 0.01369863 0.010273973
5 0.01712329 0.00000000 0.006849315 0.017123288
6 0.02054795 0.02054795 0.023972603 0.020547945
7 0.02397260 0.01369863 0.020547945 0.017123288
8 0.01369863 0.02739726 0.027397260 0.030821918
```

^ correspondence matrix P

```
library(ca) # INSTALL LIBRARY {ca} from CRAN
?ca
CA <- ca(P)
summary(CA)
```

```
> summary(CA)
```

Principal inertias (eigenvalues):

inertias >

dim	value	%	cum%	scree plot
1	0.536732	69.7	69.7	*****
2	0.096182	12.5	82.2	****
3	0.072097	9.4	91.6	***
4	0.045547	5.9	97.5	**
5	0.011066	1.4	98.9	
6	0.004539	0.6	99.5	
7	0.003882	0.5	100.0	

Total: 0.770046 100.0

mass = row and column sums of correspondence matrix P X 1000

Rows:

	name	mass	qlt	inr	k=1	cor	ctr	k=2	cor	ctr
1	1	123	752	142	-815	746	152	73	6	7
2	2	110	930	246	-1179	804	284	467	126	249
3	3	182	634	74	-344	378	40	-283	256	151
4	4	171	785	50	124	67	5	-403	717	290
5	5	106	87	54	-184	86	7	14	1	0
6	6	116	936	120	860	929	160	-75	7	7
7	7	75	882	138	970	669	132	548	213	235
8	8	116	918	175	1007	874	220	225	44	61

Columns:

	name	mass	qlt	inr	k=1	cor	ctr	k=2	cor	ctr
1	V1	92	905	122	-934	860	150	215	46	44
2	V2	92	900	123	-929	841	149	248	60	59
3	V3	82	711	126	-893	678	122	197	33	33
4	V4	110	907	72	-665	875	90	-127	32	18
5	V5	103	674	75	-253	114	12	-563	561	338
6	V6	92	716	91	487	314	41	-551	402	292
7	V7	130	452	54	376	439	34	65	13	6
8	V8	103	805	86	718	799	99	-62	6	4
9	V9	92	986	140	1012	878	176	355	108	121
10	V10	103	888	111	811	793	126	280	95	84

Other values in this table have not yet been identified.

```
?plot.ca
plot(CA, arrows = c(FALSE, TRUE))
```

