TOPIC 9 UNCONSTRAINED ORDINATION: CORRESPONDENCE ANALYSIS

- see <u>https://sites.google.com/site/mb3gustame/indirect-gradient-analysis/ca</u> and <u>https://www.davidzeleny.net/anadat-r/doku.php/en:ca_dca</u>
- also an eigenvector method
- handles nonlinear species responses better than PCA
 - therefore better for community data
- CA is based on similar regression techniques as PCA, but with χ^2 -standardised data and weights
 - (... i.e. PCA uses an intermediate correlation matrix)
- then subjected to either a Singular Value Decomposition (SVD) or eigenvalue decomposition, and the eigenvalues and eigenvectors reported
- the ordination preserves χ^{2-} (D₁₆) rather than Euclidian (D₁) distance between sites
 - χ^2 -distance is not influenced by double Os
 - no pre-transformation needed
 - suitable for species counts and presence/absence data
- CA maximises the correspondence between species scores and sample scores, whereas a PCA maximises the variance explained

- CA produces one axis fewer than min[n, p]
- as with PCA, orthogonal axes ranked in **decreasing order of importance**
- the variation represented is the total inertia, which is the SS of all the values in the χ^2 matrix
 - (... i.e. not the sum of the eigenvalues along the diagonal as in a PCA)
- individual eigenvalues will always be <1
- the variation represented along an axis is given by dividing the eigenvalues of the axis by the total inertia

- CA approximates a **unimodal** response model
 - i.e. matches gradients better (fits better to env. data)
 - the species scores give the species maximum and the abundance decreases in every direction from the centroid of the species score
 - (... in PCA species close to the origin (zero) change little and is poorly presented by the ordination, but in CA it may have its optimum there)
- the horseshoe effect is weaker, but still not entirely gone

- scaling of ordination plots
 - **scaling 1**—site scaling (rows the centroids of columns)
 - i.e. sites that plot close together are similar i.t.o. their species relative frequencies
 - any site near a point representing a species will have a relatively large contribution by that species
 - **scaling 2**—species scaling (columns are the centroids of rows)
 - i.e. species points that are close together will have relatively similar frequencies among the objects (sites)
 - any species plotting close-by a point that represents a site is more likely to be found at that object, or have a higher frequency there than at sites appearing further away in the ordination plot

> :	spe																			
# A tibble: 29 x 27																				
	Cogo	Satr	Phph	Babl	Thth	Teso	Chna	Pato	Lele	Sqce	Baba	Albi	Gogo	Eslu	Pefl	Rham	Legi	Scer	Суса	Titi
	<int></int>	<int> <</int>	<int></int>																	
1	0	3	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
2	0	5	4	3	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
3	0	5	5	5	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0
4	0	4	5	5	0	0	0	0	0	1	0	0	1	2	2	0	0	0	0	1
5	0	2	3	2	0	0	0	0	5	2	0	0	2	4	4	0	0	2	0	3
6	0	3	4	5	0	0	0	0	1	2	0	0	1	1	1	0	0	0	0	2
7	0	5	4	5	0	0	0	0	1	1	0	0	0	0	0	0	0	0	0	0
8	0	0	1	3	0	0	0	0	0	5	0	0	0	0	0	0	0	0	0	1
9	0	1	4	4	0	0	0	0	2	2	0	0	1	0	0	0	0	0	0	0
10	1	3	4	1	1	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0
#	wit	:h 19 m	nore ro	ows, ar	nd 7 ma	ore vai	riables	: Abbr	<int></int>	, Icme	<int></int>	, Gyce	<int>,</int>	Ruru	<int></int>	, Blbj	<int></int>	, Alal	<int></int>	1
#	Anan	<int></int>																		

> (spe.ca <- cca(spe))
Call: cca(X = spe)</pre>

Inertia Rank Total 1.167 Unconstrained 1.167 26 Inertia is scaled Chi-square

Eigenvalues for unconstrained axes:

CA1 CA2 CA3 CA4 CA5 CA6 CA7 CA8 0.6010 0.1444 0.1073 0.0834 0.0516 0.0418 0.0339 0.0288 (Showed only 8 of all 26 unconstrained eigenvalues)

<pre>> summary(spe.ca)</pre>	# default scaling 2											
Call: cca(X = spe)												
Partitioning of scaled Chi-square:												
Inertia	Proportion											
Unconstrained 1.16/	1											
Eigenvalues, and their contribution to the scaled Chi-square												
Importance of component	nts:											
	CA1 CA2 CA3 CA4 CA5 CA6 CA7 CA8 CA9 CA10 CA11 CA12											
Eigenvalue	0.601 0.1444 0.10729 0.08337 0.05158 0.04185 0.03389 0.02883 0.01684 0.010826 0.010142 0.007886											
Proportion Explained	0.515 0.1237 0.09195 0.07145 0.04420 0.03586 0.02904 0.02470 0.01443 0.009278 0.008691 0.006758											
Cumulative Proportion	0.515 0.6387 0.73069 0.80214 0.84634 0.88220 0.91124 0.93594 0.95038 0.959655 0.968346 0.975104											
	CA13 CA14 CA15 CA16 CA17 CA18 CA19 CA20 CA21 CA22 CA23											
Eigenvalue	0.006123 0.004867 0.004606 0.003844 0.003067 0.001823 0.001642 0.001295 0.0008775 0.0004217 0.0002149											
Proportion Explained	0.005247 0.004171 0.003948 0.003294 0.002629 0.001562 0.001407 0.001110 0.0007520 0.0003614 0.0001841											
Cumulative Proportion	0.980351 0.984522 0.988470 0.991764 0.994393 0.995955 0.997362 0.998472 0.9992238 0.9995852 0.9997693											
	CA24 CA25 CA26											
Eigenvalue	0.0001528 8.949e-05 2.695e-05											
Proportion Explained 0.0001309 7.669e-05 2.310e-05												
Cumulative Proportion 0.9999002 1.000e+00 1.000e+00												
Scaling 2 for species * Species are scaled * Sites are unscaled:	and site scores proportional to eigenvalues weighted dispersion equal on all dimensions											

...continue

Species scores

Cogo 1.50075 -1.410293 0.26049 -0.307203 0.271777 -0. Satr 1.66167 0.444129 0.57571 0.166073 -0.261870 -0.	003465 326590 200732
Satr 1.66167 0.444129 0.57571 0.166073 -0.261870 -0.	326590 200732
	200732
Phph 1.28545 0.285328 -0.04768 0.018126 0.043847 0.	
Babl 0.98662 0.360900 -0.35265 -0.009021 -0.012231 0.	253429
Thth 1.55554 -1.389752 0.80505 -0.468471 0.471301 0.	225409
Teso 0.99709 -1.479902 -0.48035 0.079397 -0.105715 -0.	332445
Chna -0.54916 -0.051534 0.01123 -0.096004 -0.382763 0.	134807
Pato -0.18478 -0.437710 -0.57438 0.424267 -0.587150 0.	091866
Lele 0.01337 -0.095342 -0.57672 0.212017 0.126668 -0.	389103
Sqce 0.01078 0.140577 -0.34811 -0.538268 0.185286 0.	167087
Baba -0.33363 -0.300682 -0.04929 0.170961 -0.157203 0.	103068
Albi -0.38357 -0.255310 -0.20136 0.374057 -0.385866 0.	239001
Gogo -0.32152 -0.034382 -0.07423 -0.031236 0.014417 -0.	156351
Eslu -0.26165 0.187282 0.00617 0.183771 0.295142 -0.	262808
Pefl -0.28913 0.121044 -0.18919 0.367615 0.218087 -0.	163675
Rham -0.60298 -0.057369 0.20341 0.214299 -0.050977 0.	211926
Legi -0.58669 -0.082467 0.21198 0.050175 -0.120456 0.	108724
Scer -0.61815 0.124733 0.13339 0.147190 0.317736 -0.	340380
Cyca -0.57951 -0.110732 0.20173 0.308547 0.006854 0.	153224
Titi -0.37880 0.138023 -0.07825 0.095793 0.256285 -0.	029245
Abbr -0.70235 0.011155 0.40242 0.211582 0.138186 0.	132297
Icme -0.73238 -0.009098 0.55678 0.321852 0.281812 0.	172271
Gyce -0.69300 0.038971 0.37688 -0.183965 -0.051945 -0.	011126
Ruru -0.44181 0.176915 -0.23691 -0.345104 0.129676 -0.	043802
Blbj -0.70928 0.032317 0.40924 0.030224 0.049050 0.	114560
Alal -0.63114 0.053594 0.15204 -0.661381 -0.414796 -0.	206611
Anan -0.63578 -0.041894 0.30093 0.224044 0.030444 0.	203160

continue...

...continue

Site	scores	(weig	ghted	aver	ages	of sp	peci	es sco	ores)		
	C	CA1	С	A2		CA3		CA4	CA5	CA6	
sit1	2.764	88 3	3.0763	06	5.365	57489	1.	99192	-5.07714	-7.80447	
sit2	2.275	540 2	2.5655	31	1.265	59130	0.	87538	-1.89139	-0.13887	
sit3	2.018	323 2	2.4412	24	0.514	4079	0.	79436	-1.03741	0.56015	
sit4	1.284	85 1	1.9356	64 -	0.250	9482	0.	76470	0.54752	0.10579	
sit5	0.088	375 1	1.0151	82 -	1.455	5434	0.	47672	2.69249	-2.92498	
sit6	1.031	.88 1	1.7121	63 –	0.954	4059	0.	01584	0.91932	0.39856	
sit7	1.914	27 2	2.2562	08 -	0.000	1407	0.	39844	-1.07017	0.32127	
sit8	0.255	5 91 1	1.4430	08 -	2.577	7721	-3.	41400	2.36613	2.71741	
sit9	1.245	517 1	1.5263	91 -	1.963	35663	-0.	41230	0.69647	1.51859	
sit10	2.145	601 0	0.1102	78	1.610	8693	-0.	82023	0.53918	1.01153	
sit11	2.174	18 -0	0.2516	49	1.584	5397	-0.	81483	0.52623	1.05501	
sit12	2.309	944 -2	2.0344	39	1.918	31448	-0.	60481	0.64435	-0.14844	
sit13	1.871	41 -2	2.2625	03	1.106	6796	-0.	80840	1.09542	0.11038	
sit14	1.346	59 -1	1.8059	67 –	0.644	1505	-0.	52803	0.76871	-0.67165	
sit15	0.702	214 -1	1.5011	67 –	1.973	35888	0.	98502	-0.93585	-1.27168	
sit16	0.287	75 -0	0.8368	03 -	1.225	59108	0.	73302	-1.57036	0.57315	
sit17	0.052	299 -6	0.6479	50 -	0.923	34228	0.	35770	-0.95401	0.77738	
sit18	8 -0.205	584 -6	0.0072	52 -	1.015	54343	0.	07041	-1.03450	0.51442	
sit19	-0.578	879 (0.0428	49 -	0.366	0551	-0.	15019	-0.61357	0.10115	
sit20	0.673	820 6	0.0388	75	0.119	4956	0.	17256	-0.14686	-0.12018	
sit21	-0.719	933 (0.0146	94	0.220	4186	0.	13598	0.09459	-0.02068	
sit22	-0.704	38 (0.7353	98 -	0.654	6250	-6.	61523	-2.49441	-1.73215	
sit23	9 -0.839	76 (0.3901	20	0.560	5295	-4.	38864	-2.56916	-0.96702	
sit24	-0.684	76 (0.4188	42 -	0.286	60819	-2.	80336	-0.37540	-3.93791	
sit25	-0.758	808 6	0.2102	04	0.589	4091	-0.	70004	-0.01880	-0.10779	
sit26	-0.750)46 (0.1008	69	0.553	31191	-0.	12946	0.29164	0.11280	
sit27	/ -0.778	878 (0.0889	76	0.737	9012	0.	05204	0.40940	0.43236	
sit28	8 -0.608	815 -0	0.2032	35	0.552	2726	0.	43621	0.15010	0.25618	
sit29	-0.808	860 -0	0.0195	92	0.668	36542	0.	88136	0.52744	0.16456	



site scaling

- most interested in sites
- sites that plot close together are similar i.t.o. their species relative frequencies
- any site near a point representing a species will have a relatively large contribution by that species
- species scaling
- most interested in species
- species points that are close together will have relatively similar frequencies among the objects (sites)

sit1

sit3

sit10

sit11

sit12

3

sit13

2

sit7

Satr

sit14

1

sit4

sit9

· species plotting close-by a point that represents an object (site) are more likely to be found at that object, or have a higher frequency there than at objects appearing further away in the ordination plot

Fit and Plot Smooth Surfaces of Variables on Ordination

```
require('viridis')
palette(viridis(8))
par(mar = c(4, 4, 0.9, 0.5) + .1, mfrow = c(2, 2))
with(spe, tmp <- ordisurf(spe.ca ~ Satr, bubble = 3,</pre>
                           family = quasipoisson, knots = 2, col = 6,
                           display = "sites", main = "Satr"))
abline(h = 0, v = 0, lty = 3)
with(spe, tmp <- ordisurf(spe.ca ~ Scer, bubble = 3,</pre>
                           family = quasipoisson, knots = 2, col = 6,
                           display = "sites", main = "Scer"))
abline(h = 0, v = 0, lty = 3)
with(spe, tmp <- ordisurf(spe.ca ~ Teso, bubble = 3,</pre>
                           family = quasipoisson, knots = 2, col = 6,
                          display = "sites", main = "Teso"))
abline(h = 0, v = 0, lty = 3)
with(spe, tmp <- ordisurf(spe.ca ~ Cogo, bubble = 3,</pre>
                           family = quasipoisson, knots = 2, col = 6,
                           display = "sites", main = "Cogo"))
abline(h = 0, v = 0, lty = 3)
# A posteriori projection of environmental variables in a CA
# The last plot produced (CA scaling 2) must be active
(spe.ca.env <- envfit(spe.ca, env, scaling = 2)) # Scaling 2 is default</pre>
plot(spe.ca.env)
# Plot significant variables with a different colour
plot(spe.ca.env, p.max = 0.05, col = "red")
```



