

ACC Analysis

Project Effect Fire Dung Beetle

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```
ACCDB <- read.csv("DataACC.csv", header = T, sep = ";")
ID <- ACCDB[1]
```

Data Frame Summary

```
dfSummary(ACCDB, plain.ascii = FALSE, style = "grid", graph.magnif = 0.75,
  valid.col = FALSE, tmp.img.dir = "/tmp")
```

temporary images written to 'C:\tmp'

```
### Data Frame Summary
#### ACCDB
**Dimensions:** 15 x 18
**Duplicates:** 0
```

No	Variable	Stats / Values	Freqs (% of Valid)	Graph	Missing
1	ID\	1\. 13\	1 (6.7%)\	! [] (/tmp/ds0399.png)	0\
	[character]	2\. 14\	1 (6.7%)\		(0.0%)
		3\. 18\	1 (6.7%)\		
		4\. 2\	1 (6.7%)\		
		5\. 3\	1 (6.7%)\		
		6\. 6\	1 (6.7%)\		
		7\. 9\	1 (6.7%)\		
		8\. JTCP1\	1 (6.7%)\		
		9\. S1.3b\	1 (6.7%)\		
		10\. S3\	1 (6.7%)\		
		[5 others]	5 (33.3%)		
2	Cover\	1\. Biodiverse pasture\	5 (33.3%)\	! [] (/tmp/ds0400.png)	0\
	[character]	2\. Grass Monoculture\	5 (33.3%)\		(0.0%)
		3\. Secondary Forest	5 (33.3%)		
3	FireRegime\	1\. High\	5 (33.3%)\	! [] (/tmp/ds0401.png)	0\
	[character]	2\. Low\	5 (33.3%)\		(0.0%)
		3\. Null	5 (33.3%)		

4	Abundance\ [integer]	Mean (sd) : 71 (48)\ min < med < max:\ 13 < 70 < 179\ IQR (CV) : 73.5 (0.7)	15 distinct values	![] (/tmp/ds0402.png)	0\ (0.0%)
5	Richness\ [integer]	Mean (sd) : 7 (1.8)\ min < med < max:\ 4 < 7 < 10\ IQR (CV) : 2 (0.3)	4 : 1 (6.7%)\ 5 : 2 (13.3%)\ 6 : 4 (26.7%)\ 7 : 2 (13.3%)\ 8 : 3 (20.0%)\ 9 : 1 (6.7%)\ 10 : 2 (13.3%)	![] (/tmp/ds0403.png)	0\ (0.0%)
6	q0\ [integer]	Mean (sd) : 7.1 (1.8)\ min < med < max:\ 4 < 7 < 10\ IQR (CV) : 2 (0.3)	4 : 1 (6.7%)\ 5 : 2 (13.3%)\ 6 : 4 (26.7%)\ 7 : 1 (6.7%)\ 8 : 4 (26.7%)\ 9 : 1 (6.7%)\ 10 : 2 (13.3%)	![] (/tmp/ds0404.png)	0\ (0.0%)
7	q1\ [numeric]	Mean (sd) : 4.2 (1.2)\ min < med < max:\ 2.4 < 4.1 < 6\ IQR (CV) : 2 (0.3)	14 distinct values	![] (/tmp/ds0405.png)	0\ (0.0%)
8	q2\ [numeric]	Mean (sd) : 3.3 (1.2)\ min < med < max:\ 1.7 < 3 < 5.1\ IQR (CV) : 1.9 (0.3)	14 distinct values	![] (/tmp/ds0406.png)	0\ (0.0%)
9	pH\ [numeric]	Mean (sd) : 6 (0.3)\ min < med < max:\ 5.5 < 5.9 < 6.7\ IQR (CV) : 0.3 (0)	14 distinct values	![] (/tmp/ds0407.png)	0\ (0.0%)
10	Humidity\ [numeric]	Mean (sd) : 29.9 (6.8)\ min < med < max:\ 20.8 < 28.7 < 43.3\ IQR (CV) : 8 (0.2)	15 distinct values	![] (/tmp/ds0408.png)	0\ (0.0%)
11	DA\ [numeric]	Mean (sd) : 2 (0.1)\ min < med < max:\ 1.8 < 2.1 < 2.2\ IQR (CV) : 0.2 (0.1)	13 distinct values	![] (/tmp/ds0409.png)	0\ (0.0%)
12	N\ [numeric]	Mean (sd) : 0.2 (0.1)\ min < med < max:\ 0 < 0.2 < 0.5\ IQR (CV) : 0.1 (0.7)	14 distinct values	![] (/tmp/ds0410.png)	0\ (0.0%)
13	C\ [numeric]	Mean (sd) : 2.8 (1.4)\ min < med < max:\	15 distinct values	![] (/tmp/ds0411.png)	0\ (0.0%)

		1.2 < 2.5 < 5.9\				
		IQR (CV) : 1.7 (0.5)				
14	P_disp\	Mean (sd) : 6.1 (9.5)\	14 distinct values	![] (/tmp/ds0412.png)	0\	
	[numeric]	min < med < max:\			(0.0%)	
		0.7 < 2.8 < 34.8\				
		IQR (CV) : 2.2 (1.5)				
15	P_total\	Mean (sd) : 306.9 (79.8)\	15 distinct values	![] (/tmp/ds0413.png)	0\	
	[numeric]	min < med < max:\			(0.0%)	
		156.7 < 304.4 < 447.9\				
		IQR (CV) : 95.3 (0.3)				
16	Arena\	Mean (sd) : 37.7 (8.8)\	12 distinct values	![] (/tmp/ds0414.png)	0\	
	[numeric]	min < med < max:\			(0.0%)	
		24.8 < 36.8 < 52.8\				
		IQR (CV) : 10.5 (0.2)				
17	Arcilla\	Mean (sd) : 44.5 (8.9)\	12 distinct values	![] (/tmp/ds0415.png)	0\	
	[numeric]	min < med < max:\			(0.0%)	
		30.2 < 45.2 < 61.2\				
		IQR (CV) : 14.5 (0.2)				
18	Limo\	Mean (sd) : 17.8 (3.8)\	13 : 1 (6.7%\	![] (/tmp/ds0416.png)	0\	
	[integer]	min < med < max:\	14 : 4 (26.7%\		(0.0%)	
		13 < 17 < 24\	15 : 1 (6.7%\			
		IQR (CV) : 6.5 (0.2)	17 : 2 (13.3%\			
			19 : 2 (13.3%\			
			20 : 1 (6.7%\			
			21 : 1 (6.7%\			
			22 : 1 (6.7%\			
			24 : 2 (13.3%\			

Analysis ACC

Obtenemos los valores de las correlaciones entre variables

Conocemos las variables de coprofagos a relacionar

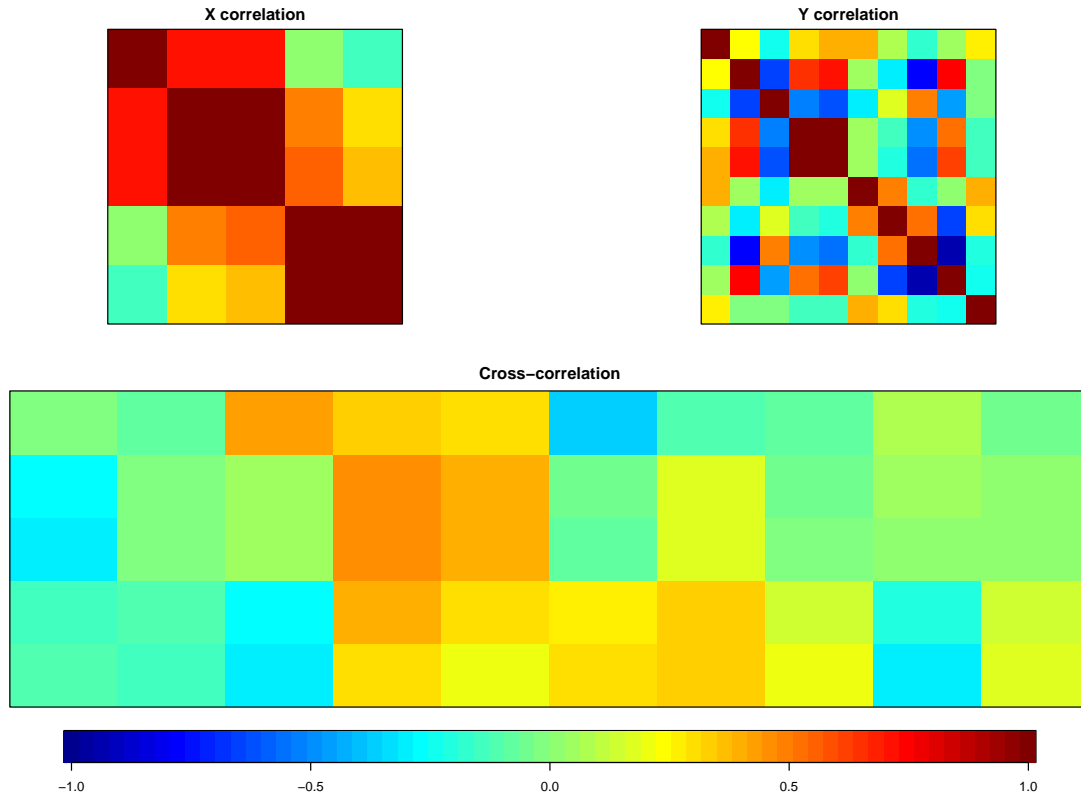
```
names(ACCDB[, 4:8])
```

```
[1] "Abundance" "Richness"  "q0"         "q1"         "q2"
```

Conocemos las variables de suelo a relacionar

```
names(ACCDB[, 9:18])
```

```
[1] "pH"          "Humidity" "DA"        "N"          "C"          "P_disp"
[7] "P_total"    "Arena"    "Arcilla"   "Limo"
```



Calculamos las correlaciones canónicas.

```
cc1 <- cc(dungbeetles, soil)
cc1$cor
```

```
[1] 0.9991714 0.9177246 0.7935859 0.6399311 0.3757272
```

Calculamos los vectores canónicos

```
cc1[3:4]
```

\$xcoef

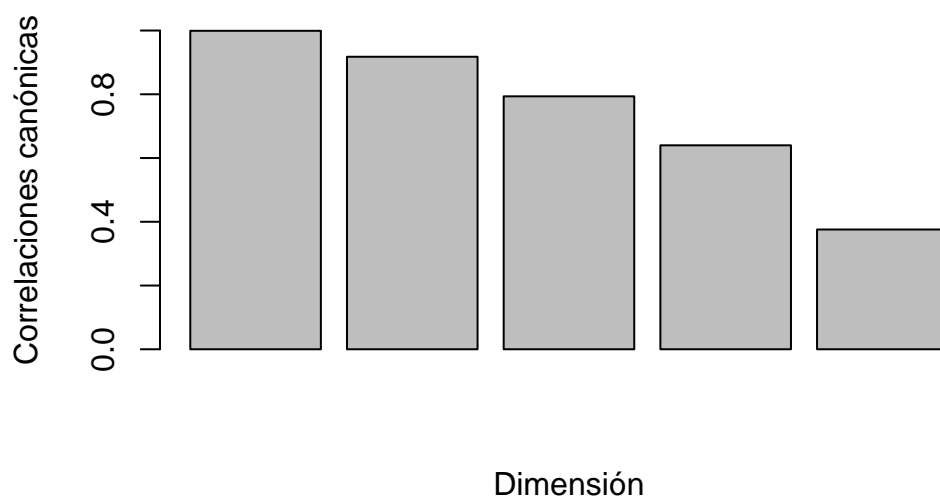
	[,1]	[,2]	[,3]	[,4]	[,5]
Abundance	0.006272365	-0.02566235	-0.01636723	-0.0211524	0.003455225
Richness	0.699480607	0.37125319	-1.19936491	-1.0290469	-3.949423043
q0	-0.053908847	0.44016596	3.17233910	1.1690544	3.745015457
q1	-2.272684000	-2.91841159	-9.47006677	4.0408423	0.477779784
q2	2.642174667	2.99564029	8.27438012	-4.7142311	-0.458064794

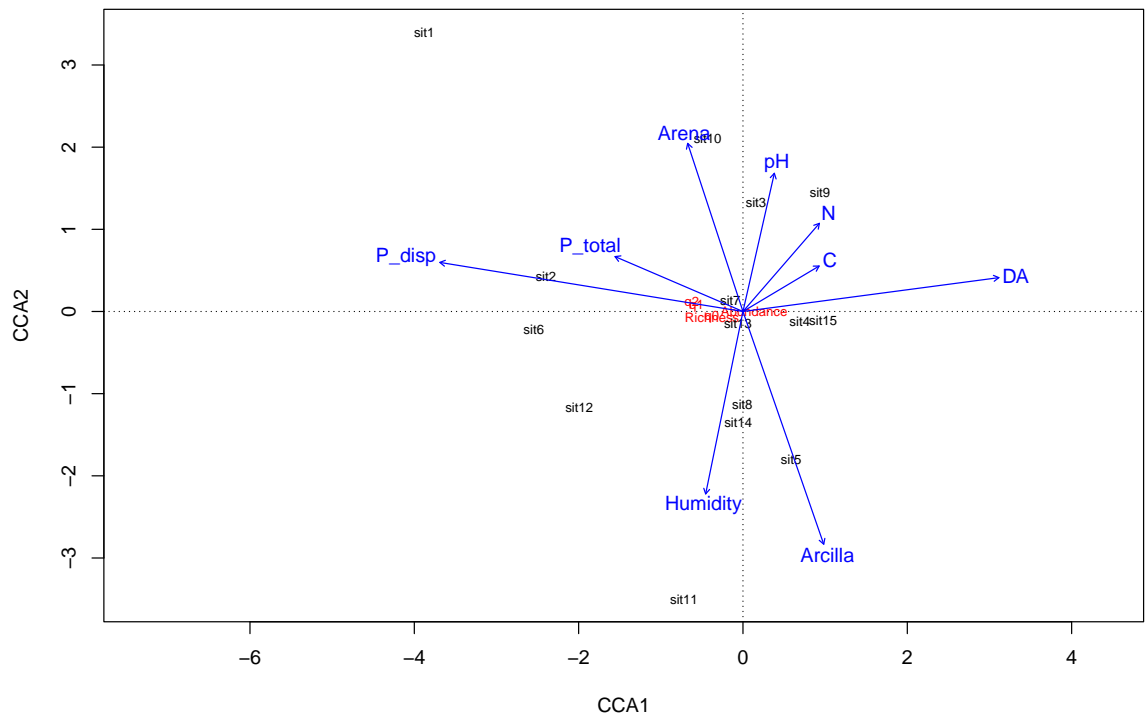
\$ycoef

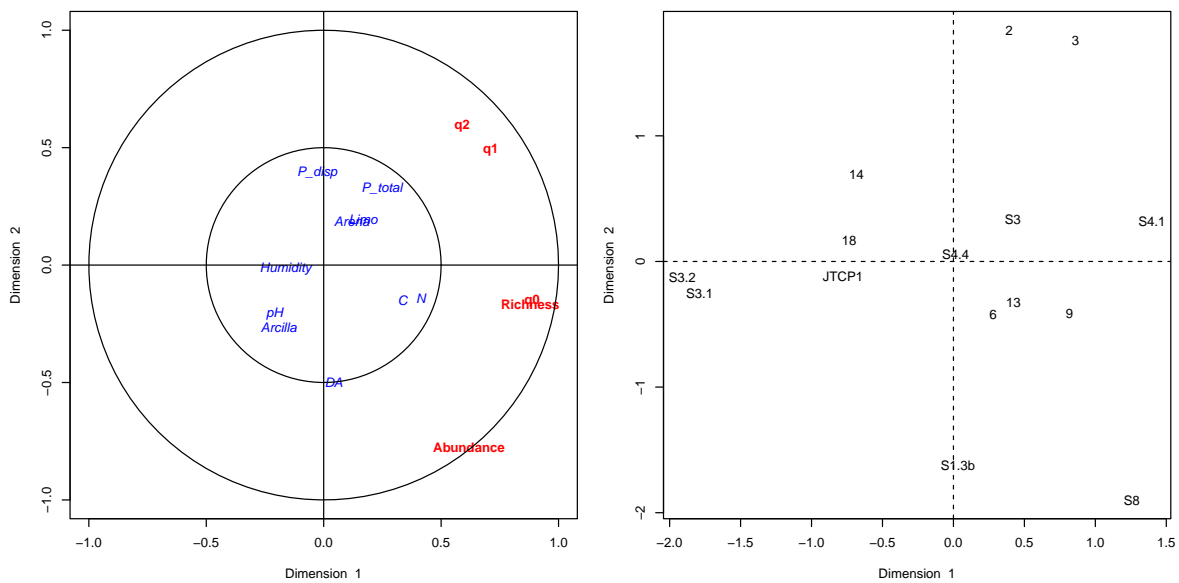
	[,1]	[,2]	[,3]	[,4]	[,5]
pH	-4.435709132	-1.653753573	-2.259058e+00	-1.670308549	2.054296921
Humidity	-0.058955136	0.061734637	4.506629e-04	-0.009833856	-0.097582393

DA	6.033541000	-6.253917331	1.795131e+00	0.015617629	-6.283923838
N	-22.936628986	4.648850573	-2.937635e+01	-5.352747436	46.890802493
C	3.648240062	-0.669148264	2.639224e+00	0.457785279	-5.111987653
P_disp	0.070893486	0.058256198	-5.690386e-02	-0.009732090	-0.107305961
P_total	-0.005923043	-0.004098261	5.496685e-03	0.012075794	0.008560338
Arena	0.104668435	0.060868829	1.985373e-02	-0.087900457	-0.117955005
Arcilla	-0.026390175	-0.047091599	2.232582e-02	0.048477598	0.031151203
Limo	0.204506469	0.040476187	1.301213e-01	-0.163761427	-0.122422052

```
barplot(cc1$cor, xlab = "Dimensión", ylab = "Correlaciones canónicas", ylim = c(0, 1))
```







Prueba de dimensiones canónicas

```
rho <- cc1$cor

n <- dim(ACCDB)[1]
p <- length(dungbeetles)
q <- length(soil)

p.asym(rho, n, p, q, tstat="Wilks")
```

Wilks' Lambda, using F-approximation (Rao's F):

	stat	approx	df1	df2	p.value
1 to 5:	4.907268e-05	0.5253122	50	3.364210	0.8653439
2 to 5:	2.962345e-02	0.2373145	36	5.484777	0.9967445
3 to 5:	1.877498e-01	0.2081070	24	6.401825	0.9979303
4 to 5:	5.071284e-01	0.1732453	14	6.000000	0.9966933
5 to 5:	8.588291e-01	0.1095840	6	4.000000	0.9899378