# FATHOM for Matlab

A Matlab Toolbox for Multivariate Ecological & Oceanographic Data Analysis

by

Dave Jones

## Description

FATHOM: Matlab Toolbox for Ecological & Oceanographic Data Analysis by Dave Jones <a href="mailto:cdipones@rsmas.miami.edu">cdipones@rsmas.miami.edu</a>

#### References

Jones, D. L. 2002. Users manual for FATHOM: a MATLAB toolbox for Multivariate Ecological & Oceanographic Data Analysis. Available from:

http://www.rsmas.miami.edu/personal/djones/

#### **Details**

The FATHOM Toolbox is a collection of Matlab functions and scripts I've written for my everyday use. I'm releasing them to the public in order to encourage the sharing of code and prevent duplication of effort. If you find this toolbox useful, drop me a line. I'd also appreciate bug reports and suggestions for improvements.

While I've made every attempt to write functions that provide accurate and precise results, the functions in this toolbox are provided as is, with no guarantees and are only intended for non-commercial use. These routines are currently being developed and tested with Matlab 6.5 (Release 13) under Windows XP Pro SP1.

Please note that this users manual is currently under construction and is presently incomplete.

#### Note

There is currently only one issue related to installation that you need to be aware of:

f\_nmds: NonMetric Multidimensional Scaling

This function calls Mark Steyver's NMDS routine. For it to work you must install his toolbox from:

http://www-psych.stanford.edu/ msteyver/programs\_data/mdszip.zip

I've been able to obtain better results with this program by editing Steyver's mds.m file and changing

```
randn( 'state', seed );
to
```

# rand( 'state', seed );

This allows you to draw from  $Uniformly\ distributed\ random\ numbers$  for initial configurations rather than  $Normally\ distributed\ random\ numbers.$ 

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#### f\_anosim

## Description

One-way Analysis of Similarity

## Usage

```
[r,p] = f_anosim(xDis,grps,rank,iter,pw,plt)
```

## Arguments

xDis	symmetric distance matrix
grps	row vector designating group membership for objects in xDis
rank	optionally rank distances in xDis (default $= 1$ )
iter	# iterations for permutation test (default = 1000)
pw	do pairwise tests (default $= 1$ )
plt	make diagnostic plot (default $= 0$ )

#### Details

This function performs a multivariate ANOSIM by computing an unstandardized Mantel Statistic between an (optionally ranked) distance matrix and a model matrix; the model matrix is derived from a row vector designating group membership. Results are equivalent to Clarke's method. The permutation test permutes the rows/columns of the distance matrix xDis. Pairwise tests between each group are also optionally run. Permutation tests are based on the complete permutation distribution when it is < 5000, otherwise it is randomly sampled the number of times specified by iter.

ANOSIM assmumes that under the null hypothesis distances within groups are smaller than those between groups, thus significant differences can arise between groups having different dispersions but identical centroids. Diagnostic boxplots are provided as a way to check this and prevent Type I error.

Care must be taken when coding grouping factors (i.e., grps and xDis must be sorted ascending prior to running the function); see example below.

This program has been tested against Clarke's *Primer 5 for Windows* and gives the same results.

#### Value

The function returns the following values:

- r strength of relationship (ranges from -1 to 1)
- p permutation-based probability of no difference between groups

#### Author(s)

Dave Jones

#### References

Clarke, K. R. 1993. Non-parametric multivariate analyses of changes in community structure. Aust. J. Ecol. 18: 117-143.

Clarke, K. R. 2002. Personal Communication.

Legendre, P. & L. Legendre. 1998. Numerical ecology. 2nd English ed. Elsevier Science BV, Amsterdam. (pp.552; 561-562)

#### See Also

f\_anosim2, f\_npManova, f\_mantel, and f\_modelMatrix

#### Examples

Load the file, anosim.mat, in the data folder. This data is from Gray et al. (1990) and ships with Clarke's *Primer* program. There is 1 response variable: species, representing abundances of 174 species (rows) from 39 collection sites (columns), and 1 factor, grps, a row vector specifying group membership (i.e. categories representing distance from the center of an oilfield).

- >> load anosim.mat
- >> species2 = f\_transform(species,1); % square-root transform species abundances:
- >> dis = f\_braycurtis(species2); % Bray-Curtis symmetric distance matrix
- >> [r,p] = f\_anosim(dis,grps)

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## 1-way ANOSIM RESULTS:

-----

## Groupings:

#### Global Test:

R = 0.5413 p = 0.0010 (1000 of 408316439225392430000 possible perms)

Pair-Wise Tests:

```
1 2: R = 0.5539 p = 0.0020 (1000 of 8008 possible perms)

1 3: R = 0.8200 p = 0.0010 (1000 of 18564 possible perms)

1 4: R = 0.9277 p = 0.0010 (1000 of 12376 possible perms)

2 3: R = 0.1596 p = 0.0890 (1000 of 646646 possible perms)

2 4: R = 0.7635 p = 0.0010 (1000 of 352716 possible perms)

3 4: R = 0.5584 p = 0.0010 (1000 of 1352078 possible perms)
```

\_\_\_\_\_

```
r = 0.54128

p = 0.001
```

Gray, J. S., K. R. Clarke, R. M. Warwick, & G. Hobbs. 1990. Detection of initial effects of pollution on marine benthos: an example from the Ekofisk and Eldfisk oilfields, North Sea. Mar. Ecol. Prog. Ser. 66:285–299.

f\_anosim2 Two-way Analysis of Similarity

# Description

Two-way crossed ANOSIM with no replication

## Usage

```
[r,p] = f_anosim2(dis,fac1,fac2,{rank},iter)
```

## Arguments

dis	symmetric distance matrix
fac1	vector of integers (or chars) specifying levels of factor 1 for rows/cols of distance matrix $$
fac2	vector of integers (or chars) specifying levels of factor 2 for rows/cols of distance matrix $$
rank	optionally rank distances in xDis (default $= 1$ )
iter	# iterations for permutatin test (default = 1000)

## **Details**

This program handles missing data in a 2-way layout design that would occur when one (or more) of the treatment levels is missing from a block. The one-tailed permutation-based significance test permutes the treatment levels *separately* within each block.

This program has been tested against Clarke's *Primer 5 for Windows* and gives the same results.

Please be patient when running this program with large datasets and/or high values for iter.

#### Value

The function returns the following values:

r strength of treatment effect (averaged across all blocks)p permutation-based significance test

#### Note

fac1 and fac2 must be equal to row/col size of dis.

## Author(s)

Dave Jones

#### References

Clarke, K. R. & R. M. Warwick. 1994. Similarity-based testing for community pattern: the two-way layout with no replication. Mar. Bio. 118: 167-176.

#### See Also

```
f_anosim, f_npManova, f_mantel, and f_modelMatrix
```

#### Examples

For an example of a Two-Way ANOSIM with no replication load the file, anosim2.mat, in the data folder. This data is from Warwick (1971) and ships with Clarke's PRIMER program. There is 1 response variable, dis, representing a Bray-Curtis symmetric distance matrix from 4th root transformed species abundances, and 2 factors: site and time.

Warwick, R. M. 1971. Nematode associations in the Exe estuary. J. Mar. Biol. Ass. U.K. 51:439–454

f\_bartlett

Bartlett's Test

## Description

Bartlett's Test for Homogeneity of Variances

## Usage

```
[pval, chisq, df] = f_bartlett(x)
```

#### Arguments

x column vector of input data

grps column vector specifying group membership

#### **Details**

Under the null hypothesis of equal variances, the test statistic chisq approximately follows a chi-square distribution with df degrees of freedom; pval is the p-value (1 minus the CDF of this distribution at chisq) of the test.

This program has been tested against the SAS code bartlett.sas and gives similar results.

#### Value

The function returns the following values:

pval probability that the null hypothesis is true

chisq test statistic

df degrees of freedom

## Author(s)

Original Octave code bartlett\_test.m by KH (Kurt.Hornik@ci.tuwien.ac.at). Ported to Matlab by Dave Jones.

# Examples

Load the file, bartlett.mat, in the data folder.

Biotic		Environm	antal	Campai	lation
Dionic	-	Environing	ении	Correct	auton

#### f\_bioenv

## Description

Correlation of primary (biotic) symmetric distance matrix with all possible subsets of secondary (environmental) matrix.

## Usage

```
[res,resLabels] = f_bioenv(dis,matrix,labels,metric,trim,out)
```

## **Arguments**

dis	symmetric distance matrix
matrix	$2^{\circ}$ matrix (rows = variables, cols = samples)
labels	cell array of variable labels of $2^{\circ}$ matrix; e.g., labels = {'temp' 'sal' 'depth' '02'}
metric	distance metric to use for 2° matrix; 0 = Euclidean (default); 1 = Bray-Curtis
trim	return only this many of the top Rho's per subset size class; $0=$ return all (default)
out	<pre>send results to screen (= 1, default) or cell array with filename; e.g., out = {'results.txt'}</pre>

#### Value

The function returns the following values:

res cell array, 1st column = Rho, 2nd:end are variable indices

resLabels cell array of variable names

#### Note

This function requires combnk.m from the *Matlab Statistics Toolbox*. The code could be modified to utilize the freely available choosenk.m instead.

For out, an existing file with the same name as specified will be **deleted**. Tabulated results are also sent to the screen or file, depending on the value of out.

The # of rows of dis must equal the # columns of matrix.

## Author(s)

Dave Jones

#### References

Clarke, K. R. & M. Ainsworth. 1993. A method of linking multivariate community structure to environmental variables. Mar. Ecol. Prog. Ser. 92:205-219.

Legendre, P. & L. Legendre. 1998. Numerical ecology. 2nd English ed. Elsevier Science BV, Amsterdam. xv + 853 pp.

#### See Also

f\_braycurtis, f\_euclid

#### Examples

Load the file, spiders.mat, in the data folder. This is data from van der Aart & Smeenk-Enserink (1975) and is Hunting spider abundances for 12 species (spiders) taken from 28 sites (site\_labels) and associated environmental data (env).

```
>> load spiders.mat
>> spiders2 = f_transform(spiders',3)'; % log-transform abundances
>> dis = f_brayCurtis(spiders2'); % Bray-Curtis dissimilarity matrix
>> [res,resLabels] = f_bioenv(dis,env',env_labels,0,5,1);
There are 63 possible subsets of 6 variables
 Processing 6 subsets of
                        1 variables
 Processing 15 subsets of 2 variables
 Processing 20 subsets of 3 variables
 Processing 15 subsets of 4 variables
 Processing 6 subsets of 5 variables
 Processing
            1 subsets of 6 variables
_____
      Variables
```

0.7037 water

0.5486 light

0.5197 sand

0.4962 cover

0.4097 twigs

0.2611 herbs

```
2
0.7591 twigs water
0.7578 light water
0.7115 light sand
0.7073 cover water
0.7018 sand water
0.8110 sand twigs water
0.7870 light sand water
0.7561 cover light sand
0.7505 cover sand water
0.7496 cover twigs water
0.8159 cover sand twigs water
0.7995 cover light sand water
0.7982 light sand twigs water
0.7845 cover herbs sand water
0.7751 herbs light sand water
0.8065 cover light sand twigs water
0.8061 cover herbs light sand water
0.8047 cover herbs sand twigs water
0.7723 herbs light sand twigs water
0.7270 cover herbs light twigs water
0.7996 cover herbs light sand twigs water
```

van der Aart, P. J. M. & N. Smeenk-Enserink. 1975. Correlations between distributions of hunting spiders (Lycosidae, Ctenidae) and environmental characteristics in a dune area. Netherlands Journal of Zoology 25: 1–45.

Eigenvector-based 2d distance biplot

f\_biplot

## Description

This function is a general function used to distance biplot based on eigenvectors. Vectors are plotted for each variable comprising the original data matrix for which an eigen-analysis was performed. The direction each vector points indicates the direction of increase of each variable (i.e., gradient). Length of vectors indicate the relative contribution each variable contributes to the formation of the reduced space plotted (e.g., that space defined by the  $1^{st}$  2 Principal Component axes).

## Usage

```
f_biplotPca2(evects,scale,offset,sLabels);
```

#### Arguments

## Author(s)

Dave Jones

#### References

Legendre, P. & L. Legendre. 1998. Numerical ecology.  $2^{nd}$  English ed. Elsevier Science BV, Amsterdam.

#### See Also

```
f_pca, f_biplotEnv2, f_biplotSpecies, f_vectorfit
```

f\_biplotEnv2

 $Biplot\ for\ 2\text{-}d\ Ordination$ 

# Description

Create a distance biplot consisting of a 2-d ordination with environmental correlation vectors.

## Usage

```
[biplot,Rsq] = f_biplotEnv2(crds,env,special,iter,scale,offset,sLabels)
```

## Arguments

crds	matrix of ordination coordinates (rows $=$ sites; cols $=$ dimensions)
env	matrix of (transformed) environmental variables (rows = sites, $cols = variables$ )
special	type of correlation; $0 = Pearson's$ , $1 = Spearman's$ (default)
iter	# of iterations for permutation test (default = 0)
scale	scaling factor for env vectors (default $= 1$ )
offset	offset of labels in plot (default $= 0$ )
sLabels	cell array of vector labels; if empty they are autocreated e.g., sLabels
	= {'sal' 'tmp' 'elev'}

#### Value

The function returns the following values:

biplot	2 column matrix sepcifying x-y coordinates of endpoints for scaled env
	vectors to overlay on ordination
Rsq	column matrix of correlation with each axis (a permutation test is performed when iter>0)

# Author(s)

Dave Jones

#### References

```
Legendre, P. & L. Legendre. 1998. Numerical ecology. 2nd English ed. Elsevier Science BV, Amsterdam. xv + 853 pp. (page 586)
Legendre, P. 2001. (personal communication).
```

# See Also

f\_biplotEnv3, f\_biplotSpecies, f\_vectorfit

f\_biplotEnv3

Biplot for 3-d Ordination

## Description

Create a distance biplot consisting of a 3-d ordination with environmental correlation vectors.

## Usage

```
[biplot,Rsq] = f_biplotEnv3(crds,env,special,iter,scale,offset,sLabels,plotflag,minP)
```

# Arguments

crds	matrix of ordination coordinates (rows $=$ sites; cols $=$ dimensions)
env	matrix of (transformed) environmental variables (rows = sites, cols = variables)
special	type of correlation; $0 = Pearson's$ , $1 = Spearman's$ (default)
iter	# of iterations for permutation test (default = 0)
scale	scaling factor for env vectors (default $= 1$ )
offset	offset of labels in plot (default $= 0$ )
sLabels	<pre>cell array of vector labels; if empty they are autocreated e.g., sLabels = {'sal' 'tmp' 'elev'}</pre>
minP	if p-value of correlation > minP, then correlation is NOT used (default = 0.05)

# Value

The function returns the following values:

biplot 3 column matrix sepcifying xyz coordinates of endpoints for scaled env

vectors to overlay on ordination

Rsq column matrix of correlation with each axis (permutation-based sig-

nificance provided when iter>0)

# Author(s)

Dave Jones

# References

Legendre, P. & L. Legendre. 1998. Numerical ecology. 2nd English ed. Elsevier Science BV, Amsterdam. xv + 853 pp. (page 586)

Legendre, P. 2001. (personal communication).

# See Also

 ${\tt f\_biplotEnv2}, \, {\tt f\_biplotSpecies}, \, {\tt f\_vectorfit}$ 

 $f_biplotSpecies$   $S_I$ 

 $Species\ Biplot$ 

## Description

Create species vectors for ordination distance biplot.

## Usage

```
[biplot,rsq] = f_biplotSpecies(crds,species,special,iter,scale,offset,sLabels);
```

## **Arguments**

crds	matrix of ordination coordinates (rows = sites; cols = eigenvectors or dimensions)
species	matrix of (transformed) species abundances (rows = sites, $cols = variables$ )
special	type of correlation; $0 = Pearson's$ (default), $1 = Spearman's$
iter	# of iterations for permutation test (default = 0)
scale	scaling factor for species vectors (default $= 1$ )
offset	offset of labels in plot (default $= 0$ )
sLabels	<pre>cell array of species labels (if empty autocreate) e.g., sLabels = {'sp1' 'sp2' 'sp3'}</pre>

#### Value

The function returns the following values:

biplot 2 column matrix sepcifying x-y coordinates of endpoints for scaled

species vectors to overlay on ordination

Rsq column matrix of correlation with each axis (permutation-based sig-

nificance provided when iter>0)

# Author(s)

Dave Jones

## References

Legendre, P. & L. Legendre. 1998. Numerical ecology. 2nd English ed. Elsevier Science BV, Amsterdam. xv + 853 pp.

Legendre, P. & E. Gallagher. Ecologically meaningful transformations for ordination biplots of species data. Oecology 129: 271-280.

# See Also

f\_biplotEnv2, f\_biplotEnv3, f\_vectorfit

f\_braycurtis

Bray-Curtis Dissimilarity

## Description

This functions calculates a Bray-Curtis symmetric dissimilarity matrix from an input data matrix specifying species abundances per sample site.

#### Usage

```
dist = f_braycurtis(X);
```

## Arguments

X

species x site data matrix

#### Details

#### Value

The function returns the following values:

dist

symmetric dissimilarity matrix

#### Note

The input data matrix should be codes as species (rows) versus sites (columns). An  $n \times n$  symmetric dissimilarity matrix will be returned where n is the number of sites.

## Author(s)

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All standard disclaimers apply.

Slightly modified by Dave Jones after distance.m in the *EDAT Toolbox* to only calculate a Bray-Curtis distance matrix between columns.

# See Also

f\_euclid

f\_brokenstick

Broken-Stick model

## Description

This function is used to determine the # of significant ordination dimensions via the Broken-Stick model.

## Usage

```
f_brokenstick(nvars)
```

#### Arguments

nvars

# of variables (e.g., sample sites)

#### Value

The function returns the following values:

varExplained % variance explained

## Author(s)

Original Matlab code, brokestk.m, by R. E. Strauss, modified by Dave Jones.

#### References

Frontier, S. 1976. Etude de la decroissance des valeurs propres dans une analyze en composantes principales: comparison avec le modele de baton brise. J. Exp. Mar. Biol. Ecol. 25:67–75.

Jackson, D. A. 1993. Stopping rules in principal components analysis: a comparison of heuristial and statistical approaches. Ecology 74:2204–2214.

Legendre, P. & L. Legendre. 1998. Numerical ecology. 2nd English ed. Elsevier Science BV, Amsterdam. xv + 853 pp. (p. 410)

## See Also

f\_pca

## f\_cap

# Description

The function performs a Canonical Analysis of Principal Coordinates using any distance (dissimilarity) matrix.

# Usage

```
[crds,trc,ccor,H,p1,p2,centroids] = f_cva(yDis,x,rank,iter,plt,verb)
```

# Arguments

yDis	square symmetric distance matrix derived from response variables
x	<ul> <li>(1) vector of integers specifying group membership for objects in yDis,</li> <li>(2) ANOVA design matrix specifid by dummy coding, or (3) matrix of explanatory variables (rows = observations, cols = variables)</li> </ul>
rank	optionally rank distances in yDis (default = $0$ )
iter	# iterations for permutation test (default = 0)
plt	optionally plot results (default $= 1$ )
verb	optionally send results to display (default $= 1$ )

## Value

The function returns the following values:

crds	coordinates of canonical axes $(= Qstar)$
trc	trace statistic
ccor	canonical eigenvalues (1st value is greatest root statistic)
p1	randomized probability of trace statistic
p2	randomized probability of greatest root statistic
centroids	centroids of groups defined in x

#### Note

This program performs nonparametric Multiple Discriminant Analysis on **any** symmetric distance matrix when the input for  $\mathbf{x}$  is (1) a vector specifying group membership. It performs generalized Canonical Variates Analysis when  $\mathbf{x}$  is (2) an ANOVA design matrix or (3) a matrix of explanatory variables.

Use **f\_designMatrix** to create an ANOVA design matrix for input as **x**; the matrix should be full rank (not singular) and do **not** include an intercept term (a column of all 1's).

The program asks the user to specify how many axes of Q to retain for the analysis (m). Examine the EIGENVALUES and % VARIATION EXPLAINED output in the command window and try to include as much information in Q as possible with as few axes as possible.

This program has been tested against the numerical example in Legendre & Legendre (1998:p.626) and provides similar output. The % variation explained for each canonical axis, calculated by f\_cap, are almost identical to that computed by manoval.m in the *Matlab Statistical Toolbox* (version 3) for the Iris data.

This program requires ortha.m by Andrew Knyazev <knyazev@na-net.ornl.gov> & Rico Argentati, which is included the present toolbox and available from:

http://www-math.cudenver.edu/aknyazev/software/MATLAB/

#### Author(s)

Dave Jones

## References

Anderson, M. J. 2002. CAP: a FORTRAN program for canonical analysis of principal coordinates. Dept. of Statistics University of Auckland. Available from:

http://www.stat.auckland.ac.nz/PEOPLE/marti/

Anderson, M. J. & T. J. Willis. 2003. Canonical analysis of principal coordinates: a useful method of constrained ordination for ecology. Ecology 84(2): 511-525.

Legendre, P. & L. Legendre. 1998. Numerical ecology. 2nd English ed. Elsevier Science BV, Amsterdam. pp.616–631.

#### See Also

f\_cda,f\_npManova, f\_anosim, f\_anosim2, f\_mantel, f\_designMatrix

## Examples

The example dataset, iris.mat, is Fisher's Iris data and can be found in the data folder

```
>> load iris.mat
>> dis = f_euclid(iris');
>> [crds,trc,ccor,H,p1,p2,centroids] = f_cap(dis,grps,0,1000,1,1);
      Matrix Q:
-----
+ Eigenvalues: % Explained: Axis:
   630.008 92.46187
36.15794 97.76852
11.65322 99.47878
3.551429 100
                        1
                               2

    3.551429
    100

    4.391243e-013
    100

                               4
                          5 ...
1.034584e-016
                   100
                              126
_____
Specify # of Axes in Q to retain (1-126) ? 2
Permuting the data 999 times...
_____
Nonparametric Canonical Discriminant Analysis:
Trace Stat = 368.1700 p = 0.00100
Greatest Root = 366.1265 p = 0.00100
No. of permutations = 1000
_____
No. of axes of Q used (m) = 2
Canonical Correlations:
 366.1265 2.0435
_____
```

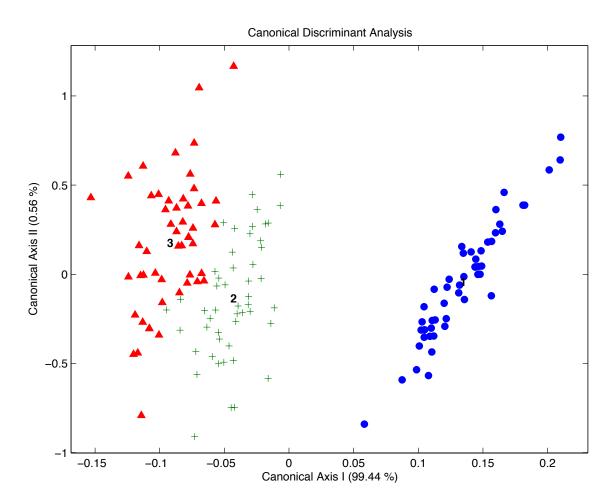


Figure 1: Example of  $f_cap$ 

Canonical Discriminant Analysis

#### f\_cda

## Description

This function is used to perform a classical Canonical Discriminant Analysis.

## Usage

```
[scores,centroids,Cvects] = f_cda(x,y,method,pflag)
```

## Arguments

```
x input data (rows = objects, cols = variables)
```

y column vector of integers specifying group membership

method center (=1, default) or stardardize (=2)

pflag make canonical plot (default = 0)

# Details

method = 1: variables comprising matrix x are centered, so columns of the canonical eigenvectors (Cvects) are *Identification Functions*. Use this method when you plan to place new objects in the canonical space.

method = 2: variables comprising matrix x are standardized, so columns of the canonical eigenvectors (Cvects) are *Discriminant Functions*. Use this method to assess the relative importance of the original variables of x in discriminating groups.

#### Value

The function returns the following values:

scores coordinates in canonical space (= canonical variates)

centroids group centers

cvects canonical eigenvectors

#### Note

scores are the coordinates of the original (centered, or standardized) data projected in new canonical space. They are obtained by multiplying the original data by the Canonical Eigenvectors.

centroids are the coordinates of the group means projected in the new canonical space.

Cvects (Canonical Eigenvectors) are the normalized orthogonal eigenvectors defining the canonical space of the discriminant analysis.

## Author(s)

Dave Jones

#### References

Legendre, P. & L. Legendre. 1998. Numerical ecology.  $2^{nd}$  English ed. Elsevier Science BV, Amsterdam.

## See Also

f\_cap

# Examples

The example dataset, iris.mat, is Fisher's Iris data and can be found in the data folder

```
>> load iris.mat
>> [scores,centroids,Cvects] = f_cda(iris,grps,2,1);
```

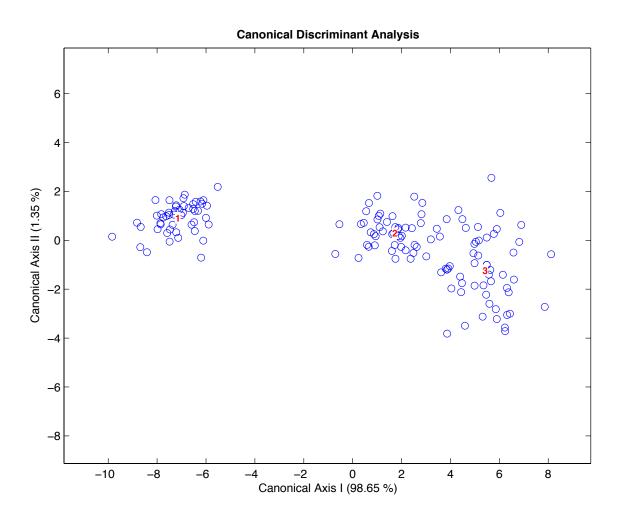


Figure 2: Example of f\_cda  $\,$ 

f\_figArea

Area of Figure Objects

## Description

This function is used to compute the Area of figure objects in a plot and optionally returns a list of handles sorted ascending by Area.

## Usage

```
[area,h2] = f_figArea(h);
```

## Arguments

h handle of figure object(s)

#### Value

The function returns the following values:

area for object specified by h

h2 figure handles sorted by area (ascending)

#### Note

This is particularly useful when you've plotted a number of patches but, because of the stacking order, the smaller ones are obscured by the larger.

#### Author(s)

Dave Jones

#### Examples

After creating a plot with potentially overlapping patches, run the following commands:

f\_importSurfer

Import Surfer Blanking File

## Description

This program is used to import a Golden Software's *Surfer for Windows* bln file into Matlab in a format that can be used by the M\_Map Toolbox.

#### Usage

```
[ncst,k,Area] = f_importSurfer('fname');
```

## **Arguments**

fname

name of Surfer \*.bln blanking file

#### Details

Be sure to save variables ncst,k, and Area as a usercoast.mat file.

#### Value

The function returns the following values:

ncst,k,Area required components for M\_Map usercoast.mat file

#### Note

Since Surfer can import ArcView Shape files, this is a good way to get shp data into Matlab. The suggested procedure is to turn off the display of axes in Surfer, then export the data as a bln file (select the option *Break Apart Compound Areas*), import using f\_importSurfer, save as a usercoast file, and plot using m\_usercoast.m.

## Author(s)

Dave Jones

#### References

Portions of this code are from the comments in mu\_coast.m from Rich Pawlowicz's <rich@ocgy.ubc.ca> M\_Map Toolbox available from:

```
http://www2.ocgy.ubc.ca/ rich/map.html
```

# Examples

Example code to plot as filled patches using M\_Map toolbox:

```
>> m_proj('mercator','longitudes',[-81 -80],'latitudes',[24.5 25.5]);
>> m_usercoast('fmri.mat','patch',[0 0 0],'edgecolor','none');
>> m_grid('box','fancy','fontsize',8,'linestyle','none','xtick',[-81:-80],'ytick',[24.5:25.5]);
```

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Multinle	Linear	Regression	n
IVI WUUUPUC	Dorocar	1 CCG1 COOLC	,,,

## ${\tt f\_mregress}$

# Description

This program performs Multiple Linear Regression using Least Squares Estimation via Matlab's QR factorization.

# Usage

```
[F,t,R2,yfit,b,resid] = f_mregress(x,y,iter,perm,verb);
```

## Arguments

X	matrix of independent variables (column-wise)
У	column vector of dependent variable
iter	# of iterations for permutation test (default = 0)
perm	permute residuals instead of raw data (default $= 1$ )
verb	verbose output of results to display (default $= 1$ )

#### **Details**

This function solves the equation such that:

```
y = b(0) + b(1)*(X(:,1)) + b(2)*(X(:,2))...+ b(k)*(X(:,k)), where k = # of predictor variables.
```

## Value

The function returns the following values (F and t are structures):

F.stat	F-statistic
F.para_p	parametric p-value for 1-tailed test of F
F.perm_p	permutation p-value for 1-tailed test of F
t.stat	t-statistic for partial regression coefficients
t.para_p	parametric p-value for 1-tailed test of t
t.perm_p	permutation p-value for 1-tailed test of t
R2	coefficient of multiple determination $(R^2, goodness-of-fit)$
yfit	fitted values of y
b	regression coefficients (1st value is the y-intercept)
resid	residuals

#### Note

The regression coefficients are computed using Least Squares Estimation (via the \ operator), which is preferred over methods that require taking the inverse of a matrix. R2, the coefficient of multiple determination, is a measure of goodness-of-fit and gives the proportion of variance of Y explained by X.

Parametric (and optional permutation) tests of significance for the F- and t-statistics are performed. The permutation test is conducted when  $\mathtt{iter} > 0$  and allows for permutation of either the raw data or the residuals of the full regression model. Permutation of the raw data involves random permutation of the rows (= observations) of Y relative to the rows of X. The permutation test is preferred over the parametric test when the data are non-normal. Permutation of the residuals (vs. the raw data) is preferred when data have extreme values (i.e., outliers).

This function has been tested against Legendre & Casgrain's regressn.exe program and gives similar output.

Calculation of parametric p-values for F and t require fpdf and tcdf from the *Matlab Statistics Toolbox*; these could be replaced by df and dt from the free *Stixbox Toolbox*.

## Author(s)

Dave Jones with help from news://comp.soft-sys.matlab

#### References

Legendre, P. & L. Legendre. 1998. Numerical ecology. 2nd English ed. Elsevier Science BV, Amsterdam. xv + 853 pp. (pp. 517, 606-612)

Legendre, P. 2002. Program for multiple linear regression (ordinary or through the origin) with permutation test - User's notes. Depart. of Biological Sciences, University of Montreal. 11 pages. Available from: http://www.fas.umontreal.ca/biol/legendre/

Neter, J., W. Wasserman, & M. H. Kutner. 1989. Applied linear regression models. 2nd Edition. Richard D. Irwin, Inc. Homewood, IL.

Sokal, R. R. & F. J. Rohlf. 1995. Biometry - The principles and practice of statistics in biological research. 3rd ed. W. H. Freeman, New York. xix + 887 pp.

## Examples

The example dataset, mregress.mat, can be found in the *data* folder and is an excerpt from Table 16.1 of Sokal & Rohlf (1995).

```
>> load mregress.mat
>> tic;
>> [F,t,yfit,coefs,resid] = f_mregress(x,y,10000);
>> toc;
```

Permuting the data 9999 times...

-----Multiple Linear Regression via QR Factorization:

R2 F-	Stat	parametric-p	permutation-p
0.51611 20	).26553	0.00000	0.00010

Variable	b	t-stat	parametric-p	permutation-p
_	77.23671	3.59122	0.00045	0.00060
	-1.04805	-2.80777	0.00384	0.00220
	0.02430	5.07607	0.00000	0.00050

<sup>#</sup> permutations of residuals = 9999 All significance tests are One-Tailed

elapsed\_time = 57.092

Minimum Spanning Tree

#### f\_mst

#### Description

This function uses Kruskal's algorithm to calculate a Minimum Spanning Tree for objects based on pair-wise distances specified in a symmetric distance matrix.

#### Usage

```
[mst,branch,scaled] = f_mst(dis,crds)
```

# Arguments

dis symmetric distance matrix

crds Euclidean coordinates of an ordination based on dis

#### Value

The function returns the following values:

mst pairs of objects connected by MST branches

branch corresponding branch lengths
scaled branch lengths rescaled to 0-100

#### Note

MST's are particularly useful for checking or interpreting the results of an ordination by overlying a MST on an ordination plot.

#### Author(s)

Dave Jones

#### Examples

Load the file, spiders.mat, in the data folder. This is data from van der Aart & Smeenk-Enserink (1975) and is Hunting spider abundances for 12 species (spiders) taken from 28 sites (site\_labels) and associated environmental data (env).

[van der Aart, P. J. M. & N. Smeenk-Enserink. 1975. Correlations between distributions of hunting spiders (Lycosidae, Ctenidae) and environmental characteristics in a dune area. Netherlands Journal of Zoology 25: 1–45.]

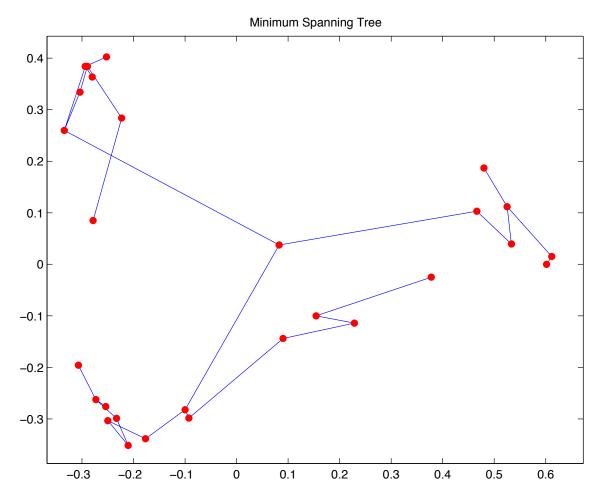


Figure 3: Example of f\_mst

```
>> load spiders.mat
>> dis = f_braycurtis(spiders');
>> evects = f_pcoa(dis,0,0,0);
>> mst = f_mst(dis,evects);
```

Non-metric Multidimensional Scaling

#### $f_nmds$

#### Description

This function is used to run Mark Steyvers' Non-metric Multidimensional Scaling program.

#### Usage

```
config = f_nmds(dist,ndims,initial,plotflag,maxiter,conv,rotate);
```

# Arguments

dist	symmetric dissimilarity matrix
ndims	number of dimensions of solution (default $= 2$ )
initial	initial config: 1 = Torgeson-Young scaling (default); 0 = random
plotflag	plot results (default $=1$ )
maxiter	$\max \# \text{ iterations (default} = 50)$
conv	convergence criterion (default = $0.001$ )
rotate	rotate final configuration to Principal Coordinates (default $= 1$ )

#### Value

The function returns the following values (config is a structure):

#### Note

This program requires the *Matlab Optimization Toolbox* and Mark Steyvers' Nonmetric Scaling Toolbox, available from:

http://www-psych.stanford.edu/~msteyver/programs\us data/mdszip.zip

I've been able to obtain best results with this program by editing Steyver's mds.m file and changing randn('state', seed') to rand('state', seed'). This allows you to draw from *uniformly* distributed random numbers for initial configurations rather than *normally* distributed random numbers.

# Author(s)

Modified after Mark Steyvers' original Matlab code, domds.m, by Dave Jones. Added support for variable random seed, rotate to PCA, Mantel statistic, and formatting of output.

#### See Also

f\_pcoa, f\_pca

#### Examples

Load the file, spiders.mat, in the data folder. This is data from van der Aart & Smeenk-Enserink (1975) and is Hunting spider abundances for 12 species (spiders) taken from 28 sites (site\_labels) and associated environmental data (env).

[van der Aart, P. J. M. & N. Smeenk-Enserink. 1975. Correlations between distributions of hunting spiders (Lycosidae, Ctenidae) and environmental characteristics in a dune area. Netherlands Journal of Zoology 25: 1–45.]

```
>> load spiders.mat
>> dis = f_braycurtis(spiders');
>> config = f_nmds(dis, 2, 0, 0, 50, 0.001, 1);
Values added to satisfy distance axioms 0 0
Minkowski metric r value
                                         2
Number of objects in matrix
                                         28
Number of useful similarity ratings
                                         378
Number of dimensions:
                                         2
Maximum number of iterations
                                         50
                                         0.001
Convergence Criterion:
Minimize function:
                                         stress1
Starting configuration is:
                                         random (seed=213210)
The following results apply to training file only:
(Rs is the rank order correlation coefficient
between observed and predicted dissimilarities
t=0 Stress1=0.42659 Stress2=0.99278 Rs=0.02823
```

```
t=1 Stress1=0.60586 Stress2=0.99538 Rs=-0.05871
t=2 Stress1=0.40898 Stress2=0.98782 Rs=-0.03861
t=3 Stress1=0.38038 Stress2=0.96809 Rs=0.02299
t=4 Stress1=0.36543 Stress2=0.92466 Rs=0.13417
t=5 Stress1=0.33233 Stress2=0.82448 Rs=0.34933
t=6 Stress1=0.27412 Stress2=0.66189 Rs=0.62448
t=7 Stress1=0.18919 Stress2=0.43455 Rs=0.83975
t=8 Stress1=0.15189 Stress2=0.33059 Rs=0.91154
t=9 Stress1=0.13544 Stress2=0.28669 Rs=0.94157
t=10 Stress1=0.09710 Stress2=0.20584 Rs=0.97019
t=11 Stress1=0.09284 Stress2=0.19665 Rs=0.97251
t=12 Stress1=0.09224 Stress2=0.19559 Rs=0.97280
End of simulation - reason: convergence criterion reached
>> plot(config.mds(:,1),config.mds(:,2),'bo');
>> title([['\bfNonmetric MDS (Stress = '] num2str(config.stress) [')']]);
>> xlabel('Dim 1');
>> ylabel('Dim 2');
```

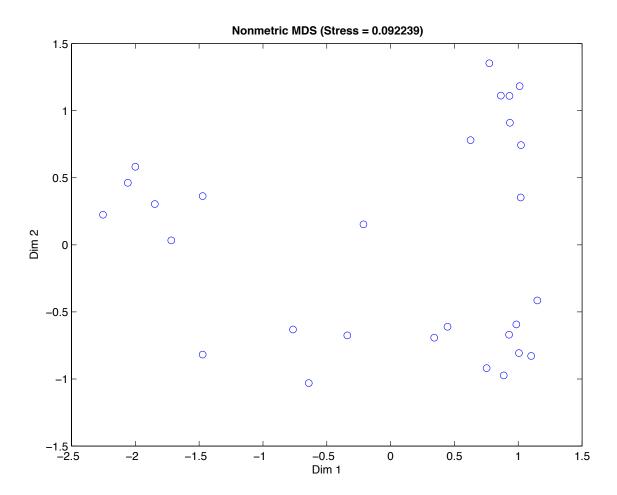


Figure 4: Example of f\_nmds

I_npManova	f_npManova	Nonparametric MANOVA
------------	------------	----------------------

# Description

This program performs a nonparametric, permutation-based MANOVA on any symmetric distance (or dissimilarity) matrix.

# Usage

```
result = f_npManova(yDis,x,type,rank,iter,pw,verb)
```

# Arguments

yDis	square symmetric distance matrix derived from response variable(s) $$
x	matrix of integers specifying group membership for objects in yDis (column-wise) or a matrix of explanatory variables
type	x specifies group membership or factor levels (default = 0); x is a matrix of explanatory variables $(=1)$
rank	optionally rank distances in yDis (default $= 0$ )
iter	# iterations for permutation test (default = 0)
verb	optionally send results to display (default $= 0$ )

# **Details**

The permutation test in this program currently only supports permutation of the raw data.

# Value

This function returns the following values (result is a structure):

result.so	source of variation
result.df	degrees of freedom
result.SS	sum of squares
result.MS	mean square
result.F	F-statistics
result.p	permutation-based significance probabilities

#### Note

Special care must be taken when coding levels of nested factors, i.e. treatment levels of a nested factor must not be **repeated** across different levels of the main factor. For example use: main factor = [1 1 1 2 2 2 3 3 3]' nested factor = [1 2 3 4 5 6 7 8 9]'

instead of: nested factor = [1 2 3 1 2 3 1 2 3]'

More examples can be found in Appendix A.

This program takes a regression approach to ANOVA using the General Linear Model (GLM) and constructs F-ratios using the **unrestricted** form of the model. The F-ratios used for each type of test are provided in Appendix B of the User's Manual. Some of these differ somewhat from those used in textbook examples, especially for balanced, mixed-model designs, but are the same as those used in most computer programs that use GLM (e.g., SAS and MINITAB).

To perform a classical (M)ANOVA use a symmetric Euclidean distance matrix as input for the response variable. The real power of this function, however, comes from its ability to use, say, a Bray-Curtis distance matrix derived from species abundances from a community ecology study.

After determining a significant factor effect, you may wish to use f\_npManovaPW to perform a posteriori multiple comparison tests.

#### Author(s)

Dave Jones

#### References

Anderson, M. J. 2002. DISTML v.2: a FORTRAN computer program to calculate a distance-based multivariate analysis for a linear model. Dept. of Statistics University of Auckland. Available from:

http://www.stat.auckland.ac.nz/PEOPLE/marti/

Anderson, M. J. 2000. NPMANOVA: a FORTRAN computer program for non-parametric multivariate analysis of variance (for any two-factor ANOVA design) using permutation tests. Dept. of Statistics, University of Auckland. Available from:

http://www.stat.auckland.ac.nz/PEOPLE/marti/

Anderson, M. J. 2001. A new method for non-parametric multivariate analysis of variance. Austral Ecology 26: 32-46.

McArdle, B. H. & M. J. Anderson. 2001. Fitting multivariate models to community data: a comment on distance-based redundancy analysis. Ecology 290-297.

Sokal, R. & F. J. Rohlf. 1995. Biometry — The principles and practice of statistics in biological research. 3rd ed. W. H. Freeman, New York. xix + 887 pp.

Underwood, A. J. 1981. Techniques of analysis of variance in experimental marine biology and ecology. Oceanogr. Mar. Biol. Ann. Rev. 19: 513-605.

Zar, J. H. 1999. Biostatistical analysis. 4th ed. Prentice Hall, Upper Saddle River, NJ.

#### See Also

f\_npManovaPW,f\_cap, f\_anosim, f\_anosim2, Appendix A, Appendix B

#### Examples

For an example of a One-Way Model I Anova with replication (balanced design) load the file, sr\_09p5.mat, in the data folder. This is data from Box 9.5 of Sokal & Rohlf (1995) and has 1 response variable, age, and 1 fixed factor: clone.

```
>> load sr_09p5.mat
>> dis = f_euclid(age');
>> result = f_npManova(dis,[clone],0,0,1000,1);
```

Permuting the data 999 times...

Nonparametric	(Termudation	Daseu)	TIANOVA.	

'Source'	'df'	'SS'	'MS'	'F'		'n,	,
'factor 1'	[ 1]	[0.006428	6] [0.0064286]	[0.0	014063]	[0	.872]
'residual'	[12]	[ 5.485	7] [ 0.45714]	[	NaN]	[	NaN]
'total'	[13]	[ 5.492	1] [ NaN]	[	NaN]	[	NaN]

# iterations = \_\_\_\_\_

(Note: NaNs are placeholders for the ANOVA table)

1000

(Data with replication)

See Appendix A for more examples.

#### Description

This function is used to perform PCA using either the covariance or the correlation matrix of the input data. Use the *covariance matrix* when the variables are of the same kind, type, and scale; otherwise, use the *correlation matrix*.

#### Usage

```
[scores, evects, evals, expl] = f_pca(x,pflag,method);
```

#### **Arguments**

x	data matrix (rows = objects, $cols = variables$ )
pflag	no plot (=0, default); scores (=1); scores +scree (=2); scores, scree, + equilibrium circle (=3)
method	use covariance (=1 default) or correlation matrix (=2)

### **Details**

Principal Component **scores** are the coordinates of the **objects** from the input data matrix in the new space defined by the Principal Component Axes. The **eigenvectors** are the loadings of the original **variables** which, when multiplied by the original data, yield the Principal Component **scores**. The matrix of **eigenvectors** has a row for each **variable** of the original data matrix and a column for each Principal Component Axis. The **eigenvalues** give the variance of the **scores** along each Principal Component Axis.

The SCREE PLOT is a graphical method of evaluating how may PC Axes you need to retain in order to adequately represent the variation in the original data matrix—it provides the same information as expl.

The EQUILIBRIUM CIRCLE provides a graphical method of determining the relative contribution each **variable** makes to the formation of the reduced spaced defined by PC Axis I & II. Only **variables** whose vectors extend to or beyond the radius of the Equilibrium Circle are considered to have made a significant contribution to that reduced space.

#### Value

The function returns the following values:

scores coordinates of objects on each PC axis

evects eigenvectors (= loadings)
evals eigenvalues for each PC axis

expl % percent variation explained by each PC axis (row 1); cumulative %

variation explained (row 2)

#### Note

This function implements PCA via Singular Value Decomposition of an association matrix formed from the original data. This matrix is decomposed into object space (U), variable space (V), and a diagonal matrix with singular variables along its diagonal (D) such that [association matrix] =  $[U^*D^*V]$ , where eigenvectors = U and eigenvalues = the diagonal elements along D.

### Author(s)

Dave Jones

#### References

Legendre, P. & L. Legendre. 1998. Numerical ecology.  $2^{nd}$  English ed. Elsevier Science BV, Amsterdam.

#### See Also

```
f_biplot, f_nmds, f_pcoa, f_cap
```

### Examples

Load the file, spiders.mat, in the data folder. This is data from van der Aart & Smeenk-Enserink (1975) and is Hunting spider abundances for 12 species (spiders) taken from 28 sites (site\_labels) and associated environmental data (env).

[van der Aart, P. J. M. & N. Smeenk-Enserink. 1975. Correlations between distributions of hunting spiders (Lycosidae, Ctenidae) and environmental characteristics in a dune area. Netherlands Journal of Zoology 25: 1–45.]

```
>> load spiders.mat
>> f_pca(env,3,2);
```

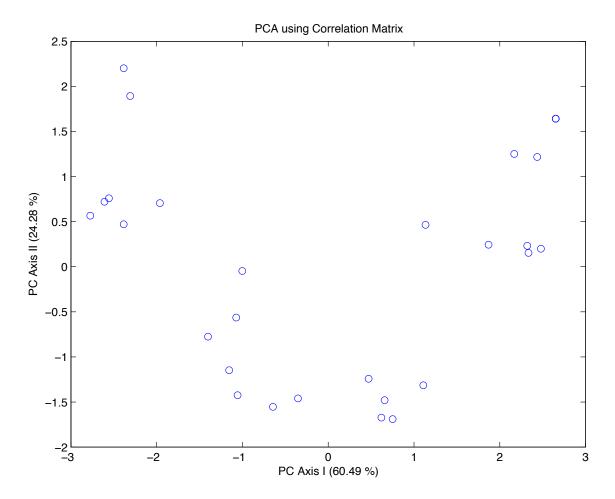


Figure 5: Example of f\_pca plot

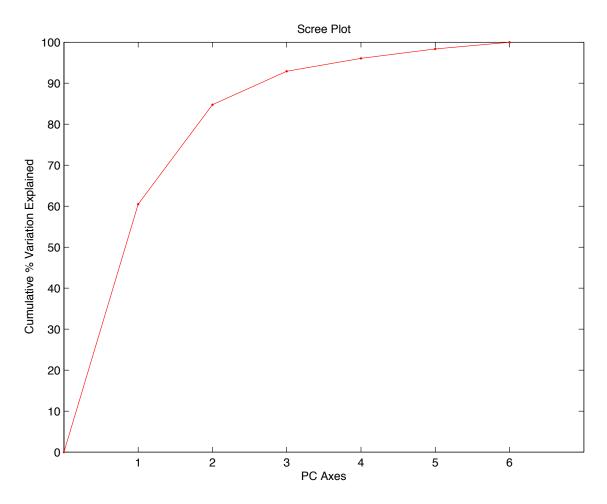


Figure 6: Example of f\_pca Scree plot

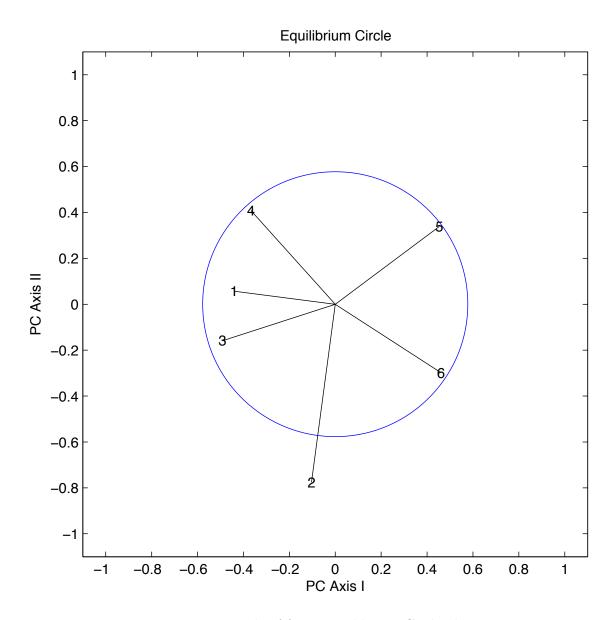


Figure 7: Example of f\_pca Equilibrium Circle plot

f	procrustes
	procrabics

Procrustes Analysis

#### Description

Generalized orthogonal Procrustes rotation of matrix Y to X, maximizing similarity and minimizing sum-of-squared distances.

## Usage

```
[Xscl, Yrot, m2, resid, prob] = f_procrustes(X, Y, stnd, iter, plotflag)
```

## Arguments

X	reference matrix (rows = observations, $cols = variables$ )
Y	target matrix to rotate
stnd	standardize variables (default $= 0$ )
iter	# iterations for permutation test (default = 0)
plotflag	plot fitted results (default $= 0$ )

#### **Details**

This function performs an orthogonal least-squares Procrustes analysis on 2 rectangular data matrices (X & Y) by minimizing the sum-of-squared distances between corresponding elements of the 2 matrices. This is done by translating, scaling, mirroring, and rotating Y to fit X. The symmetric orthogonal Procrustes statistic  $(m^2)$  is a measure of goodness-of-fit of Y to X after rotation and provides the residual sum-of-squares, which varies from 0 to 1. Smaller values of  $m^2$  indicate better fit.

#### Value

The function returns the following values (resid is a structure):

Xscl	centered & scaled form of X
Yrot	centered, scaled, & rotated form of Y
m2	symmetric Procrustes statistic, ranges from 0–1 (smaller values indicate better fit)
resid	structure of residuals (*.dim, *.obs, *.sse)
prob	permutation-based significance of m2

#### Note

If the # of variables (columns) in X < Y, it is padded with 0's and allows one to, say, rotate an ordination to an environmental variable.

When stnd = 1 each variable is standardized to mean 0 and variance 1 so they will contribute equal weight to the fitting process. This, however, may distort the final configurations.

An optional permutation-based significance test of  $m^2$  is performed when iter > 0 to assess the statistical concordance between X & Y.

resid.dim provides the length of each observation along each dimension after rotation. This allows interpretation of the direction of increase when Y codes for, say, an environmental gradient.

resid.obs provides the total length of each observation, which is a measure of goodness-of-fit for each observation; (smaller values = better fit).

resid.sse (like m2) is a measure of total concordance between X and Y (smaller values = better fit);

# Author(s)

Dave Jones

#### References

Cox, M. A. A. & T. F. Cox. Local minima in nonmetric multidimensional scaling. Submitted to Statistics & Computing. Available from http://www.ncl.ac.uk/mds/

Legendre, P. & L. Legendre. 1998. Numerical ecology. 2nd English ed. Elsevier Science BV, Amsterdam.

Oksanen, J. 2002. Users manual for Vegan: R functions for vegetation ecologists. Available from:

http://cc.oulu.fi/~jarioksa/softhelp/vegan.html

Peres-Neto, P. R. 2000. Documentation for program PROTEST.EXE. Dept. of Zoology, University of Toronto. Available from: http://www.zoo.utoronto.ca/jackson/software/.

Peres-Neto, P. R. & D. A. Jackson. 2001. How well do multivariate data sets match? The advantages of a Procrustean superimposition approach over the Mantel test. Oecologia 129: 169–178.

Rohlf, F. J. & D. Slice. 1990. Extensions of the Procrustes method for the optimal superimposition of landmarks. Syst. Zool. 39(1): 40–59.

#### See Also

```
f_mantel, f_bioenv
```

#### Examples

The first example follows Example 1 of Peres-Neto & Jackson (2001) and uses data from Table 1 of Losos (1990). Load the file, losos.mat, in the data folder.

[Losos, J. B. 1990. Ecomorphology, performance capability, and scaling of West Indian Anolis lizards: an evolutionary analysis. Ecol. Monogr. 60: 368–388.]

```
>> load losos.mat
% log transform variables:
>> morph_log = f_transform(morph,3);
>> perform_log = f_transform(perform,3);
% PCA on correlation matrix:
>> [morph_scores,morph_evects] = f_pca(morph_log,0,2);
>> [perform_scores,perform_evects] = f_pca(perform_log,0,2);
% scale variance of scores on each axis = 1:
>> morph_scores = f_transform(morph_scores',7)';
% Procrustes Analysis:
>> [morph_scl,perform_scl,m2,resid,prob] = f_procrustes(morph_scores(:,1:2),...
           perform_scores(:,1:2),1,1000,1);
Permuting the data 999 times...
>> prob
prob =
         0.0030
% add labels to Superimposition plot:
>> f_labelplot([morph_scl(:,1) morph_scl(:,2)],sLabels);
% directions of variation:
>> figure;
>> f_biplotPca2(morph_evects,1,0,mLabels,0);
>> f_biplotPca2(perform_evects(:,1:2)*H,1,0,pLabels,0);
>> box on;
```

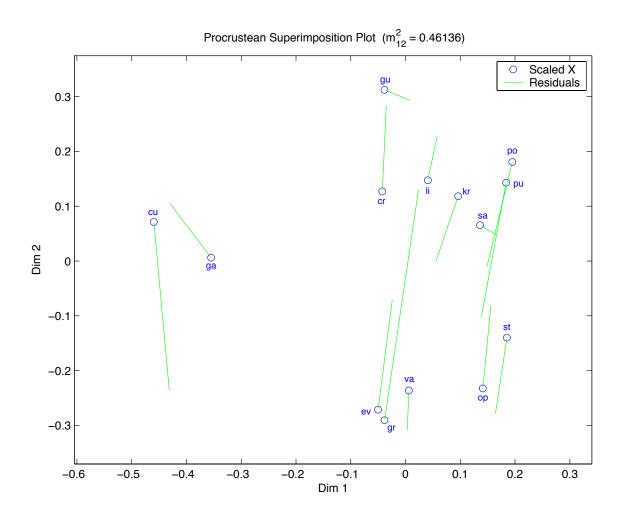


Figure 8: Example of f $\_$ procrustes

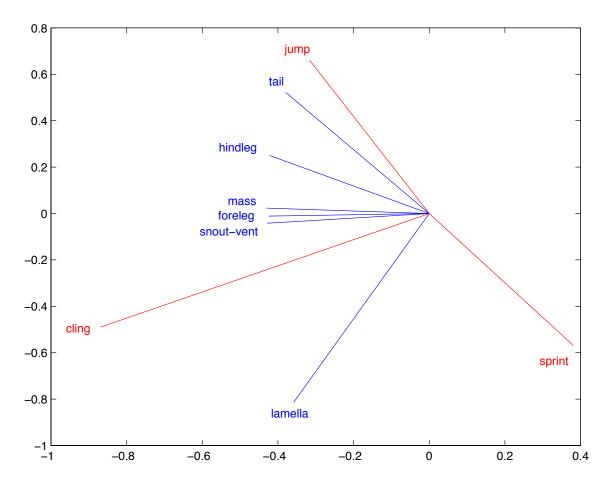


Figure 9: Example of f\_procrustes

f\_shadeBox Shade subsets of a time series plot

#### Description

This function is used to shade subsets of a time series plot in order to highlight specific time periods. It should be called after creating a time series plot and should use the same scaling factor.

## Usage

```
f_shadebox(region)
```

# Arguments

region 2-d matrix defining regions along the Y-axis to shade (column 1 = start, column 2 = stop)

scale scaling factor used in time series plot (defaut = 1)

#### **Details**

region specifies a variable currently loaded in the Matlab workspace.

'region' specifies a space-delimited ASCII file in the Matlab path.

#### Value

The function modifies the current figure.

#### Author(s)

Dave Jones

#### Examples

Load the file, shadeBox.mat, in the data folder.

```
>> load shadeBox.mat
>> plot(time,speed,'b-');
>> f_shadeBox(region);
```

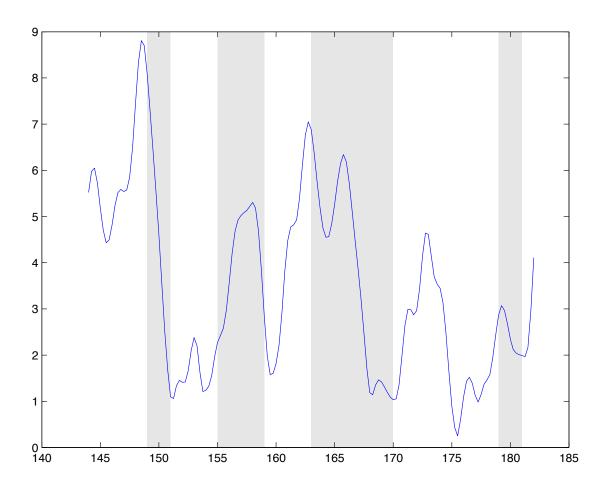


Figure 10: Example of f\_shadeBox

Moving Average Smoothing Filter

#### f\_smooth

## Description

This function is used to smooth columns of a matrix via a Moving Average. It utilizes the Matlab function smooth from the *Curve Fitting Toolbox*. Each column of data is smoothed separately but using the same span.

## Usage

```
y = f_smooth(x,span,rep);
```

## Arguments

x input matrix

span size of filter (default = 5), should be odd

rep # of times to repeat filter

## Details

Blocks of data separated by rows of NaN's are smoothed *separately*. The **smooth** function differs from similar filters, such as ones that utilize the Matlab function *filter*, as it *preserves* the endpoints of the data set.

Using the repeat option (rep) to run a smaller-sized filter multiple times may provide better results than running a larger-sized filter once.

#### Value

The function returns the following values:

y smoothed data

#### Note

This function requires the Matlab Curve Fitting Toolbox.

# Author(s)

Dave Jones

# Examples

Load the file, smooth.mat, in the data folder. This is synthetic data representing, say, the geographical coordinates of the shorelines of two islands. The coordinates of the two islands are separted from one another in the matrix dat by a row of NaN's, thus they are smoothed independently.

```
>> load smooth.mat
% smooth twice with a span of 5:
>> datsm = f_smooth(dat,5,2);
>> plot(dat(:,1),dat(:,2),'r-',datsm(:,1),datsm(:,2),'b-');
>> axis equal;box on;
>> legend('Original','Smoothed',2);
>> title('\bfSmoothing via Moving Average');
```

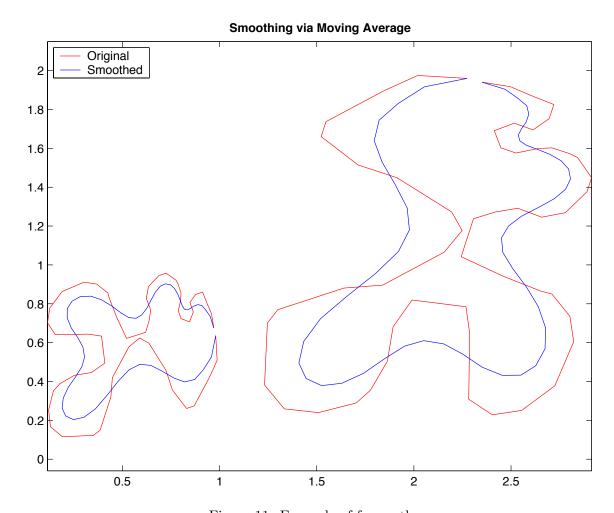


Figure 11: Example of f-smooth

Counter-clockwise angle between 2 points

f\_vecAngle

#### Description

This function is used to determine the counter-clockwise angle (in degrees) between points A and B.

## Usage

```
theta = f_vecAngle(a,b);
```

## Arguments

a 2-d row vector for point a (= [xa ya]) b 2-d row vector for point a (= [xb yb])

#### Value

The function returns the following values:

theta angle between a & b in degrees

#### Note

The points are considered to be Cartesian coordinates of the heads of vectors starting at the origin.

Note that the counter-clockwise angle from (A -> B) is not necessarily equal to that from (B -> A).

This function is vectorized, so A and B may each be 2-d matrices specifying multiple pairs of points.

# Author(s)

Dave Jones

#### References

Feldman, M. 1997. The Win95 Game Programmer's Encyclopedia. Available from:

http://www.geocities.com/SiliconValley/2151/win95gpe.html

# See Also

```
f_vecMagDir, f_vecTrans, f_vecUV
```

# Examples

```
>> a = [0 10; -10 -10];
>> b = [-10 -10; 0 10];
>> theta = f_vecAngle(a,b)
theta =
   135
   225
```

Progressive Vector Diagrams

f\_vecDiagram

# Description

This function is used to create Progressive Vector Diagrams from time series data of wind or moored current meter velocity vectors. This type of diagram is used to produce a Lagrangian display of Eulerian measurements.

# Usage

```
f_vecDiagram(u,v,units)
```

# Arguments

u,v unrotated vector components

units m/s (=1) or cm/s (=2) (default = 0)

# Value

The function creates a new figure

# Note

units is an optional parameter that allows calculation of the spatial units in the plot. A velocity vector specifying 1 m/s covers 3.6 km/hr (there is 3600 sec in an hour).

# Author(s)

Dave Jones

#### See Also

f\_vecPlot

#### Examples

Load the file vecPlot.mat from the data folder.

```
>> load vecPlot.mat
>> f_vecDiagram(u,v,1);
```

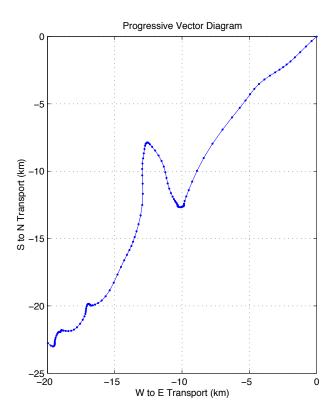


Figure 12: Example of f\_vecPlot

Magnitude & Direction from U, V vector components

f\_vecMagDir

# Description

This function is used to obtain Polar coordinates (magnitude & direction) of a vector given its Cartesian coordinates (U & V vector components). The direction is the counter-clockwise angle of rotation.

# Usage

```
[mag,dir] = f_vecMagDir(u,v);
```

#### Arguments

u,v

column vectors of Cartesian coordinates of heads of vector components

#### **Details**

The programs uses the Matlab function ATAN2 which relies on the sign of both input arguments to determine the quadrant of the result.

#### Value

The function returns the following values:

mag length of vector

dir angle of rotation (in degrees between 0–360)

# Author(s)

Dave Jones

#### See Also

```
f_vecAngle, f_vecTrans, f_vecUV,
```

# Examples

f\_vecPlot Plot time series of velocity vectors

# Description

This function is used to plot time series of wind or current meter velocity vectors using Matlab's quiver function.

#### Usage

```
f_vecPlot(jdate,u,v,scale,units,jRange);
```

#### Arguments

jdate	column vector of Julian dates
u,v	corresponding vector components
scale	scale factor (default $= 1$ )
units	Y-axis label; e.g., units = 'm/s' (default = none)
jRange	limits of dates to plot; e.g., jRange = [min max] (default = auto)

#### Value

The function produces a new figure;

# Note

This function is necessary in order to obtain vectors that have the proper length and angle of rotation. An optional scaling factor can be applied allowing the user control over the amount of overlap among vectors and/or the scaling of vectors relative to the overall time series. The X-axis is scaled accordingly. The Y-axis allows easy, visual interpretation of vector length.

U,V components of velocity vectors can be extracted from data specifying only Speed and Direction using f\_vecUV.

### Author(s)

Dave Jones

#### See Also

```
f_julian, f_vecUV, f_shadeBox
```

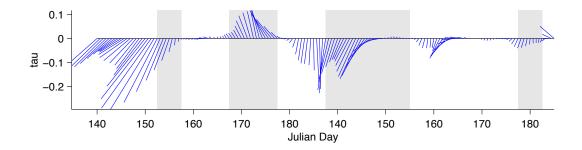


Figure 13: Example of f\_vecPlot

# Examples

Load the file vecPlot.mat from the data folder.

```
>> load vecPlot.mat
>> f_vecPlot(date,u,v,20,'tau',[140 182]);
>> axis tight;
>> f_shadeBox(subsets,20);
```

 $f\_vecTrans$  Transform 2-d vector coordinates

# Description

This function is used to rotate, translate, and/or scale 2dimensional Cartesian coordinates. If the input coordinates specify the heads of (velocity) vectors, they may additionally be converted to unit length.

# Usage

```
[tx,ty] = f_vecTrans(x,y,rot,transl,scale,unit);
```

# Arguments

x,y	column vectors specifying coordinate pairs
rot	angle of rotation in degrees (default $= 0$ )
transl	translation [dx dy] $(default = [0 \ 0])$
scale	scaling factor [sx sy] (default = $[1 \ 1]$ )
unit	convert vectors to unit length (default $= 0$ )

### **Details**

This function uses the following matrices for transforming coordinates:

Rotation matrix: ( $\theta$  is in radians):

$$\begin{pmatrix}
\cos(\theta) & -\sin(\theta) & 0 \\
\sin(\theta) & \cos(\theta) & 0 \\
0 & 0 & 1
\end{pmatrix}$$

Translation matrix:

$$\left(\begin{array}{ccc}
1 & 0 & tx \\
0 & 1 & ty \\
0 & 0 & 1
\end{array}\right)$$

Scaling matrix:

$$\left(\begin{array}{ccc} sx & 0 & 0\\ 0 & sy & 0\\ 0 & 0 & 1 \end{array}\right)$$

#### Value

The function returns the following values:

```
tx,ty transformed coordinates
```

#### Note

X and Y coordinates can be translated or scaled asymmetrically if 2 values are specified for these parameters. If only 1 value is provided, coordinates are translated or scaled symmetrically.

# Author(s)

Dave Jones

#### References

Feldman, M. 1997. The Win95 Game Programmer's Encyclopedia. Available from:

```
http://www.geocities.com/SiliconValley/2151/win95gpe.html
```

#### See Also

```
f_vecAngle, f_vecMagDir, f_vecTrans3d, f_vecUV
```

### Examples

# Description

This function is used to rotate, translate, and/or scale 3 dimensional Cartesian coordinates. If the input coordinates specify the heads of (velocity) vectors, they may additionally be converted to unit length.

## Usage

```
[tx,ty,tz] = f_vecTrans3d(x,y,z,xrot,yrot,zrot,transl,scale,unit);
```

# Arguments

x,y,z	column vectors specifying coordinate triplets
xrot	rotation about X-axis in degrees (default = $0$ )
yrot	rotation about Y-axis in degrees (default $= 0$ )
zrot	rotation about Z-axis in degrees (default = $0$ )
transl	translation [dx dy dz] (default = $[0 \ 0 \ 0]$ )
scale	scaling factor [sx sy sz] (default = $[1 \ 1 \ 1]$ )
unit	convert vectors to unit length (default $= 0$ )

# Details

This function uses the following matrices for transforming coordinates:

Rotation Matrices ( $\theta$  is in radians):

X-axis:

$$\begin{pmatrix}
1 & 0 & 0 & 0 \\
0 & \cos(\theta) & -\sin(\theta) & 0 \\
0 & \sin(\theta) & \cos(\theta) & 0 \\
0 & 0 & 0 & 1
\end{pmatrix}$$

*Y-axis:* 

$$\begin{pmatrix}
\cos(\theta) & 0 & \sin(\theta) & 0 \\
0 & 1 & 0 & 0 \\
-\sin(\theta) & 0 & \cos(\theta) & 0 \\
0 & 0 & 0 & 1
\end{pmatrix}$$

Z-axis:

$$\begin{pmatrix}
\cos(\theta) & -\sin(\theta) & 0 & 0 \\
\sin(\theta) & \cos(\theta) & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 1
\end{pmatrix}$$

Translation matrix:

$$\left(\begin{array}{cccc}
1 & 0 & 0 & tx \\
0 & 1 & 0 & ty \\
0 & 0 & 1 & tz \\
0 & 0 & 0 & 1
\end{array}\right)$$

Scaling matrix:

$$\left(\begin{array}{cccc}
sx & 0 & 0 & 0 \\
0 & sy & 0 & 0 \\
0 & 0 & sz & 0 \\
0 & 0 & 0 & 1
\end{array}\right)$$

#### Value

The function returns the following values:

tx,ty transformed coordinates

#### Note

X, Y, & Z coordinates can be translated or scaled asymmetrically if 3 values are specified for these parameters. If only 1 value is provided, coordinates are translated or scaled symmetrically.

# Author(s)

Dave Jones

# References

Feldman, M. 1997. The Win95 Game Programmer's Encyclopedia. Available from:

http://www.geocities.com/SiliconValley/2151/win95gpe.html

# See Also

f\_vecAngle, f\_vecMagDir, f\_vecTrans, f\_vecUV

U, V vector components from Magnitude  $\mathcal{C}$  Direction

f\_vecUV

# Description

This function is used to obtain Cartesian coordinates (the U,V vector components) of a vector given its Polar coordinates (Magnitude & Direction).

## Usage

```
[u,v] = f_vecUV(mag,dir);
```

# Arguments

mag	column vector specifying magnitude of vectors (in arbitrary units)
dir	column vector indicating angle of rotation (in degrees from 0–360)

## **Details**

This function is used to obtain Cartesian coordinates (the U,V vector components) of a vector given its Polar coordinates (magnitude & direction). Direction is the counterclockwise angle of rotation.

# Value

The function returns the following values:

u, v Cartesian coordinates of heads of vector components

# Author(s)

Dave Jones

## See Also

```
f_vecAngle, f_vecMagDir, f_vecTrans
```

# Examples

# Description

Here are a number of ANOVA Examples using f\_npManova. These are provided to help the user become familiar with the syntax used for running the function and for coding of ANOVA factors. Note that f\_npManova uses a GLM approach to ANOVA and relies on the *unrestricted* model for constructing proper F-ratios, so output may differ from that found in textbook examples, which typically use the restricted model with balanced designs. The GLM approach used here should provide output similar to that given by SAS and MINITAB.

Remember that the true power of this function is utilized when the response variable consists of a symmetric distance matrix based one of the metrics commonly used in community ecology (e.g., Bray-Curtis, etc.).

# References

K. A. Brownlee. 1960. Statistical theory and methodology in science and engineering, John Wiley & Sons, Inc., New York.

Neter, J., M. H. Kutner, C. J. Nachtsheim, & W. Wasserman. 1996. Applied linear statistical models, 4th Edition, Richard D. Irwin, Inc., Burr Ridge, Illinois.

Sokal, R. R. & F. J. Rohlf. 1995. Biometry — The principles and practice of statistics in biological research. 3rd ed. W. H. Freeman, New York. xix + 887 pp.

Zar, J. H. 1999. Biostatistical analysis. 4th ed. Prentice Hall, Upper Saddle River, NJ.

Winer, B. J. 1971. Statistical Principles in Experimental Design, Second Edition, McGraw-Hill.

# Examples

1. For an example of a Two-Way Model I ANOVA with replication, load the file, sr\_11p1.mat, in the data folder. This data is from Table 11.1 of Sokal & Rohlf (1995) and has 1 response variable, food, and 2 fixed factors: fat and sex.

```
>> load sr_11p1.mat
>> dis = f_euclid(food');
>> result = f_npManova(dis,[sex fat],0,0,1000,1);
```

\_\_\_\_\_

# Please specify the ANOVA model for 2-way ANOVA factors 1 & 2:

```
1 & 2 are fixed [21]
```

1 is fixed or random, 2 is nested [23]

Select model...[0 will cancel]
21

Permuting the data 999 times...

### 

# Nonparametric (Permutation-based) MANOVA:

'Source'	'df'	'SS'	'MS'	'F'	'n,
'factor 1'	[ 1]	[3780.8]	[3780.8]	[2.5925]	[0.139]
'factor 2'	[ 1]	[ 61204]	[ 61204]	[41.969]	[0.001]
'factor 1x2'	[ 1]	[918.75]	[918.75]	[ 0.63]	[0.446]
'residual'	[8]	[ 11667]	[1458.3]	[ NaN]	[ NaN]
'total'	[11]	[ 77570]	[ NaN]	[ NaN]	[ NaN]

# iterations = 1000

(Note: NaNs are placeholders for the ANOVA table)

(Data with replication)

- 2. For an example of a Two-Way Mixed-Model ANOVA without replication, load the file, sr\_11p3.mat, in the data folder. This data is from Box 11.3 of Sokal & Rohlf (1995) and has 1 response variable, temp, and 2 factors: depth (fixed), and day (random).
  - >> load sr\_11p3.mat
  - >> dis = {f\_euclid}(temp');
  - >> result = f\_npManova(dis,[depth day],0,0,1000,1);

\_\_\_\_\_

Please specify the ANOVA model for 2-way ANOVA factors 1 & 2:

·

- 1 & 2 are fixed [21]
- 1 is fixed or random, 2 is random [22]
- 1 is fixed or random, 2 is nested [23]

Select model...[O will cancel]

22

<sup>1</sup> is fixed or random, 2 is random [22]

Permuting the data 999 times...

Nonparametric	(Permu	tation-based)	MANOVA:		
'Source'	df'	'SS'	'MS'	'F'	'p'
'factor 1'	[ 9]	[2119.7]	[ 235.52]	[ 2835]	[0.001]
'factor 2'	[ 3]	[ 0.562]	[ 0.18733]	[2.255]	[0.134]
'residual'	[27]	[ 2.243]	[0.083074]	[ NaN]	[ NaN]
'total'	[39]	[2122.5]	[ NaN]	[ NaN]	[ NaN]

-----

(Note: NaNs are placeholders for the ANOVA table)

(Data has NO replication)

3. For an example of a *Two-Way Model II Nested ANOVA (balanced)*, load the file, sr\_10p1.mat, in the data folder. This data is from Box 10.1 of Sokal & Rohlf (1995) and has 1 response variable, wing, and 2 factors: cage (random), and female (nested).

Please specify the ANOVA model for 2-way ANOVA factors 1 & 2:

.\_\_\_\_\_

```
1 & 2 are fixed [21]
1 is fixed or random, 2 is random [22]
1 is fixed or random, 2 is nested [23]
```

Select model...[O will cancel] 23

Permuting the data 999 times...

-----

# Nonparametric (Permutation-based) MANOVA: 'Source' 'df' 'SS' 'MS' 'F' 'factor 1' [ 2] [665.68] [332.84] [1.7409]

[0.252][ 9] [0.001]'factor 2' [1720.7][191.19] [146.88] 'residual' [12] [1.3017] [ NaN] [ 15.62] NaN] 'total' [23] [ 2402] [ NaN] NaN[ NaN]

'n,

```
# iterations = 1000
```

(Note: NaNs are placeholders for the ANOVA table)

(Data with replication)

4. For an example of a Two-Way Model II Nested ANOVA (unbalanced), load the file, sr\_10p6.mat, in the data folder. This data is from Box 10.6 of Sokal & Rohlf (1995) and has 1 response variable, ph, and 2 factors: dam (random), and sire (nested).

```
>> load sr_10p6.mat
>> dis = f_euclid(ph');
>> result = f_npManova(dis,[dam sire],0,0,1000,1);
```

-----

Please specify the ANOVA model for 2-way ANOVA factors 1 & 2:

\_\_\_\_\_\_

1	& 2 are fixed		[21]
1	is fixed or random	, 2 is random	[22]
1	is fixed or random.	2 is nested	[23]

Select model...[O will cancel] 23

Permuting the data 999 times...

# Nonparametric (Permutation-based) MANOVA:

'Source'	'df'	'SS'	'MS'	'F'	'n,
factor 1'	[ 14]	[1780.2]	[127.16]	[3.5383]	[0.003]
factor 2'	[ 22]	[ 790.6]	[35.937]	[1.4482]	[0.107]
residual'	[123]	[3052.2]	[24.814]	[ NaN]	[ NaN]
'total'	[159]	[5622.9]	[ NaN]	[ NaN]	[ NaN]

# iterations = 1000

(Note: NaNs are placeholders for the ANOVA table)

(Data with replication)

5. For an example of a Two-Way Model III ANOVA with no replication, load the file, zar\_12p4.mat, in the data folder. This data is from Example 12.4 of Zar (1999) and has 1 response variable, gain, and 2 random factors: diet, and block. This is also know as a randomized block design.

```
>> load zar_12p4.mat
>> dis = f_euclid(gain');
>> result = f_npManova(dis,[diet block],0,0,1000,1);
_____
  Please specify the ANOVA model
  for 2-way ANOVA factors 1 & 2:
_____
1 & 2 are fixed
                                     [21]
1 is fixed or random, 2 is random [22]
1 is fixed or random, 2 is nested [23]
Select model...[O will cancel]
22
Permuting the data 999 times...
_____
    Nonparametric (Permutation-based) MANOVA:
    'Source' 'df' 'SS' 'MS' 'F' 'p'
'factor 1' [ 3] [27.425] [ 9.1418] [11.825] [0.004]
'factor 2' [ 4] [62.647] [ 15.662] [20.259] [0.001]
'residual' [12] [ 9.277] [0.77308] [ NaN] [ NaN]
'total' [19] [ 99.35] [ NaN] [ NaN] [ NaN]
      # iterations = 1000
```

(Note: NaNs are placeholders for the ANOVA table)

(Data has NO replication)

6. For an example of a Three-Way Model I ANOVA with replication, load the file, zar\_14p1.mat, in the data folder. This data is from Example 14.1 of Zar (1999) and has 1 response variable, rate, and 3 fixed factors: species, temp, and sex.

```
1 & 2 fixed, 3 nested in 1 (Cross-Nested) [34]
1 &/or 2 random, 3 nested in 1 (Cross-Nested) [35]
3 nested in 2 nested in 1 (Fully Nested) [36]
```

Select model...[O will cancel]
31

Permuting the data 999 times...

### \_\_\_\_\_\_

Nonparametric	(Permutation-based)	MANUVA:

'Source'	'df'	'SS'	'MS'	'F'	'p'
'factor 1'	[ 2]	[ 1.8175]	[ 0.90875]	[24.475]	[0.001]
'factor 2'	[ 2]	[ 24.656]	[ 12.328]	[332.02]	[0.001]
'factor 3'	[ 1]	[0.0088889]	[0.0088889]	[0.2394]	[0.626]
'factor 1x2'	[ 4]	[ 1.1017]	[ 0.27542]	[7.4177]	[0.002]
'factor 1x3'	[ 2]	[ 0.37028]	[ 0.18514]	[4.9863]	[0.013]
'factor 2x3'	[ 2]	[ 0.17528]	[ 0.087639]	[2.3603]	[0.114]
'factor 1x2x3'	[ 4]	[ 0.22056]	[ 0.055139]	[ 1.485]	[0.233]
'residual'	[54]	[ 2.005]	[ 0.03713]	[ NaN]	[ NaN]
'total'	[71]	[ 30.355]	[ NaN]	[ NaN]	[ NaN]

# iterations = 1000

-----

(Note: NaNs are placeholders for the ANOVA table)

(Data with replication)

- 7. For an example of a *Three-Way Model I ANOVA with no replication*, load the file, sr\_12p1.mat, in the data folder. This data is from Box 12.1 of Zar (1999) and has 1 response variable, time, and 3 fixed factors: temp, cn, and o2.
  - >> load zar\_14p1.mat
  - >> dis = f\_euclid(rate');
  - >> result = f\_npManova(dis,[species temp sex],0,0,1000,1);

-----

Please specify the ANOVA model for 3-way ANOVA factors 1, 2, & 3:

All factors fixed [31]
1 & 2 are fixed, 3 is random [32]

1 & 2 are fixed, 3 is random [32] 1 is fixed or random, 2 & 3 are random [33]

1 & 2 fixed, 3 nested in 1 (Cross-Nested) [34] 1 &/or 2 random, 3 nested in 1 (Cross-Nested) [35] 3 nested in 2 nested in 1 (Fully Nested) [36]

```
Select model...[O will cancel]
31
```

Permuting the data 999 times...

#### -----

'Source'	df'	'SS'	'MS'	'F'	'n,
factor 1'	[ 2]	[ 1.8175]	[ 0.90875]	[24.475]	[0.001
factor 2'	[ 2]	[ 24.656]	[ 12.328]	[332.02]	[0.001
factor 3'	[ 1]	[0.0088889]	[0.0088889]	[0.2394]	[0.612
'factor 1x2'	[ 4]	[ 1.1017]	[ 0.27542]	[7.4177]	[0.001
'factor 1x3'	[ 2]	[ 0.37028]	[ 0.18514]	[4.9863]	[0.012
'factor 2x3'	[ 2]	[ 0.17528]	[ 0.087639]	[2.3603]	[0.093
'factor 1x2x3'	[ 4]	[ 0.22056]	[ 0.055139]	[ 1.485]	[0.221
'residual'	[54]	[ 2.005]	[ 0.03713]	[ NaN]	[ NaN
'total'	[71]	[ 30.355]	[ NaN]	[ NaN]	[ Nal

# iterations = 1000

(Note: NaNs are placeholders for the ANOVA table)

(Data with replication)

- 8. For an example of a *Three-Way Model II ANOVA with replication*, load the file, thick.mat, in the data folder. This data ships with MINITAB and has 1 response variable, thickness, 2 fixed factors: time and setting, and 1 random factor: operator.
  - >> load thick.mat
  - >> dis = f\_euclid(thickness');
  - >> result = f\_npManova(dis,[time setting operator],0,0,1000,1);

-----

Please specify the ANOVA model for 3-way ANOVA factors 1, 2, & 3:

All factors	fixed		[31]

1 & 2 are fixed, 3 is random [32]
1 is fixed or random, 2 & 3 are random [33]

```
1 & 2 fixed, 3 nested in 1 (Cross-Nested) [34]
1 &/or 2 random, 3 nested in 1 (Cross-Nested) [35]
3 nested in 2 nested in 1 (Fully Nested) [36]
```

Select model...[O will cancel]

Permuting the data 999 times...

'Source'	'df'	'SS'	'MS'	'F'	,p,
'factor 1'	[ 1]	[ 9]	[ 9]	[0.29032]	[0.
'factor 2'	[ 2]	[ 15676]	[7838.2]	[ 73.178]	[0.
'factor 3'	[ 2]	[1120.9]	[560.44]	[ 4.9114]	[0.
'factor 1x2'	[2]	[ 114.5]	[ 57.25]	[ 2.3854]	[0.
'factor 1x3'	[ 2]	[ 62]	[ 31]	[ 1.2917]	[0.
'factor 2x3'	[ 4]	[428.44]	[107.11]	[ 4.463]	[0.
'factor 1x2x3'	[ 4]	[ 96]	[ 24]	[ 7.082]	[0.
'residual'	[18]	[ 61]	[3.3889]	[ NaN]	[ ]
'total'	[35]	[ 17568]	[ NaN]	[ NaN]	[ ]

(Note: NaNs are placeholders for the ANOVA table)

(Data with replication)

- 9. For an example of a Three-Way Model II ANOVA with no replication, load the file, milk.mat, in the data folder. This is from Brownlee (1960:p.516) and has 1 response variable, counts, 2 fixed factors: bottle and tube, and 1 random factor: sample.
  - >> load milk.mat
  - >> dis = f\_euclid(counts');
  - >> result = f\_npManova(dis,[bottle tube sample],0,0,1000,1);

Please specify the ANOVA model for 3-way ANOVA factors 1, 2, & 3:

All factors fixe 1 & 2 are fixed, 1 is fixed or ra		1	[31] [32] [33]
1 & 2 fixed	3 nested in 1	(Cross-Nested)	[34]

```
(Cross-Nested)
1 &/or 2 random, 3 nested in 1
                                    (Cross-Nested) [35]
3 nested in 2 nested in 1
                                     (Fully Nested) [36]
```

Select model...[0 will cancel]

32

Permuting the data 999 times...

'Source'	'df'	'SS'	'MS'	'F'	'n,
factor 1'	[ 1]	[0.34722]	[0.34722]	[0.14066]	[0.736]
factor 2'	[ 2]	[ 15.361]	[ 7.6806]	[ 7.6903]	[0.003]
factor 3'	[11]	[ 93.486]	[ 8.4987]	[ 3.6208]	[ 0.08]
factor 1x2'	[ 2]	[ 1.3611]	[0.68056]	[0.60767]	[0.572]
factor 1x3'	[11]	[ 27.153]	[ 2.4684]	[ 2.2041]	[0.048]
factor 2x3'	[22]	[ 21.972]	[0.99874]	[0.89177]	[0.616]
residual'	[22]	[ 24.639]	[ 1.1199]	[ NaN]	[ NaN]
'total'	[71]	[ 184.32]	[ NaN]	[ NaN]	[ NaN]

(Note: NaNs are placeholders for the ANOVA table)

(Data has NO replication)

10. For an example of a *Three-Way Model III ANOVA with replication*, load the file, exercise.mat, in the data folder. This data is from Table 23.4 of Neter et al. (1996) and has 1 response variable, tol, and 3 random factors: gender, fat, and smoke.

```
>> load exercise.mat
>> dis = f_euclid(tol');
>> result = f_npManova(dis,[gender fat smoke],0,0,1000,1);
_____
 Please specify the ANOVA model
 for 3-way ANOVA factors 1, 2, & 3:
 _____
All factors fixed
                                         [31]
1 & 2 are fixed, 3 is random
                                         [32]
1 is fixed or random, 2 & 3 are random
                                         [33]
1 & 2 fixed,
            3 nested in 1
                           (Cross-Nested)
                                         [34]
```

1 &/or 2 random, 3 nested in 1 (Cross-Nested) [35] 3 nested in 2 nested in 1 (Fully Nested) [36]

Select model...[O will cancel]

Permuting the data 999 times...

Nonparametric	(Permutation-based)	MANOVA:
---------------	---------------------	---------

'Source'	df'	'SS'	'MS'	'F'	'n,
'factor 1'	[ 1]	[176.58]	[176.58]	[ 7.7278]	[0.037]
'factor 2'	[ 1]	[242.57]	[242.57]	[ 2.8797]	[0.084]
'factor 3'	[ 1]	[70.384]	[70.384]	[0.86198]	[0.187]
'factor 1x2'	[ 1]	[ 13.65]	[ 13.65]	[ 7.2981]	[ 0.2]
'factor 1x3'	[ 1]	[ 11.07]	[ 11.07]	[ 5.9187]	[0.233]
'factor 2x3'	[ 1]	[72.454]	[72.454]	[ 38.737]	[0.089]
'factor 1x2x3'	[ 1]	[1.8704]	[1.8704]	[0.20036]	[0.694]
'residual'	[16]	[149.37]	[9.3354]	[ NaN]	[ NaN]
'total'	[23]	[737.95]	[ NaN]	[ NaN]	[ NaN]

(Note: NaNs are placeholders for the ANOVA table)

(Data with replication)

- 11. For an example of a *Three-Way Model III ANOVA with no replication*, load the file, milk.mat, in the data folder. This is from Brownlee (1960:p.516) and has 1 response variable, counts and 3 random factors: bottle, tube, and sample.
  - >> load milk.mat
  - >> dis = f\_euclid(counts');
  - >> result = f\_npManova(dis,[bottle tube sample],0,0,1000,1);

\_\_\_\_\_

Please specify the ANOVA model for 3-way ANOVA factors 1, 2, & 3:

All factors fixed 1 & 2 are fixed, 3 is 1 is fixed or random		m	[31] [32] [33]
1 & 2 fixed, 3 no 1 &/or 2 random, 3 no 3 nested in 2 nested	ested in 1	(Cross-Nested) (Cross-Nested) (Fully Nested)	[35]

Select model...[O will cancel] 33

Permuting the data 999 times...

Nonparametric (Permutation-based) MANOVA:

'Source' 'df' 'SS' 'MS' 'F'

'n,

```
'factor 1'
                 [ 1]
                         [0.34722]
                                       [0.34722]
                                                     [0.17113]
                                                                   [0.555]
'factor 2'
                 [ 2]
                         [ 15.361]
                                       [7.6806]
                                                     [ 13.731]
                                                                   [0.025]
'factor 3'
                [11]
                         [ 93.486]
                                       [ 8.4987]
                                                     [ 3.6208]
                                                                  [0.062]
'factor 1x2'
                [ 2]
                         [ 1.3611]
                                       [0.68056]
                                                     [0.60767]
                                                                  [0.558]
'factor 1x3'
                [11]
                        [ 27.153]
                                       [2.4684]
                                                     [ 2.2041]
                                                                  [0.06]
'factor 2x3'
                [22]
                         [ 21.972]
                                       [0.99874]
                                                     [0.89177]
                                                                  [0.612]
'residual'
                 [22]
                         [ 24.639]
                                       [ 1.1199]
                                                                  [ NaN]
                                                     Γ
                                                          NaN
'total'
                 [71]
                         [ 184.32]
                                                          NaNl
                                                                   [ NaN]
                                            NaN
```

(Note: NaNs are placeholders for the ANOVA table)

(Data has NO replication)

12. For an example of a *Three-Way, Fully Nested, Mixed Model ANOVA (balanced)*, load the file, sr\_10p5.mat, in the data folder. This is from Box 10.5 of Sokal Rohlf (1995) has 1 response variable, gly, 1 fixed factor, rx and 2 nested factors: rat (nested in rx) and prep (nested in rat).

```
>> load sr_10p5.mat
>> dis = f_euclid(gly');
>> result = f_npManova(dis,[rx rat prep],0,0,5000,1);
```

Please specify the ANOVA model

for 3-way ANOVA factors 1, 2, & 3:

All factors fixed	[31]
1 & 2 are fixed, 3 is random	[32]
1 is fixed or random, 2 & 3 are random	[33]

1	& 2 fixed,	3 nested in	1	(Cross-Nested)	[34]
1	<pre>&amp;/or 2 random,</pre>	3 nested in	1	(Cross-Nested)	[35]
3	nested in 2 nes	sted in 1		(Fully Nested)	[36]

Select model...[O will cancel] 36

Permuting the data 4999 times...

\_\_\_\_\_\_

# Nonparametric (Permutation-based) MANOVA:

'Source'	'df'	'SS'	'MS'	'F'	'p'
'factor 1'	[ 2]	[1557.6]	[778.78]	[ 2.929]	[ 0.192]
'factor 2'	[ 3]	[797.67]	[265.89]	[5.3715]	[0.0148]
'factor 3'	[12]	[ 594]	[ 49.5]	[2.3386]	[0.0484]

```
'residual' [18] [ 381] [21.167] [ NaN] [ NaN] 'total' [35] [3330.2] [ NaN] [ NaN] [ NaN]
```

-----

(Note: NaNs are placeholders for the ANOVA table)

(Data with replication)

- 13. For an example of a *Three-Way Cross-Nested ANOVA with replication*, load the file, steel.mat, in the data folder. This is from Brownlee (1960:p.530) and has 1 response variable, qual, 2 fixed factors: ann and loc, and 1 nested factor: coil (nested in ann. This design is also know as *partially nested* or *partially hierarchical*.
  - >> load steel.mat
  - >> dis = f\_euclid(qual');
  - >> result = f\_npManova(dis,[ann loc coil],0,0,1000,1);

Please specify the ANOVA model for 3-way ANOVA factors 1, 2, & 3:

All factors fixed [31]
1 & 2 are fixed, 3 is random [32]

[33]

1 is fixed or random, 2 & 3 are random

1 & 2 fixed, 3 nested in 1 (Cross-Nested) [34] 1 &/or 2 random, 3 nested in 1 (Cross-Nested) [35] 3 nested in 2 nested in 1 (Fully Nested) [36]

Select model...[0 will cancel]
24

Permuting the data 999 times...

\_\_\_\_\_

# Nonparametric (Permutation-based) MANOVA:

'Source'	'df'	'SS'	'MS'	'F'	'n,
'factor 1'	[ 1]	[ 2646]	[ 2646]	[ 1.091]	[0.346]
'factor 2'	[ 1]	[1872.7]	[1872.7]	[ 35.389]	[0.004]
'factor 3'	[ 4]	[9701.3]	[2425.3]	[ 45.833]	[0.001]
'factor 1x2'	[ 1]	[16.667]	[16.667]	[0.31496]	[0.606]
'factor 2x3'	[ 4]	[211.67]	[52.917]	[0.50039]	[0.718]
'residual'	[12]	[ 1269]	[105.75]	[ NaN]	[ NaN]
'total'	[23]	[ 15717]	[ NaN]	[ NaN]	[ NaN]

(Note: NaNs are placeholders for the ANOVA table)

(Data with replication)

F-ratios used by f\_npManova

Appendix B

# Description

Here are the F-ratios used in the various ANOVA designs supported by f\_npManova.

# References

Zar, J. H. 1999. Biostatistical analysis. 4th ed. Prentice Hall, Upper Saddle River, NJ.

This program uses a GLM-approach to (M)ANOVA, so for mixed models (when there are both fixed and random factors) an unrestricted model is used.

## Two-way ANOVA's:

A = fixed, B = fixed: [f npManova2]

Source F-ratio
A MSA/MSerror
B MSB/MSerror
AB MSAB/MSerror

A = fixed, B = random:

A = random, B = random: [f npManova2]

Source F-ratio
A MSA/MSAB
B MSB/MSAB
AB MSAB/MSerror

# Two-way Nested ANOVA's:

A = fixed, B = nested in A A = random, B = nested in A: [f\_npManova2n]

Source F-ratio
A MSa/MSB
B MSB/MSerror

# Three-way ANOVA's:

=======

A = fixed, B = fixed, C = fixed: [f npManova3]

Source
A

B

MSA/MSerror
B

MSB/MSerror
C

MSc/MSerror
AB

MSAB/MSerror
AC

MSAB/MSerror
AC

MSAC/MSerror
BC

MSBC/MSerror
ABC

MSABC/MSerror

A = fixed, B = fixed, C = random:  $[f_npManova3]$ 

Source F-ratio
A MS<sub>A</sub>/MS<sub>AC</sub>
B MS<sub>B</sub>/MS<sub>BC</sub>

C  $MS_c/(MS_{AC} + MS_{BC} - MS_{ABC})$ 

AB MS<sub>AB</sub>/MS<sub>ABC</sub>

AC MSac/MSabc BC MSbc/MSabc ABC MSabc/MSerror

A = fixed, B = random, C = random:

 $A = random, B = random, C = random: [f_npManova3]$ 

Source F-ratio

A MS<sub>A</sub>/ (MS<sub>AB</sub> + MS<sub>AC</sub> - MS<sub>ABC</sub>)
B MS<sub>B</sub>/ (MS<sub>AB</sub> + MS<sub>BC</sub> - MS<sub>ABC</sub>)
C MS<sub>C</sub>/ (MS<sub>AC</sub> + MS<sub>BC</sub> - MS<sub>ABC</sub>)

AB MSAB/MSABC AC MSAC/MSABC BC MSBC/MSABC ABC MSABC/MSerror

# Three-way Cross-Nested ANOVA's:

A = fixed, B = fixed, C = nested in A: [f npManova3Nest1]

Source F-ratio
A MSA/MSC
B MSB/MSBC
C MSC/MSBC
AB MSAB/MSBC
BC MSBC/MSBC

A = fixed, B = random, C = nested in A A = random, B = fixed, C = nested in A

A = fixed, B = fixed, C = nested in A:  $[f_npManova3Nest1]$ 

Source F-ratio

 $\overline{A}$   $\overline{MS_A/MS_C}$  +  $\overline{MS_{AB}}$  -  $\overline{MS_{BC}}$ 

B MS<sub>B</sub>/MS<sub>AB</sub>
C MS<sub>C</sub>/MS<sub>BC</sub>
AB MS<sub>AB</sub>/MS<sub>BC</sub>
BC MS<sub>BC</sub>/MS<sub>error</sub>

# Three-way Fully-Nested ANOVA's:

A = fixed, B = nested in A, C = nested in B:

A = random, B = nested in A, C = nested in B:  $[f_npManova3Nest2]$ 

Source F-ratio
A MS<sub>A</sub>/MS<sub>B</sub>
B MS<sub>B</sub>/MS<sub>C</sub>
C MS<sub>C</sub>/MS<sub>error</sub>