Summary of Methods.	Inputs.	Outputs.	and Assumptions.
Summary of moundabl	mpaco,	Carpais,	and i issumptions.

Inputs	Outputs	Assumptions
-		Numerical data, independent vectors, ideally MVN with common cov matrices
g Samples of data vectors, some variables designated as response, others designated as covariate	ors, some variables gnated as response, rs designated as	
n observed data vectors, some variables designated as response, some designated as predictor	Estimated regression coefficients, tests for effects, partial correlations	Linear effects of the X's on the Y's, independent and identically distributed MVN residual vectors
Covariance or correlation matrix	PC scores that maximize variance	Reliable estimate of covariance matrix
	Estimated FA model	Reliable estimate of cov matrix, Existence of factors, Sigma = $LL^{+}\psi$, and (ideally) MVN
	" CFA "	Same as above, except Sigma = (structure determined by the CFA model) instead of LL` + ψ .
	g Samples of Data vectors g Samples of data vectors, some variables designated as response, others designated as covariate n observed data vectors, some variables designated as response, some designated as predictor Covariance or correlation matrix	g Samples of Data Tests for equality of multivariate means, MSLC, multiple comparisons g Samples of data vectors, some variables designated as response, others designated as covariate Tests for covariate-adjusted equality of multivariate means, MSLC, multiple comparisons n observed data vectors, some variables designated as response, others designated as response, some designated as predictor Estimated regression coefficients, tests for effects, partial correlations Covariance or correlation matrix PC scores that maximize variance " " " " Estimated FA model

Structural equations models		Same as CFA, except with directional paths relating latent factors instead of bi-directional correlations	Same as CFA, in addition, linearity and correct causal directional determinations of the paths.
Canonical Correlation	" " " ", with pre-defined groups of variables.	Scores that maximize correlation between groups	Reliable estimate of correlation matrix, MVN ideally needed for tests of hypotheses
Discriminant Analysis	See MANOVA input	A classification rule to assign observations to groups, estimated error probabilities	Reliable estimates of the distributions, reliable assessments of costs and priors.
Canonical Discriminant Analysis		Linear combinations that maximally separate groups	MANOVA assumptions for "optimality", although there nothing wrong with using the method as a descriptive tool.
Hierarchical Cluster Analysis	A distance matrix and a choice of clustering algorithm	Groupings of items from the distance matrix, measures of separation of the groups, tree diagram	A reasonable distance matrix
		Groupings of the observations that are rows of the matrix	Clusters have a constant radius
Multidimensional A distance matrix and choice of scaling method v		Scores (ideally two for plotting) for each item from the distance matrix, and a measure of how well the distances between scores match the distances in the distance	A reasonable distance matrix

		matrix	
Correspondence Analysis	An RxC contingency table (from n bivariate categorical data pairs)	Scores (ideally two for plotting) for each row item and column item.	Sample size large enough so that cell proportions are reliably estimated
Biplots	An (nxp) numerical data matrix	A plot of (PC1, PC2) data for all n observations, as well as variable directions as determined by the first two eigenvectors.	Two PCs capture a large proportion of
Polychoric Correlation	An RxC contingency table of ordinal row and column variables (from n bivariate ordinal data pairs)	A polychoric correlation and se	Belief in the existence of underlying latent variables, and that they are bivariate normal