TOPIC 10 UNCONSTRAINED ORDINATION: PRINCIPAL COORDINATE ANALYSIS

Principal Coordinates Analysis (PCoA)

- see <u>https://sites.google.com/site/mb3gustame/dissimilarity-based-methods/principal-coordinates-analysis</u> and <u>https://www.davidzeleny.net/anadat-r/doku.php/en:pcoa_nmds</u>
- a.k.a. 'Classical multidimensional scaling' (MDS), hence the base R function name cmdscale()
- PCoA is conceptually similar to PCA, which preserves Euclidean distanced between objects, and CA which preserves χ^2 distances between objects
- rather than using raw data as in a CA, PCoA takes a (dis)similarity matrix as input
 - the strength of the method is that it can be used with other dissimilarities (as per **vegan**'s **vegdist()** function) that are better suited for communities of species
 - if dissimilarities are Euclidean distances, then PCoA is equal to PCA
- also, (dis)similarity matrices calculated from **quantitative, semi-quantitative, qualitative, and mixed variables** can be handled by PCoA
- **info about original variables are not available** because the input is the square (site × site) dissimilarity matrix, not the raw species table
- **the proportion of a given eigenvalue to the sum of all eigenvalues** reveals the relative 'importance' of each axis

Principal Coordinate Analysis (PCoA)





- 1. produce the code to run this analysis on the seaweed species data
- 2. what is the importance of the first three Dims?
- 3. how does the interpretation compare to that in the CA?
- 4. produce a multivariate analysis of the seaweed environmental data, and use this to explain the patterns observed in the species data (reuse your earlier results if you have done this before)



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Update:

I now use the **vegan** function **capscale()** for PCoA – see the Supplementary Material at <u>https://</u>github.com/ajsmit/Quantitative_Ecology/blob/main/jupyter_lab/14-PCoA.ipynb