

3502-470 Plant Genetic Resources

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# How can individuals or populations be grouped?

- Geographic origin
- Phenotypic similarity
- Genetic similarity

**But:** Geographic origin and phenotypic similarity can be frequently misleading.

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## Caveats with genetic similarity

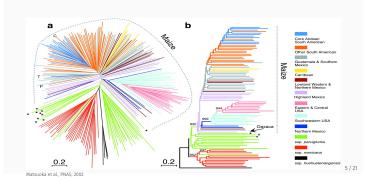
- Often a large proportion of genetic variation segregates within a (sub)population
- Only few genes may show significant structuring (i.e., adaptive trait genes)
- Many markers are needed to differentiate between genome-wide and gene-specific clustering

# Methods for population structure inference

- $\cdot$  Phylogenetic analysis
- Principal components analysis
- Model-based analysis

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## Phylogenetic analysis of maize varieties



# Phylogenetic analysis

#### Pros

- Reflects phylogenetic relationships
- Fast calculation (for distance-based methods)
- Gives measure of genetic distance
- Straightforward interpretation

## Cons

- What is the correct number of clusters?
- What if there is gene flow or reticulate evolution?

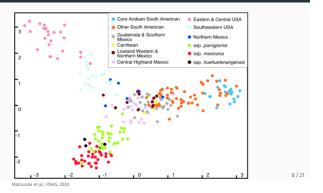
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## Principal components analysis (PCA)

### General idea:

- Reduce the dimensions of multivariate data. Here, each marker is a dimension
- PCA is model-free and parameter-free
- Approach: Find different combinations of markers such that these combinations are uncorrelated. These combinations are called principal components
- The total variation in the data is explained by all principal components
- To reduce complexity, only retain a small number of components (those with highest variances) which explain a certain amount of the total variance or just set a number of dimensions arbitrarly to see the most important principal components
- Each data point gets projected on the chosen principal components, distinguish clusters by their values on the principal components

### Principal component analysis (PCA) of maize varieties



## Pros and Cons of PCA

#### Pros

- Reduction of complexity in data
- Fast calculation
- Principal components can be used for 'downstream applications'
- Identify the important processes that affect data

#### Cons

- $\cdot\,$  Determination of distinct clusters is difficult
- $\cdot\,$  No easy interpretation of evolutionary processes
- · Gene flow and reticulate evolution are difficult to identify

## Model-based structure inference

- $\cdot$  Use an explicit population genetic model
- Assumptions:
  - Each subpopulation is in Hardy-Weinberg equilibrium (HWE)
- All markers are in complete linkage equilibrium within populations
- Any deviation from assumption is due to the presence of population structure

Implementation:

- STRUCTURE was the most popular program (Pritchard et al., 2000), but use the faster implementation *admixture* (Alexander et al., 2009)
- In STRUCTURE, two models are possible: with and without admixture
- STRUCTURE is a Bayesian method
- A similar method is implemented in the R package 'LEA'

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### How does admixture work?

#### Assumptions:

Each subpopulation is in Hardy-Weinberg equilibrium (HWE)
All markers are in complete linkage equilibrium within populati

#### Maximum likelihood approach (with admixture)

- Assume that the sample contains genetic material inherited from *K* (ancestral) populations, which had allele frequencies  $A = (A_{i,1} \dots, A_{i,K}) \text{ at the observed loci } i$
- For each individual, assume that a fraction of  $q_j$  its DNA can be traced back to the *j*th (ancestral) populations
- Compute the likelihood of the genetic data of the individuals being produced from *K* subpopulations with the assumed *q*'s and *A*'s for every possible choice of *q*'s and *A*'s
- Take the ancestry fractions q that produce the highest likelihood <sup>11/21</sup>

How to choose K?

#### Choosing the number of populations in admixture

How to choose the best K? E.g. via cross-validation (see Alexander and Lange), other methods available

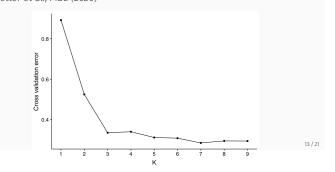
For each K, do multiple times:

- Mask a proportion of the SNP matrix (e.g. SNP position 5 in individual 3,...) at random
- Estimate the ancestries and ancestral allele frequencies as on the previous slide without the masked information
- · Predict the masked entries, record your error

Take the *K* that has smallest error averaged over the multiple cross-validations

## Example of cross-validation

Population structure analysis of wild and domesticated amaranths (Stetter et al., MBE (2020)



xample of a STRUCTURE analysis in cultivated barley (Stracke et al.)							
	rym5 rym4	Euro	pean U winter 2-rowed 6-rowed	Asian	Hs		
				lan	K = 2 Ln P(D) = -1894.0 $\alpha = 0.056$		
				[]	K = 3 Ln P(D) = -1646.9 $\alpha = 0.042$		
					K = 4 Ln P(D) = -1562.0 $\alpha = 0.036$		
					K = 5 Ln P(D) = -1588.3 $\alpha = 0.034$		
	Genotypes				K = 6 Ln P(D) = -1535.2 $\alpha = 0.035$		
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### Pros and cons of model-based inference

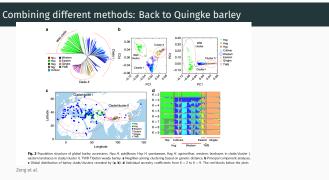
#### Pros

- $\cdot$  Is based on an explicit population genetic model
- Distinct populations can be identified
- Admixture can be detected
- Straight-forward interpretation

#### Cons

- Running times can be long
- $\cdot\,$  No (direct) distance measure between populations
- $\cdot$  Assumptions of model are frequently violated\*
- Strong influence of sampling structure, missing data (also in other methods) 15/21

\*: Some violations can be circumvented, e.g. run the method on a subset of SNPs essentially without LD



K = 9 chosen based on earlier results on worldwide barley diversity,

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### Summary

- The analysis of population structure is an important aspect of the study of genetic resources.
- A phylogenetic approach is usually fast and robust, however it does not provide well defined criteria for population subdivision and the assignment of individuals to populations.
- Principle component analysis is a fast and reliable methods, that is not based on a model of evolution.
- Model-based approaches such as the STRUCTURE program allow to estimate the number of populations and to assign individuals to different populations. May have long running times for large data sets.

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### Literature i

Cited studies using population structure inference

- Matsuoka et al. (2002) A single domestication for maize shown by multilocus microsatellite genotyping *Proc. Nat. Acad. Sc. USA* 99 (9): 6080-6084
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- Jombart, Thibaut, Sébastien Devillard, and François Balloux.
   "Discriminant analysis of principal components: a new method for the analysis of genetically structured populations." *BMC genetics* 11.1 (2010): 94.
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- D.H. Alexander, J. Novembre, and K. Lange. Fast model-based estimation of ancestry in unrelated individuals. *Genome Research* 19 (2009): 1655–1664

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### Literature iii

- Gauch Jr, H. G., Qian, S., Piepho, H. P., Zhou, L., & Chen, R. (2018). Effective principal components analysis of SNP data. bioRxiv, 393611.
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### References i

Matsuoka Y, Vigouroux Y, Goodman M, Sanchez G, Buckler E, Doebley J (2002) A single domestication for maize shown by multilocus microsatellite genotyping. Proc Natl Acad Sci USA 99(9):6080–6084, DOI 10.1073/pnas.052125199