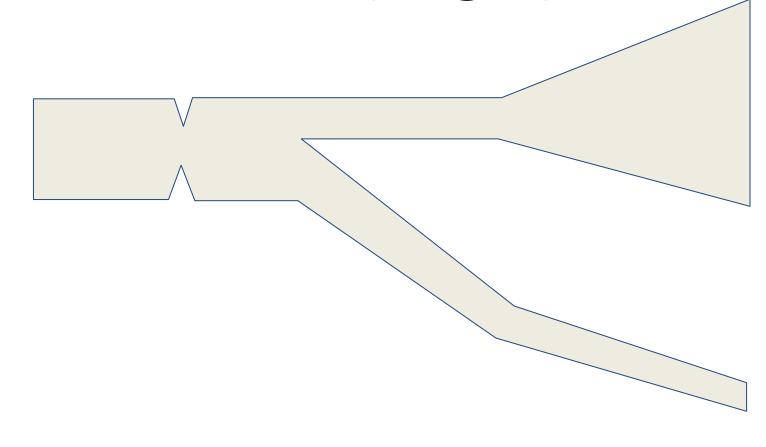
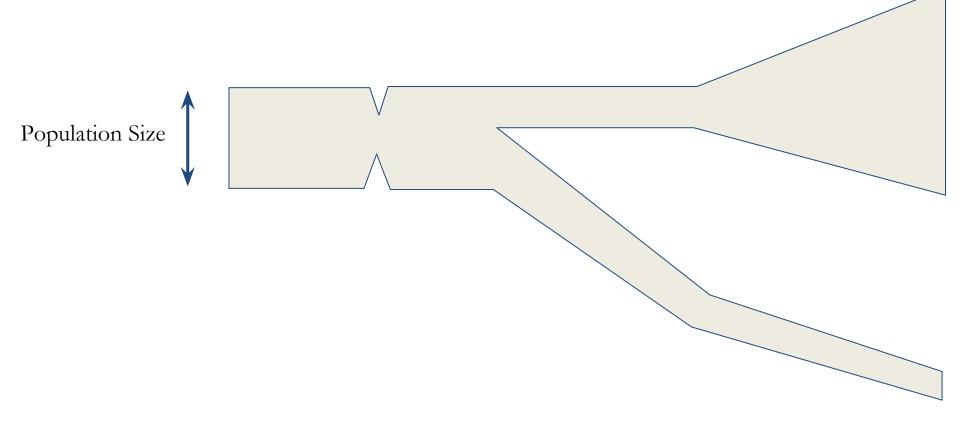
Analysing SNP data

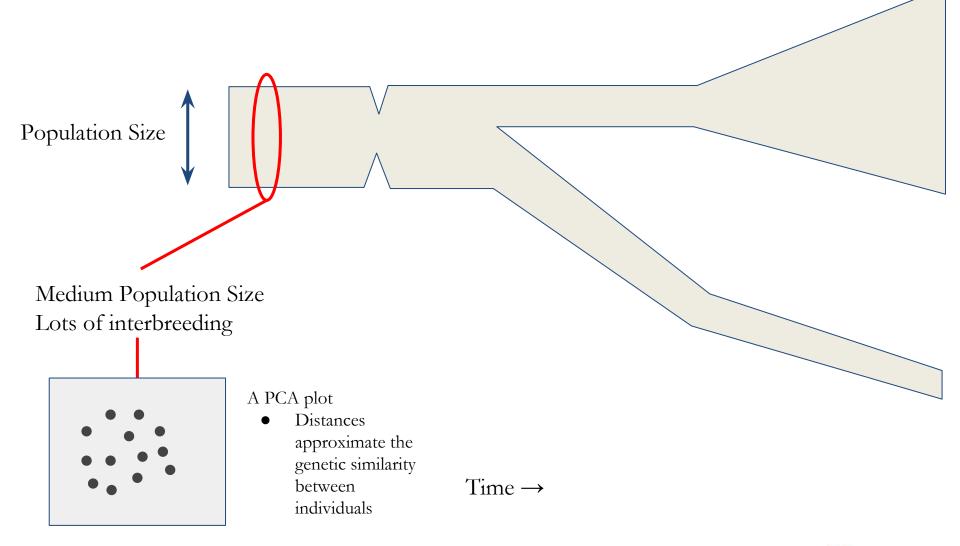
PCA and GWAS

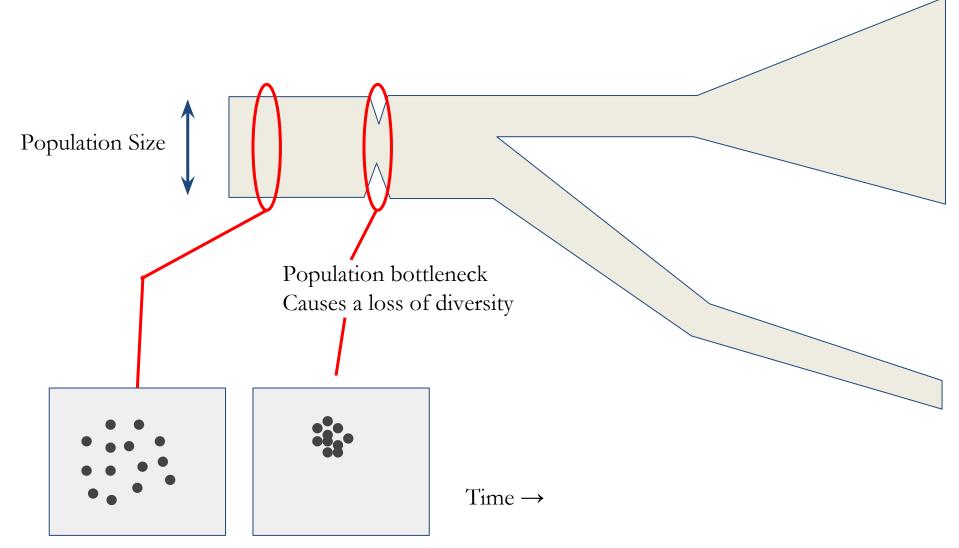


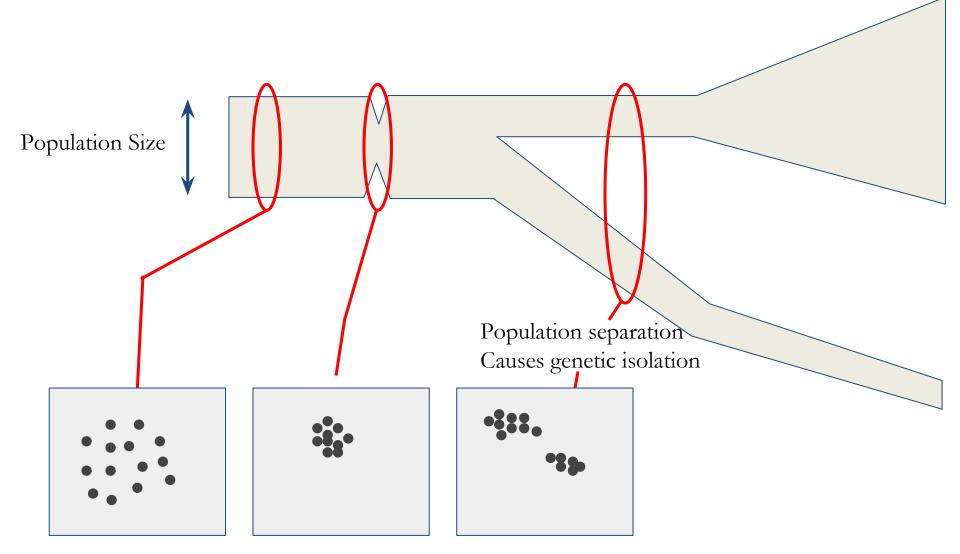
Time \rightarrow



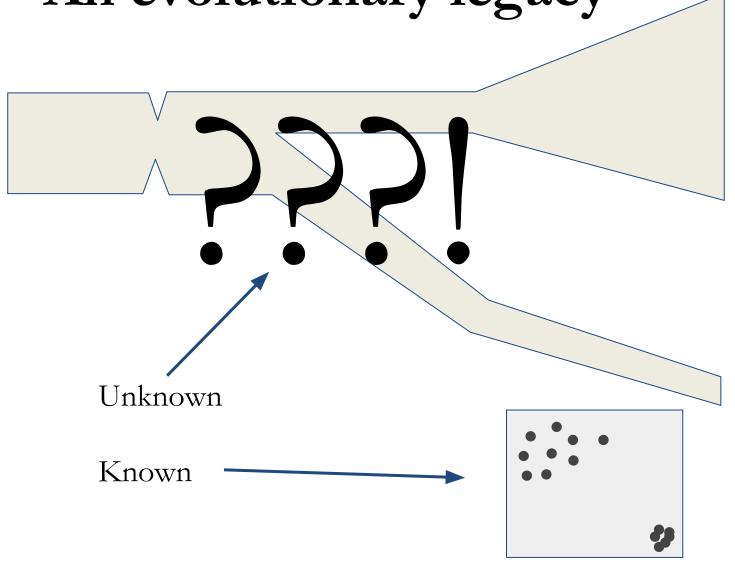
Time \rightarrow





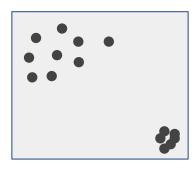


Population Genetic Diversity: An evolutionary legacy Population Size Large population gains diversity Small population has less diversity Long period of isolation



Population Genetic Diversity: A valuable resource

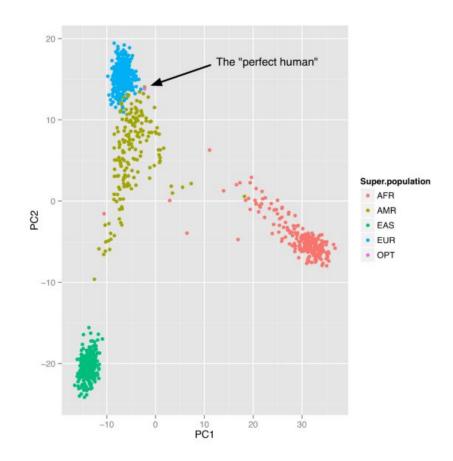
Each individual's genetic material is the product of millions of years of evolution to to tolerate different ...



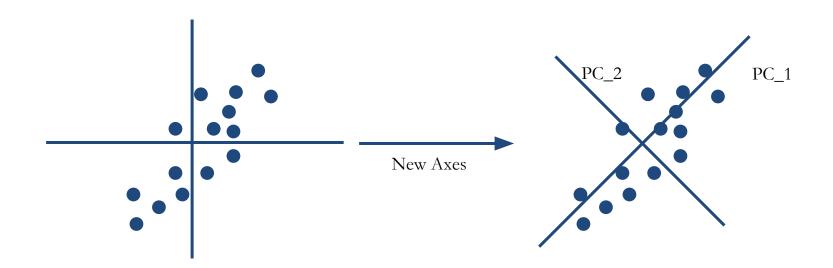
- Temperatures
- Moisture levels
- Viruses
- Soil conditions
- Day lengths
- Cold periods or frost
- Wind conditions
- Bacteria
- Nematodes, insects
- Fungal pathogens
- Seasonal extremes
- Droughts
- Nutrient deficiencies
- Nutrient excesses
- Salt levels
- Herbivores
- Symbioses
 - ... etc etc etc

PCA with SNP data:

- An excellent way to visualise diversity, and
- An efficient mathematical way to specify population structure



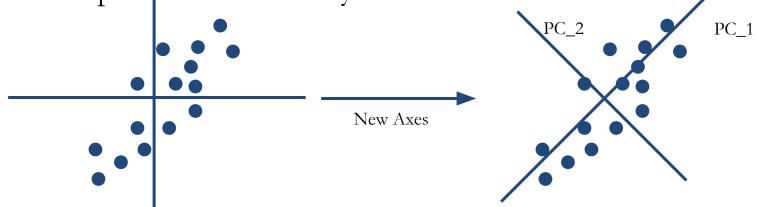
• A way of specifying new "axes" to the data, so that the new axes express the variation in the data best.



• A way of specifying "new axes" for the data, so that the new axes (or Principal Components) capture the highest possible variation in the data.

• If we need to describe the data using fewer dimensions, the new

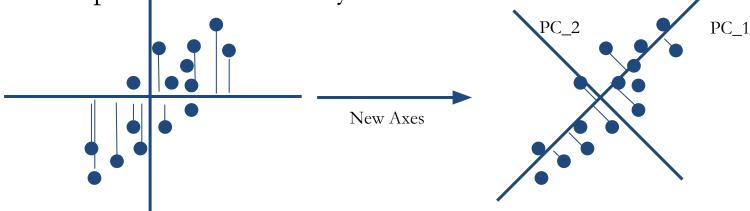
axes provide the best way to do it.



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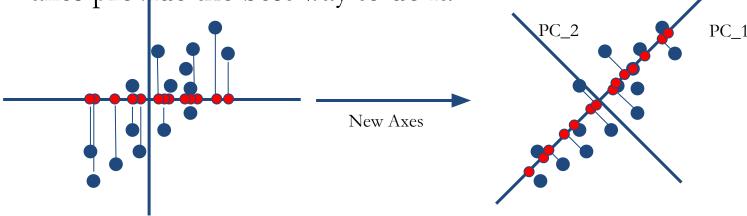


From 2D to 1D:

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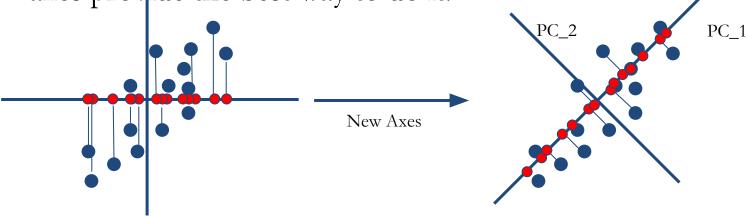


From 2D to 1D:

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From 2D to 1D:

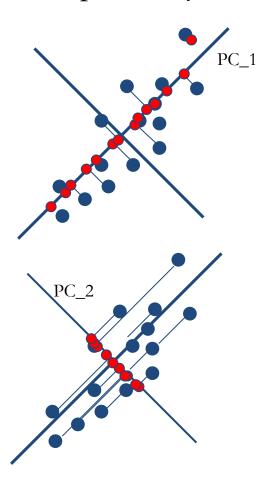


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- From 3D to 2D:

```
https://www.google.com/search?biw=1440&bih=767&tbm=isch&sa=1&ei=cw2cXLOzAZGblwT-o6TQCQ&q=eigenvectors+3d+gif&oq=eigenvectors+3d+gif&gs_l=img.3...551553.556272...556385...0.0...0.84.386.6......1....1...gws-wiz-img.2TdLFyVmTjY#imgrc=30k1xua4SqZHkM:
```

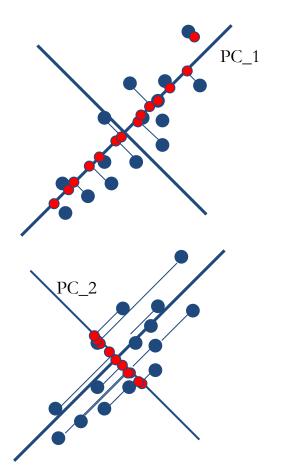
- A way of specifying "new axes" for the data, so that the new axes (or Principal Components) capture the highest possible variation in the data.
- If we need to describe the data using fewer dimensions, the new axes provide the best way to do it.
- From 3D to 2D
- A SNP dataset has as many "dimensions" as there are SNPs.
 - And the values in it are discrete (e.g., 0, 1, and 2)

• A "scree plot" shows how much of the variation in the data is captured by each PC.



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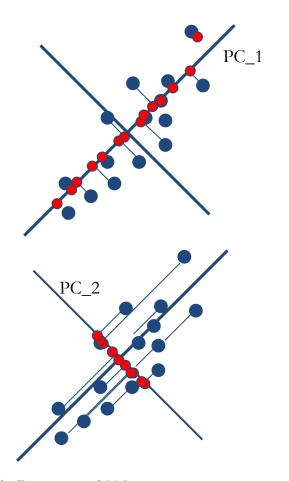
captured by each PC.

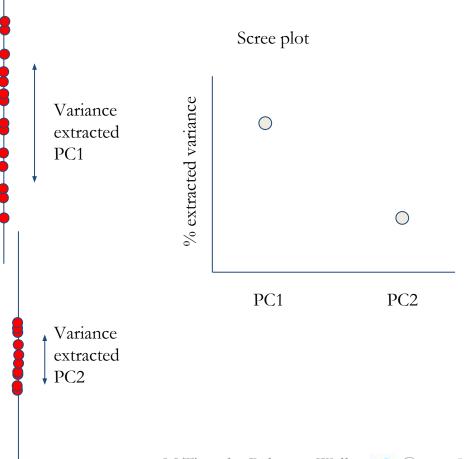


Variance extracted PC1

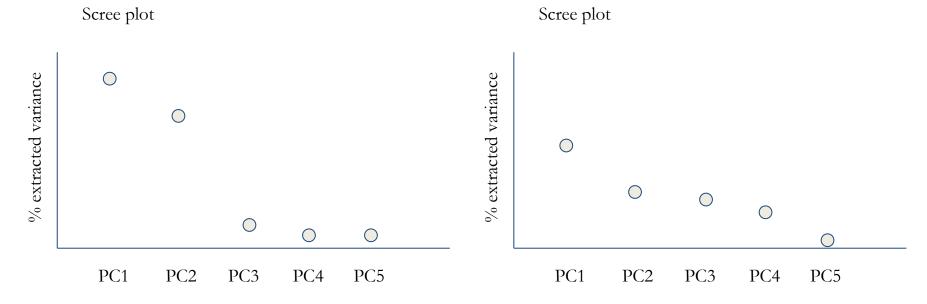
Variance extracted PC2

• A "scree plot" shows how much of the variation in the data is captured by each PC.





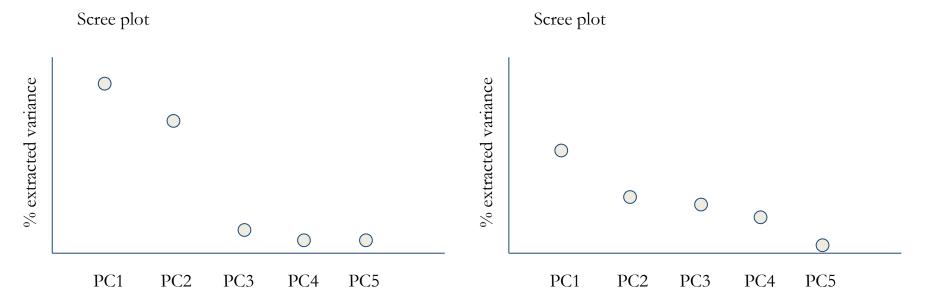
- A "scree plot" shows how much of the variation in the data is captured by each PC.
 - Interpreting scree plots



PCs 1 and 2 can be used to describe the dataset very well, because they explain most of the variance!

Need more PCs to describe most of the variation in the data. Use PCs 1--4?

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 - Interpreting scree plots



PCs 1 and 2 can be used to describe the dataset very well, because they explain most of the variance!

Need more PCs to describe most of the variation in the data. Use PCs 1--4?

A geneticist might use a rule like "use as many PCs as needed to describe 95% of the variation".

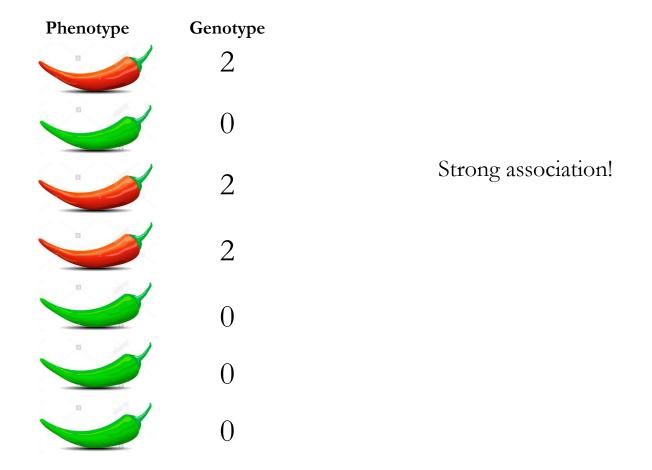
- Testing whether genetic markers (SNPs in our case) are associated with a phenotype.
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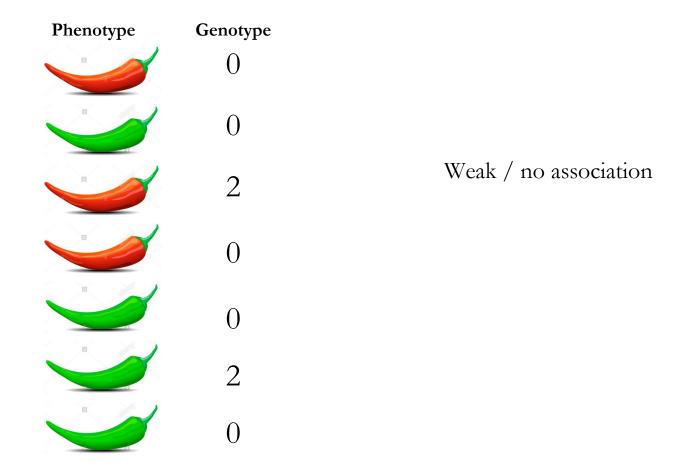
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 - That's 'associated with' ... not 'cause'!
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 - Means the SNP may be genetically near a gene that causes the phenotype \circ
 - Means the SNP allele can be used as a marker to screen for the (probable) presence of the causal gene(s)

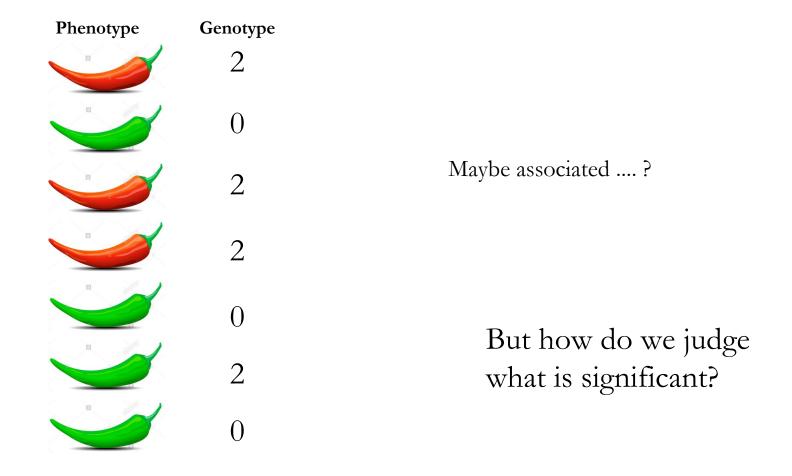
Intuition



Intuition



Intuition



Any appropriate statistical test can be used to judge whether a SNP is likely associated with a phenotype, e.g.:

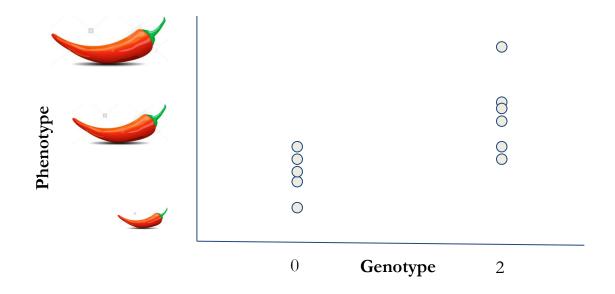
Chi squared	Phenotype		
	2	0	
type	20	52	
Phenoty	32	32	

Null hypothesis: The genotype and the phenotype are independent

... generate a p-value ...

Any appropriate statistical test can be used to judge whether a SNP is likely associated with a phenotype, e.g.:

t-test



Null Hypothesis: The mean sizes of plants of the two genotypes are equal

... generate a p-value

Any appropriate statistical test can be used to judge whether a SNP is likely associated with a phenotype, e.g.:

We will use a linear regression

- Fast
- Helps account for population structure
- You are not required to know details, just:
 - Null hypothesis: "The SNP allele is required to explain the phenotype."

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Genome Wide Association Study (GWAS)

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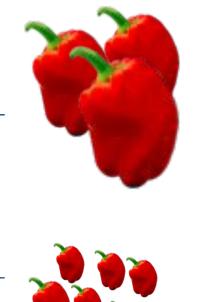
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Genome Wide Association Study (GWAS)

(we just use chromosome 1H)

CGATTCG GCGGGGCTCCTCTCAGGATGCTTAAA
CGATTCG GCGGGGCTCCTCTCAGGATGCTTAAA
CGATTCG GCGGGGCTCCTCTCAGGATGCTTAAA
CGATTCG GCGGGGCTCCTCTCAGGATGCTTGAA
CGATTCG GCGGGGCTCCTCTCAGGATGCTTGAA
CCGATTCG GCGGGGCTCCTCTCAGGATGCTTGAA
CCGATTCGCGCGCGGGGCTCCTCTCAGGATGCTTGAA
CCGATTCGCGCGCGGGCTCCTCTCAGGATGCTTGAA
CCGATTCGCGCGCGGGCTCCTCTCAGGATGCTTGAA
CCGATTCGCGCGCGGGCTCCTCTCAGGATGCTTGAA
CCGATTCGCGCGCGGGCTCCTCTCAGGATGCTTGAA



Does either SNP associate with fruit size?

AGCCCTGTCTCTAGGCTCTCGATTCGTGCGGGGCTCCTCTCAGGATGCTTAAA
AGCCCTGTCTCTAGGCTCTCGATTCGTGCGGGGGCTCCTCTCAGGATGCTTAAA
AGCCCTGTCTCTAGGCTCTCGATTCGTGCGGGGGCTCCTCTCAGGATGCTTAAA
AGCCCCGTCTCTGGGCTCTCGATTCGTGCGGGGCTCCTCTCAGGATGCTTGAA
AGCCCCGTCTCTGGGCTCTCGATTCGCGCGCGGGGCTCCTCTCAGGATGCTTGAA
AGCCCCGTCTCTGGGCTCTCGATTCGCGCGCGGGCTCCTCTCAGGATGCTTGAA
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AGCCCCGTCTCTGGGCTCTCGATTCGCGCGCGGGCTCCTCTCAGGATGCTTGAA



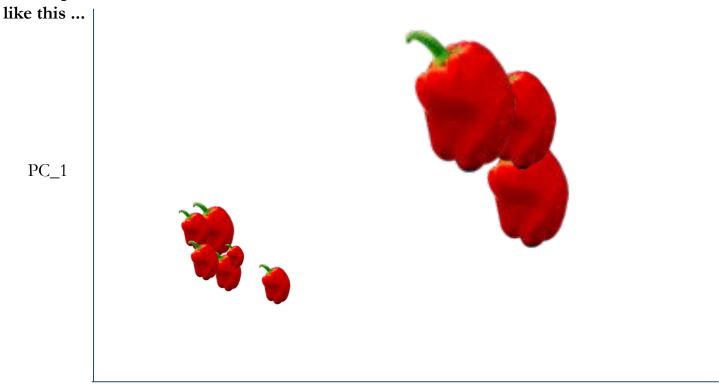


What about now?

A PCA plot would look a bit like this ... PC_1

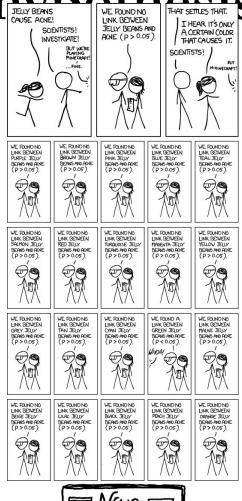
PC_2

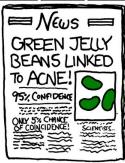
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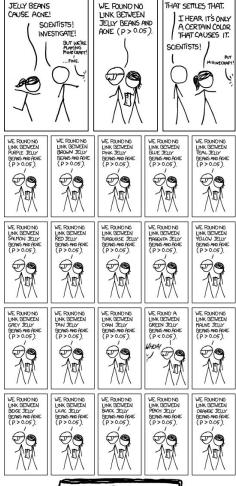
PC_2

- The principal components are able to predict the phenotype!
 - i.e., population structure is a strong influence
- We can control for this in the linear model
 - The model will test how much extra predictive power the SNP allele gives us, when the population structure (summarised by the PCs) are also used to predict the phenotype.

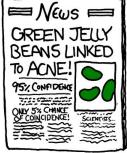




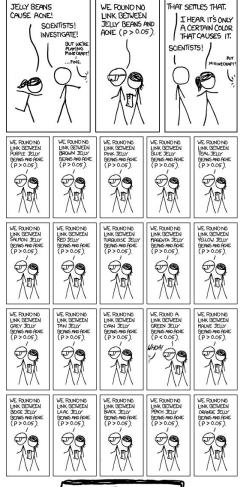
Multiple hypothesis testing ...



• The normal "significance threshold" of p-value = 0.05 causes false positives when we test many hypotheses



Multiple hypothesis testing ...

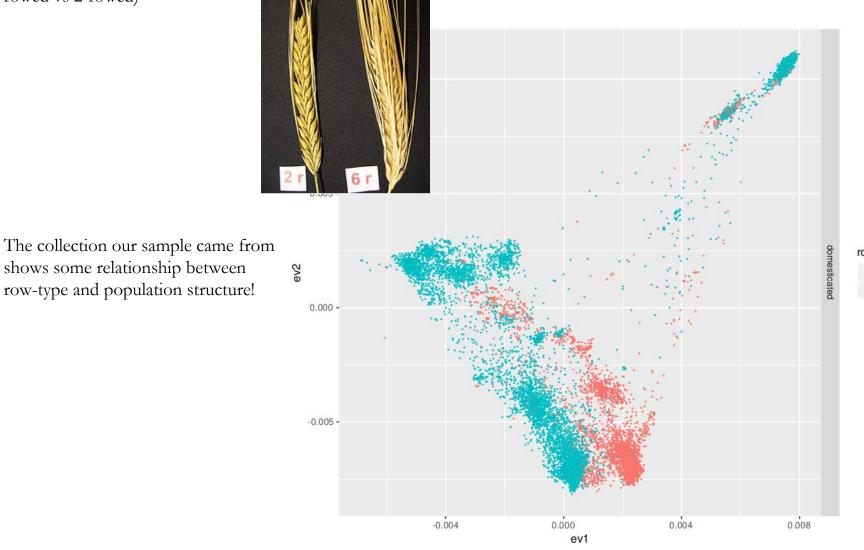


- The normal "significance threshold" of p-value < 0.05 causes false positives when we test many hypotheses.
- We correct for this by setting the threshold much more stringently.
- The "Bonferroni correction" involves simply dividing the p-value significance threshold (0.05) by the number of tests.
- The number of tests is the number of SNPs.

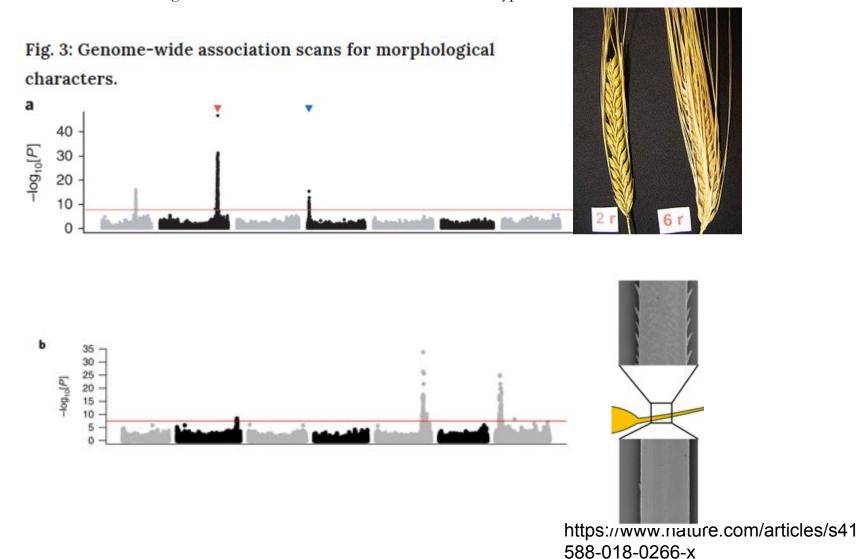
Test for differences in row type (6-rowed vs 2-rowed)



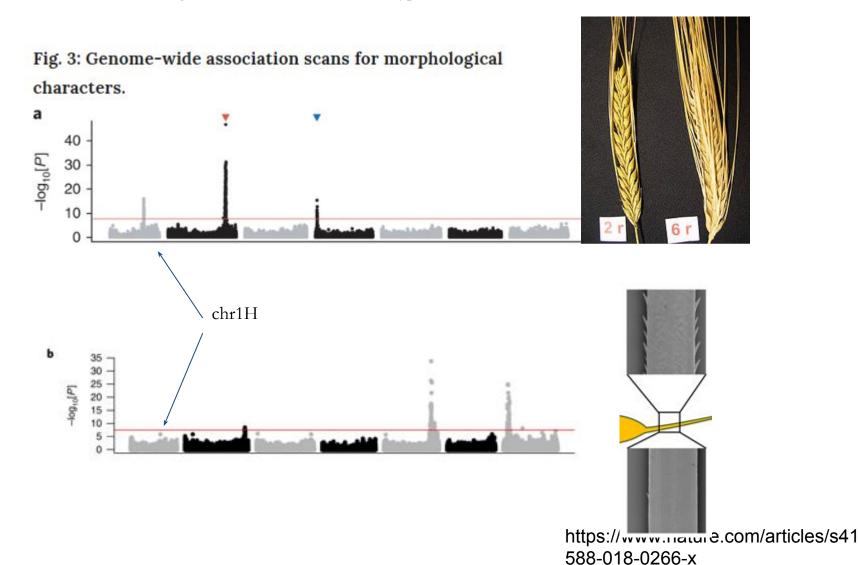
Test for differences in row type (6-rowed vs 2-rowed)



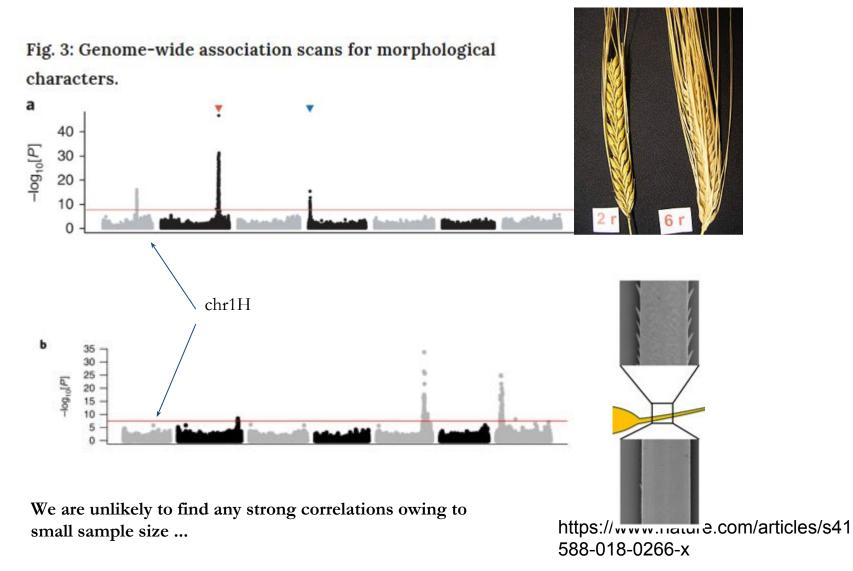
Previous GWAS results show genetic associations with awn barbs and row type



Previous GWAS results show genetic associations with row type (and also awn barbs, below)



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So ... primary aims:

- Load our cleaned SNP data into R
- Create a PCA to summarise the population structure between samples
 - Plot a PCA plot!
- Link it to phenotype data from a database
 - Using a new data.table trick ...
 - More PCA plots ...
- Link three datasets: **The PCs** that summarise the population structure, the **SNPs**, and the **phenotype data**
- Run a GWAS to test associations at each SNP
- Plot the GWAS results as a Manhattan Plot with the significance threshold shown on the plot.