# Identifying a diamond in the rough: studying allelic variation for complex traits in maize landraces

Jorge Alberto Romero Navarro
Buckler Lab
Cornell University





#### International Maize and Wheat Improvement Center (CIMMYT)

- Sarah Hearne
- Martha Wilcox
- Juan Burgueño
- Samuel Trachsel
- Ivan Ortiz-Monasterio
- Félix San Vicente
- Peter Wenzl
- Gary Atlin



Instituto Nacional de Investigaciones Forestales, Agrícolas y Pecuarias

#### Instituto Nacional de Investigaciones Forestales Agrícolas y Pecuarias (INIFAP)

- Frnesto Preciado
- Arturo Terron
- Victor Vidal
- Humberto Vallejo Delgado
- Alejandro Ortega
- Noel Orlando Gómez Montiel



#### **Cornell University**

- Edward Buckler
- Cinta Romay
- Eduardo Carrillo
- Kelly Swarts









Native varieties contain plenty of useful alleles...



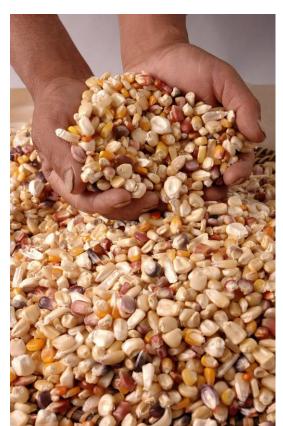
#### each linked to hundreds of undesired alleles



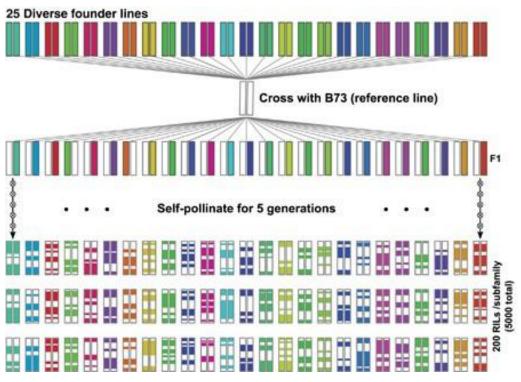


### How can we identify the diamond in the rough?

- 1989-97 Latin American Maize Project (LAMP) 11 countries characterize CIMMYT's germplasm
  - ü Description of adaptation patterns
  - ü Definition of a Breeder's core collection
  - **X** Genetic drag = limited impact
  - **X** Expensive
  - **X** Time consuming
  - **X** Marker assisted selection unavailable



### Association mapping helps identify and deploy useful variation



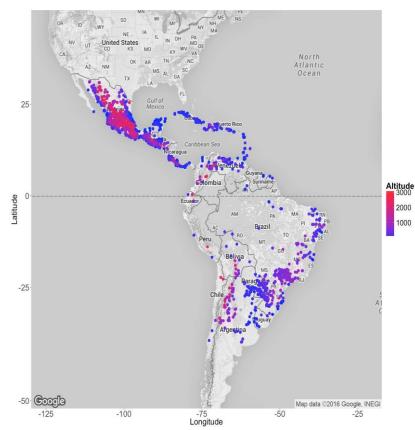


Buckler, et al, 2009

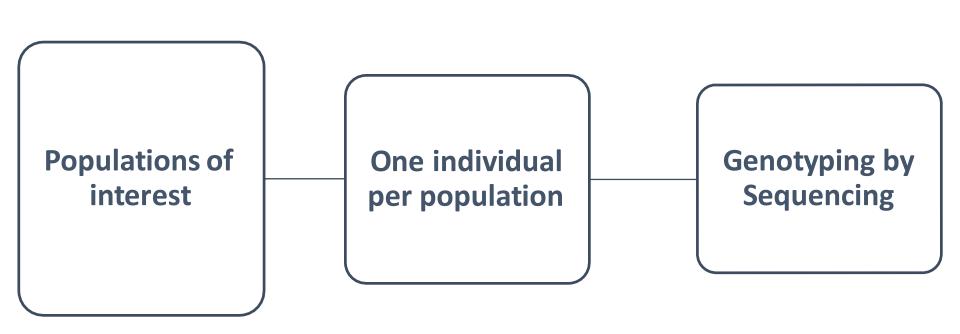
Harvestplus

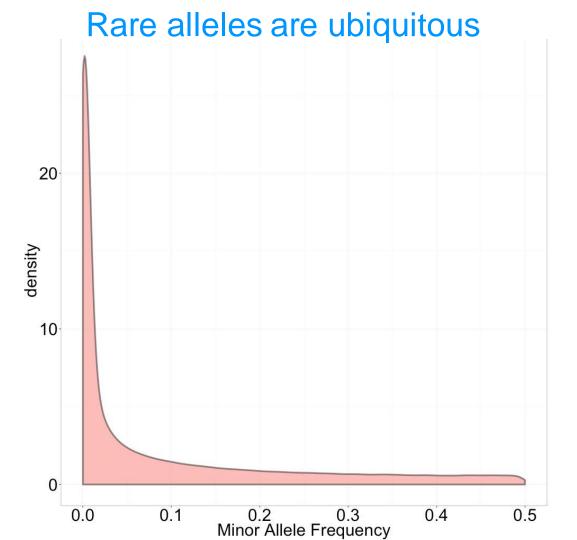
How do you evaluate a large and diverse landrace panel?

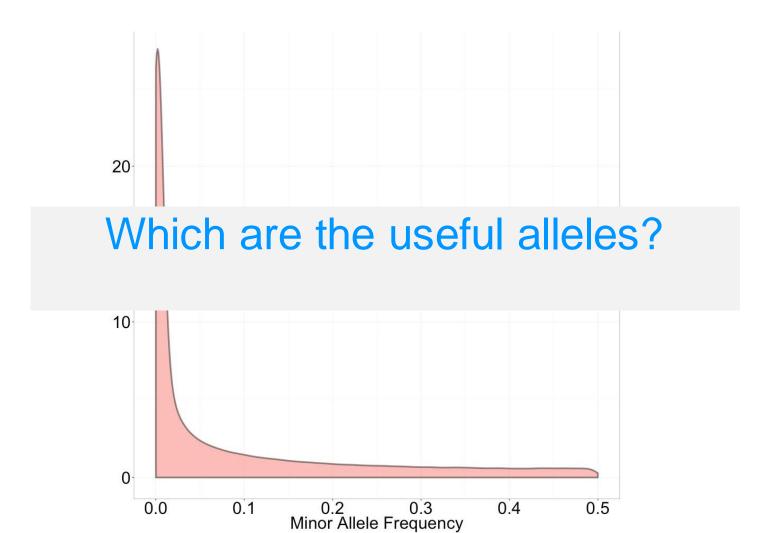
- Breeder's core collection:
   4,500 CIMMYT landrace accessions
- Adaptation:
  - Tropical (low elevation)
  - Subtropical (mid-elevation)
  - Highland



### New approach for fast identification and deployment of useful alleles from landraces





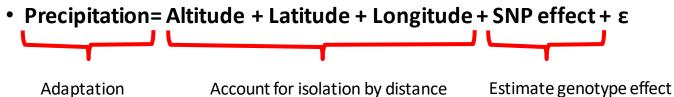


### We can use GWAS for adaptation to local environmental conditions

- Landraces have been selected for local adaptation for thousands of generations
- Passport information has been carefully collected and curated for decades for thousands of samples (and we get both amazing resources for free!)
- Example equation for environmental GWAS :
  - Average precipitation during growing season
  - Precipitation= Altitude + Latitude + Longitude + SNP effect + ε

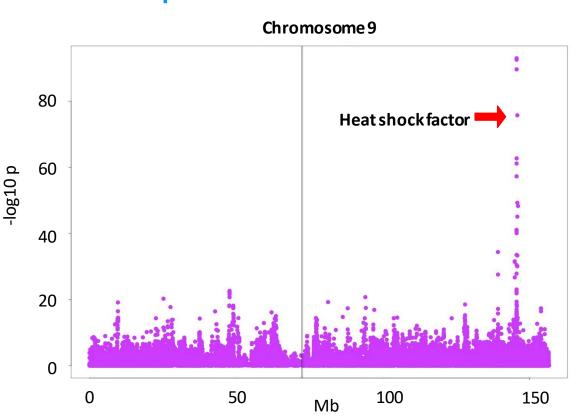
### We can use GWAS for adaptation to local environmental conditions

- Landraces have been selected for local adaptation for thousands of generations
- Passport information has been carefully collected and curated for decades for thousands of samples (and we get both amazing resources for free!)
- Example equation for environmental GWAS :
  - Average precipitation during growing season



### Using passport data we can perform high resolution GWAS for adaptation to climate

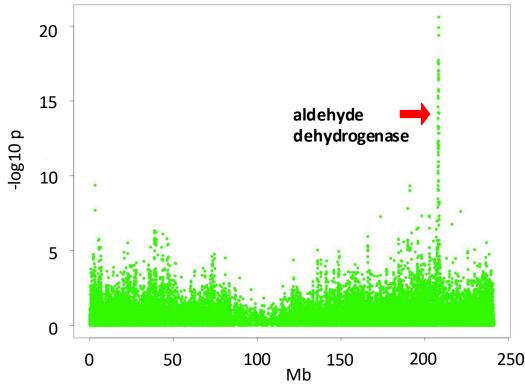
- Environmental traits
  - Altitude
  - Latitude/Longitude
  - Precipitation
  - Temperature
  - Cloud cover
  - Frost frequency
  - Vapor pressure
  - Aridity
  - Waterlog
  - pH



### Using passport data we can perform high resolution GWAS for adaptation to soils

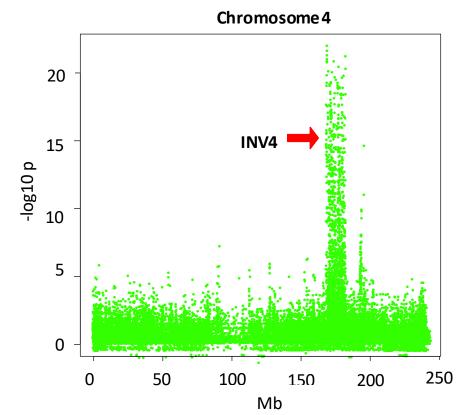
- Environmental traits
  - Altitude
  - Latitude/Longitude
  - Precipitation
  - Temperature
  - Cloud cover
  - Frost frequency
  - Vapor pressure
  - Aridity
  - Waterlog
  - pH





### Using passport data we can perform high resolution GWAS for adaptation to geography

- Environmental traits
  - Altitude
  - Latitude/Longitude
  - Precipitation
  - Temperature
  - Cloud cover
  - Frost frequency
  - Vapor pressure
  - Aridity
  - Waterlog
  - pH



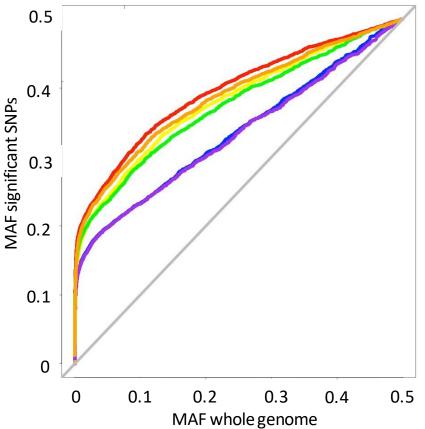
SNPs associating with adaptation are in high frequency

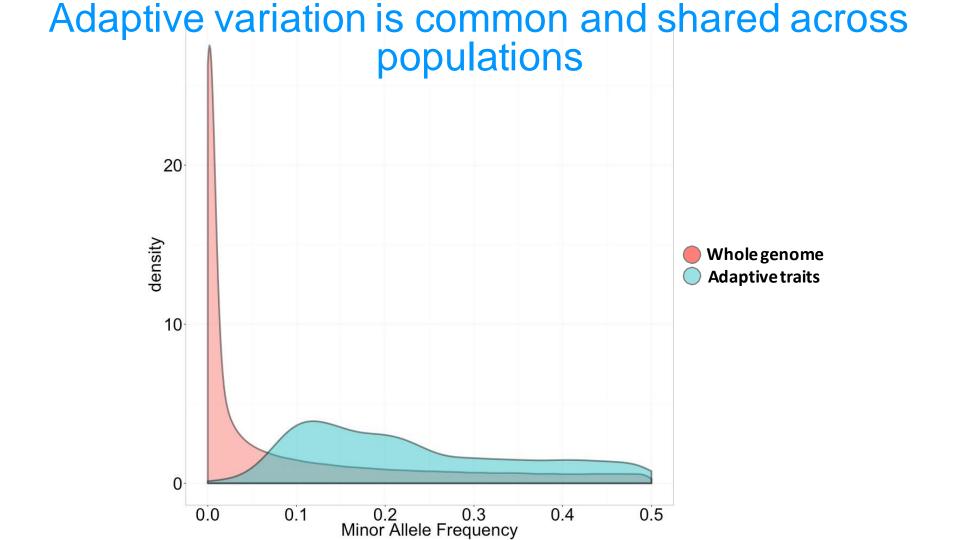
**Longitude Nutrient** 

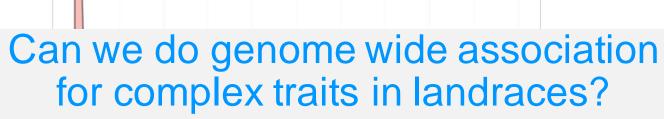
availability Latitude

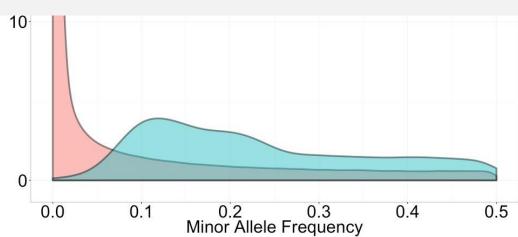
Precipitation Altitude

**Temperature** 

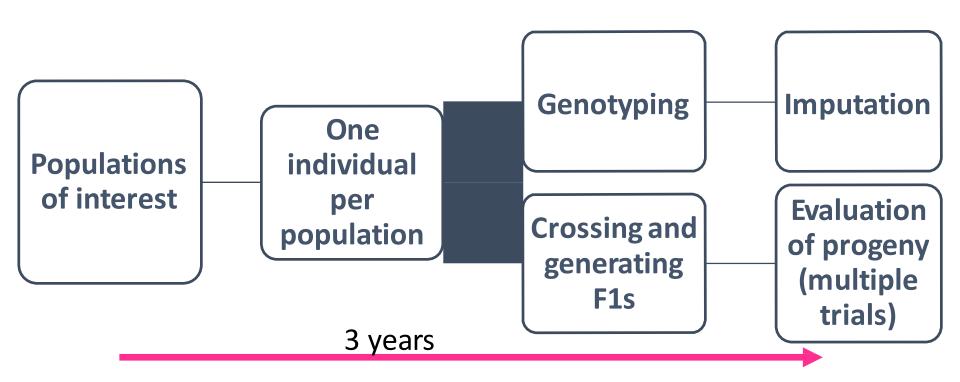




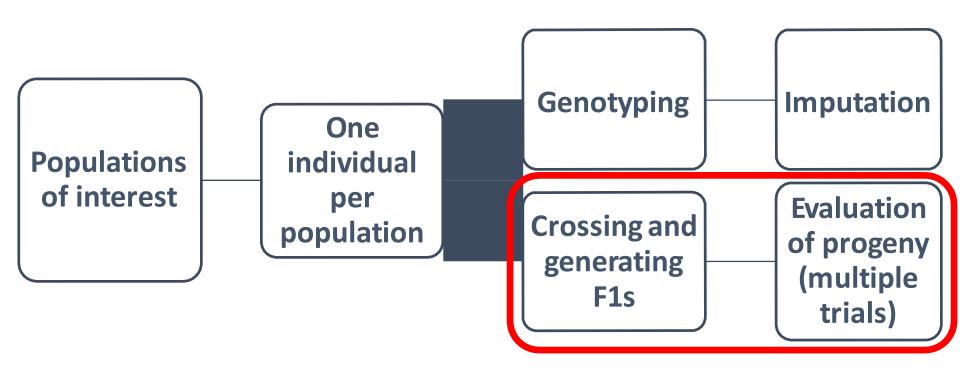


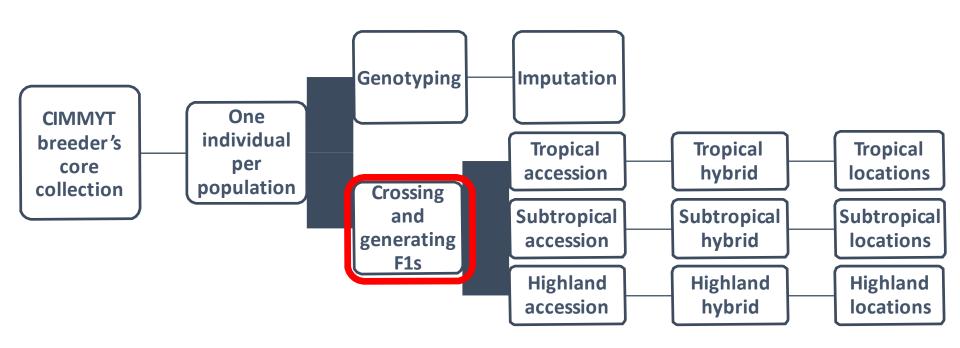


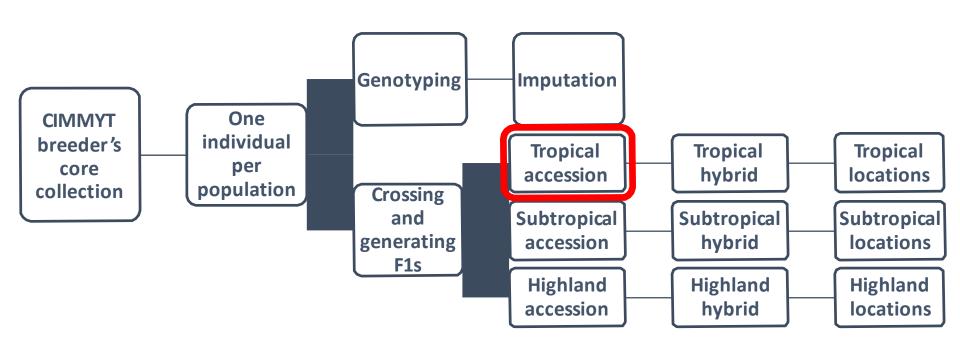
### F-One Association Mapping (FOAM) design allows fast evaluation

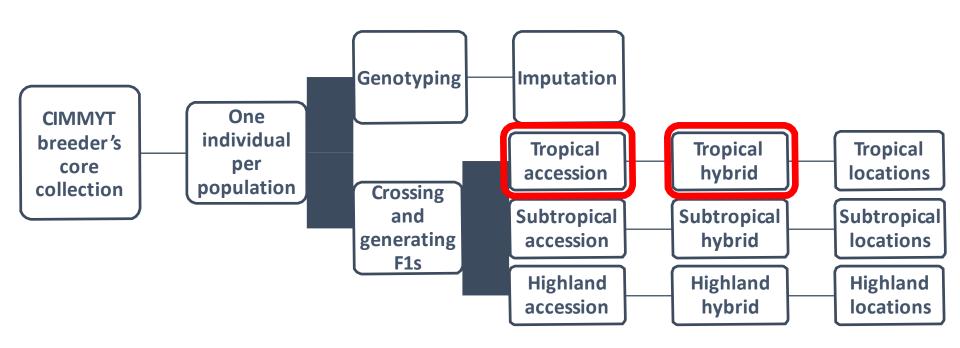


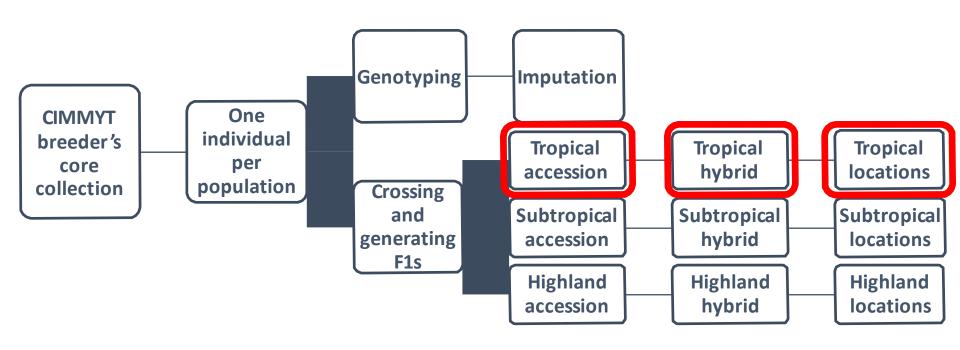
### Custom FOAM design nests adaptation class in order to do accurate evaluation













## 30 trials used to estimate ~19,000 breeding values

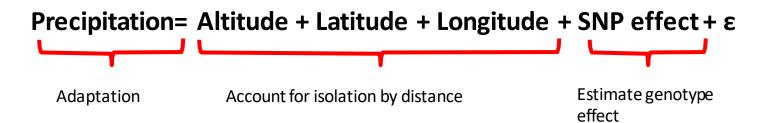


**Martha Wilcox** 

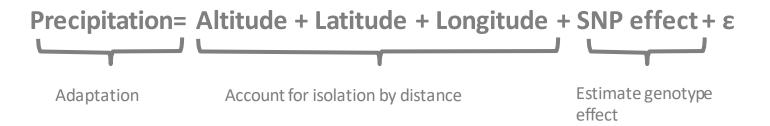
- Trial design allowed for a large number of accessions per location
  - Extended row-column
- Each trial contains a different subset of the entire collection
  - Lack of balanced replication sacrifices GxE effect estimation

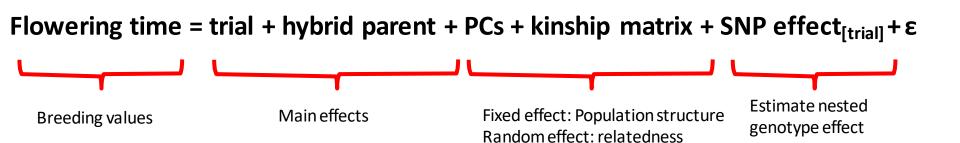


### GWAS mixed model for complex traits in FOAM design

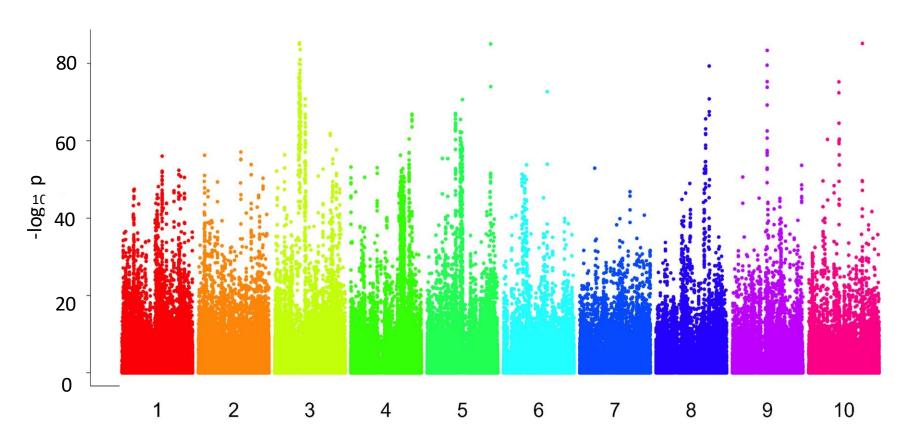


### GWAS mixed model for complex traits in FOAM design



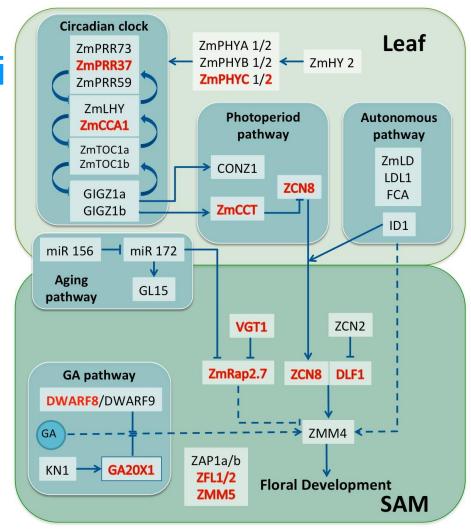


### Flowering time GWAS hits 883 genes



### Enrichment at candidate desi

- High overlap between male and female flowering time control
- Results include association at ZCN8, VGT1, and dwarf8 region

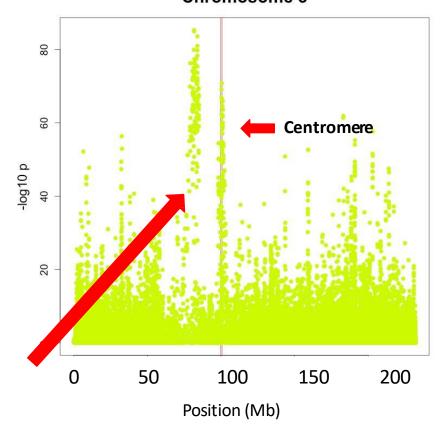


Structural variants account for 30% of explainable variance

Chromosome 3

• Centromeres 3,5 and 6

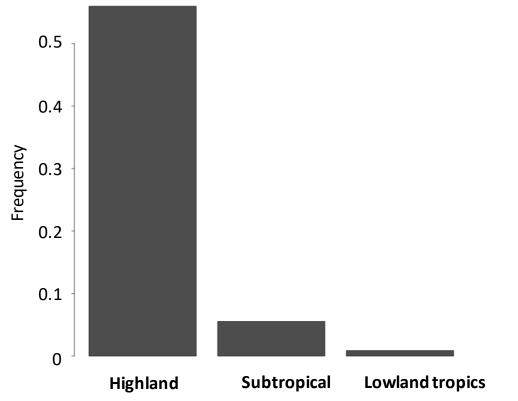
Putative unreported inversion on chromosome 3



Unreported inversion

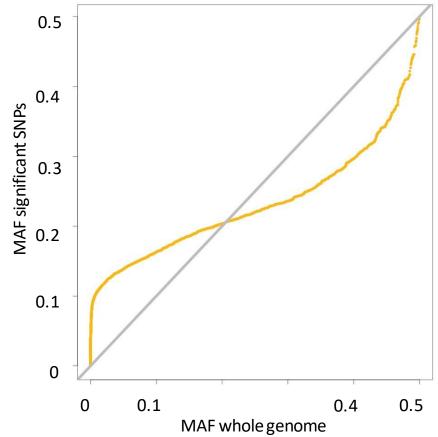
### Structural variants account for 30% of explainable variance

- Centromeres 3,5 and 6
- Putative unreported inversion on chromosome 3
- INV4
  - -3 days effect
  - Adaptation dependent frequency
  - Fixed in several tropical highland improved lines

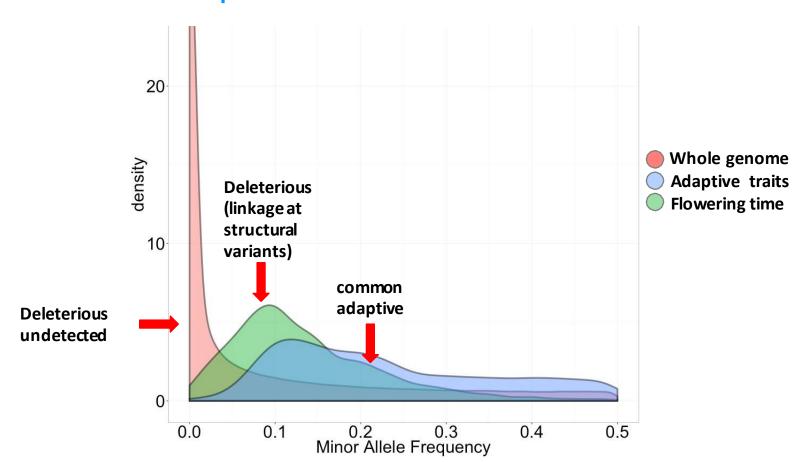


Most flowering time associating SNPs are in low frequency

- Common genesuncommon variants
- Enrichment with environmental GWAS
  - 4x temperature and altitude
  - 3.5x precipitation

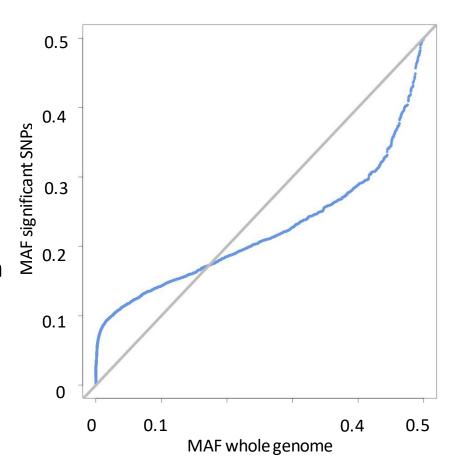


### Flowering time control mixture of common adaptive and uncommon alleles

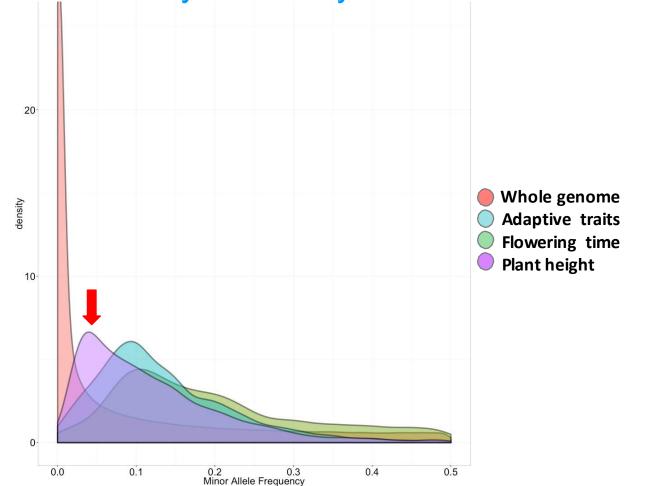


#### Plant height GWAS

- 1046 maize orthologs of *Arabidopsis* genes involved in hormonal pathways (auxin, gibberellin, brassinosteroids, abscisic acid, jasmonic acid, salicylic acid, ethylene, cytokinins)
  - 44% enrichment
- Genes associated with flowering time in SeeD
  - Three-fold enrichment



#### Height control is mostly driven by uncommon alleles



#### Summary

- FOAM (F-One Association Mapping) population design
  - Advantages: Fast, comprehensive and cost effective
  - Considerations: Adaptation and lack of replication limit GxE estimation
- Structural variation has a significant effect on local adaptation and quantitative trait variation
- GWAS
  - Adaptive alleles are in high frequency and shared across populations
  - Deleterious alleles are rare and distinct across populations

FOAM GWAS allows fast identification of useful alleles...



#### we can break linkage with undesired alleles







#### International Maize and Wheat Improvement Center (CIMMYT)

- Sarah Hearne
- Martha Wilcox
- Juan Burgueño
- Samuel Trachsel
- Ivan Ortiz-Monasterio
- Félix San Vicente
- Peter Wenzl
- Gary Atlin



Instituto Nacional de Investigaciones Forestales, Agrícolas y Pecuarias

#### Instituto Nacional de Investigaciones Forestales Agrícolas y Pecuarias (INIFAP)

- Frnesto Preciado
- Arturo Terron
- Victor Vidal
- Humberto Vallejo Delgado
- Alejandro Ortega
- Noel Orlando Gómez Montiel



#### **Cornell University**

- Edward Buckler
- Cinta Romay
- Eduardo Carrillo
- Kelly Swarts



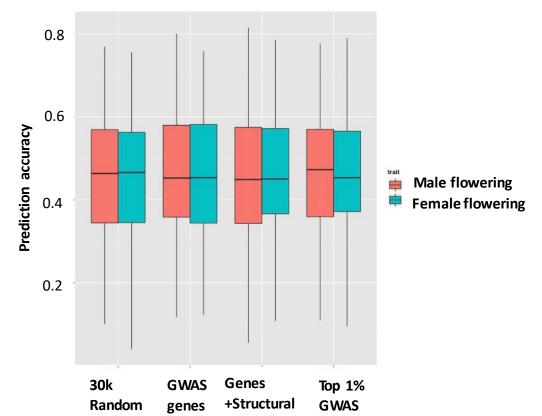




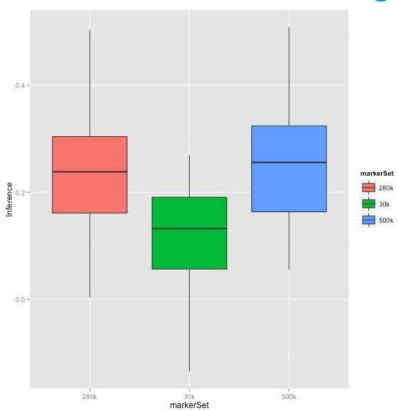


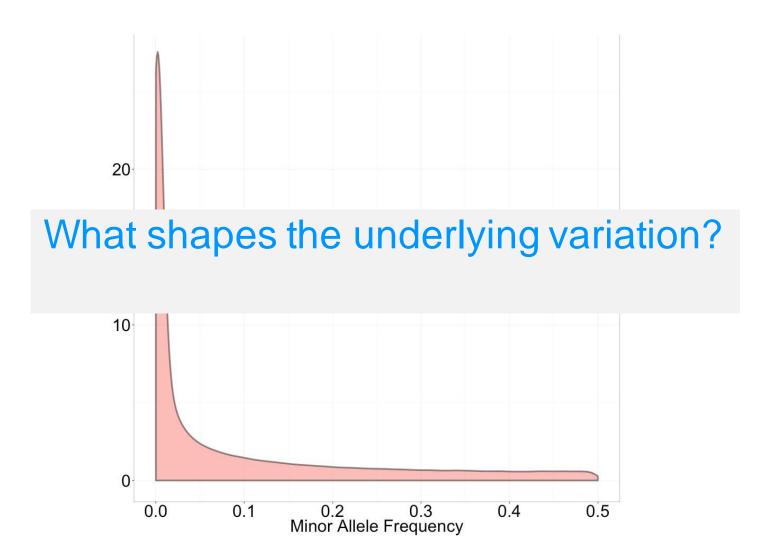
GWAS markers display prediction accuracy equivalent to 30,000 random genome wide

**SNPs** 



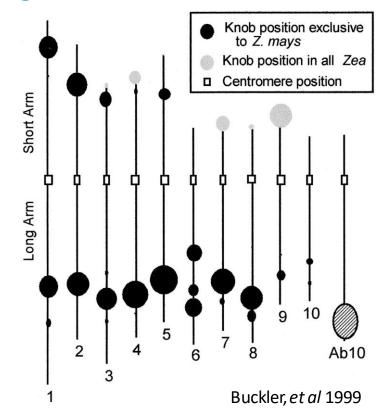
# Plant height is highly polygenic and remains hard to predict even when using all markers





### Landrace diversity has been studied extensively through cytological features



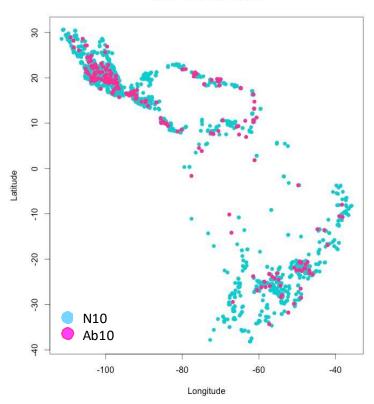


### Ab10 also segregates at low frequency

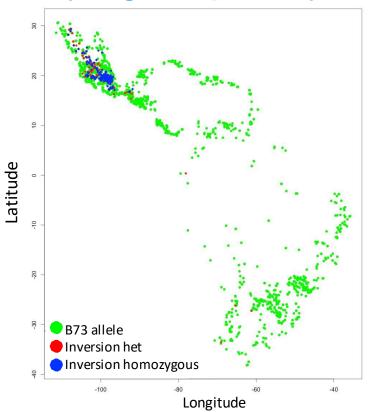


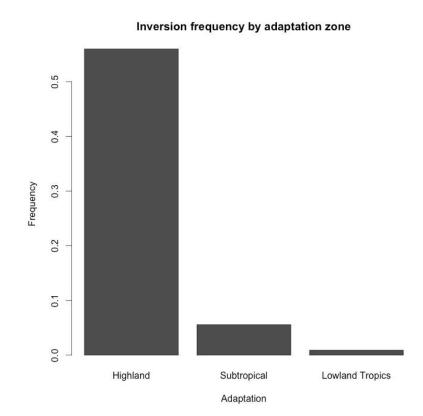
Kelly Dawe



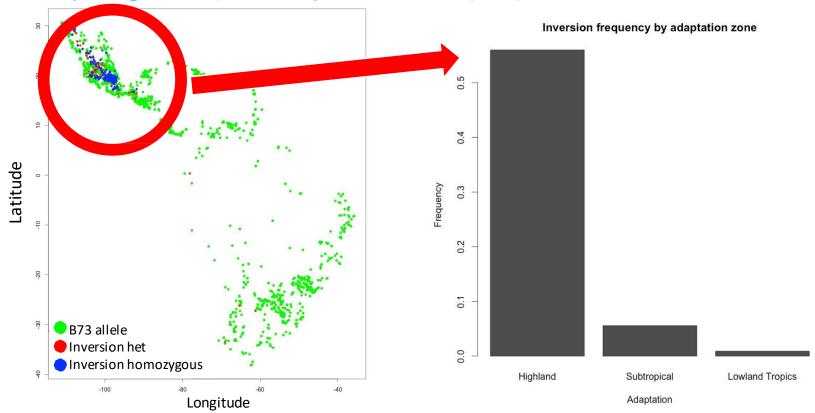


# One reported inversion (INV4) is present at varying frequency across populations

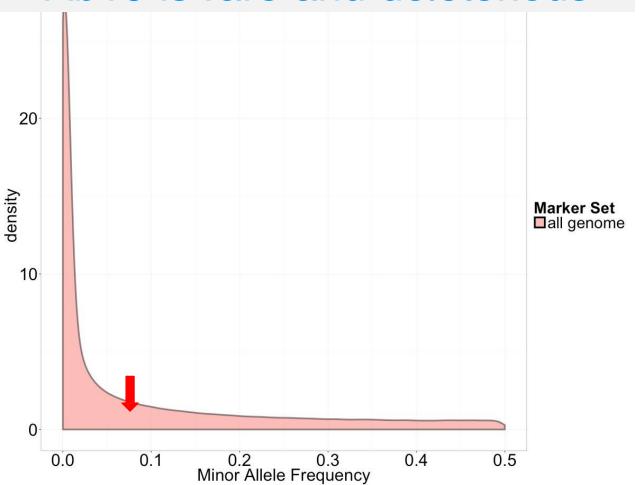


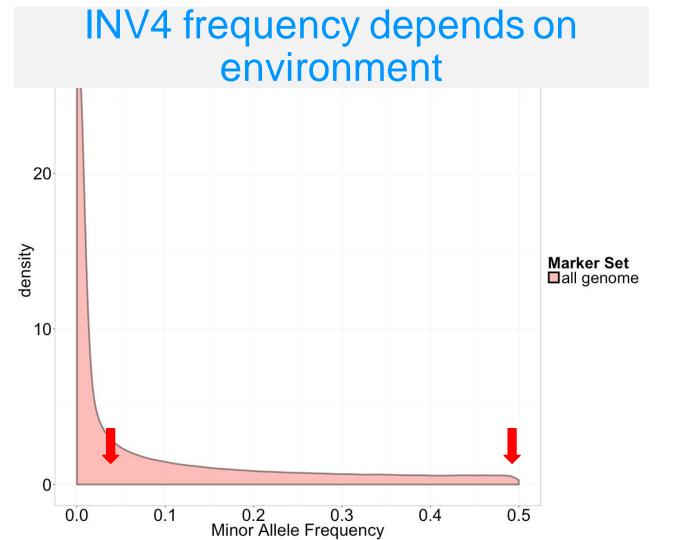


# One reported inversion (INV4) is present at varying frequency across populations



#### Ab10 is rare and deleterious



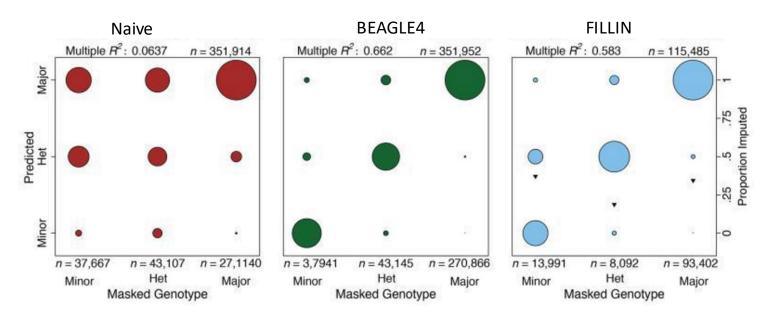


### BEAGLE4 is the best performing imputation method for maize landraces



Kelly Swarts

366 diverse landraces



• SeeD: Multiple R<sup>2</sup> FILLIN 0.53, BEAGLE4 0.68