# Genomic Predictions to Advance FHB Resistance Breeding in Barley



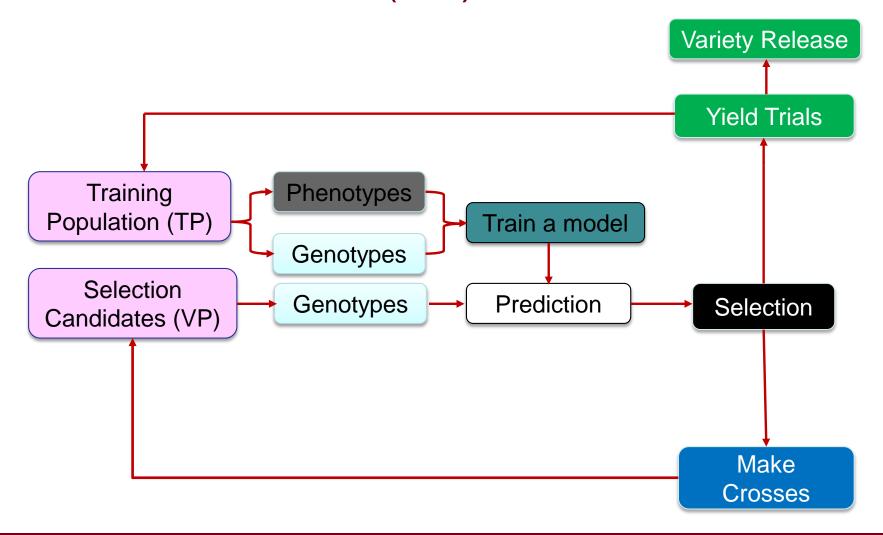


#### Genomic Predictions to Advance FHB Resistance Breeding in Barley

Comparing Phenotypic Selection to Genomic Selection Optimizing Training Population Realized Gains Predicting Variance

Photo: Brian Steffenson

#### Genomic selection (GS)

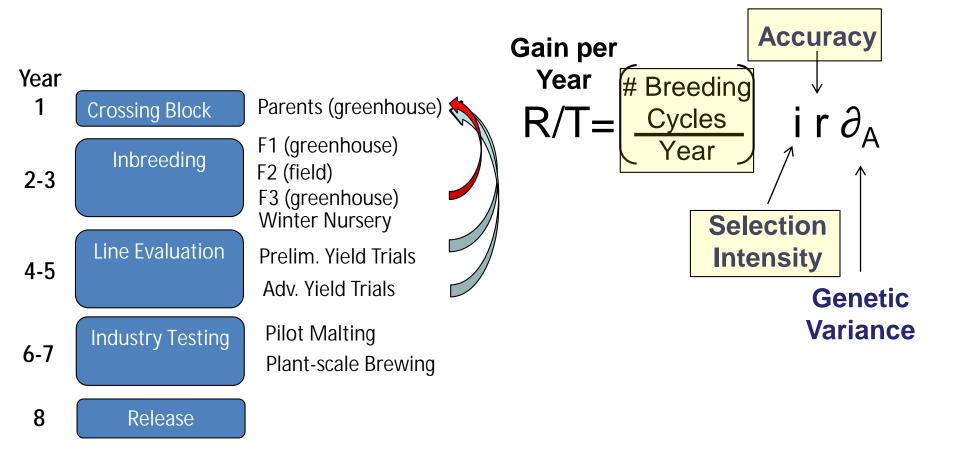


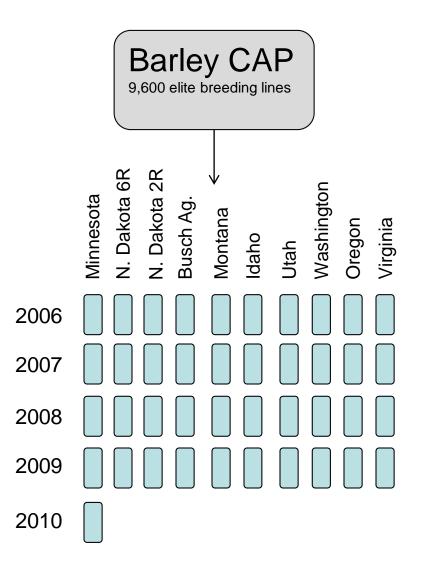
## Genomic Prediction Accuracy

Marker Density / Distribution Training Population Size Prediction Model

Training Population Composition
Training Population Environments (GxE)

### Genomic Selection





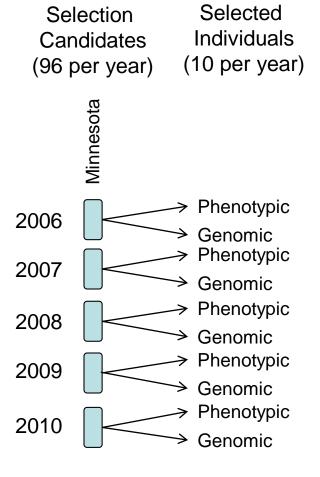
#### **Barley CAP**

10 Breeding Programs96 entries per program per year4 years

Genotyped: 3,072 SNP markers

Phenotyped: over 40 traits (FHB and DON)

#### Comparing Phenotypic (PS) to Genomic (GS) Selection



#### **Phenotypic Selection**

- 2 trials in misted/inoculated nurseries
- 2 replicate per trial

#### **Genomic Selection**

Training Pop 186 breeding lines (parents to SC) 3 trials in misted/inoculated nurseries 1 replicate per trial

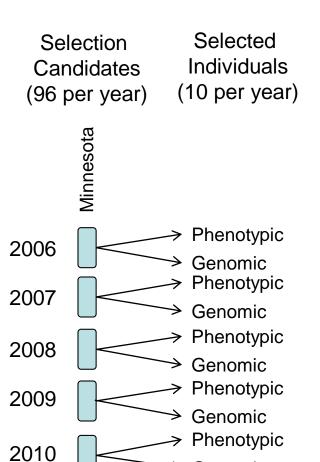
3072 SNP markers

**RR-BLUP** 

#### **Comparison Experiment**

- 4 trials in misted/inoculated nurseries
- 1 replicate per trial wit repeated checks

#### Comparing Phenotypic (PS) to Genomic (GS) Selection

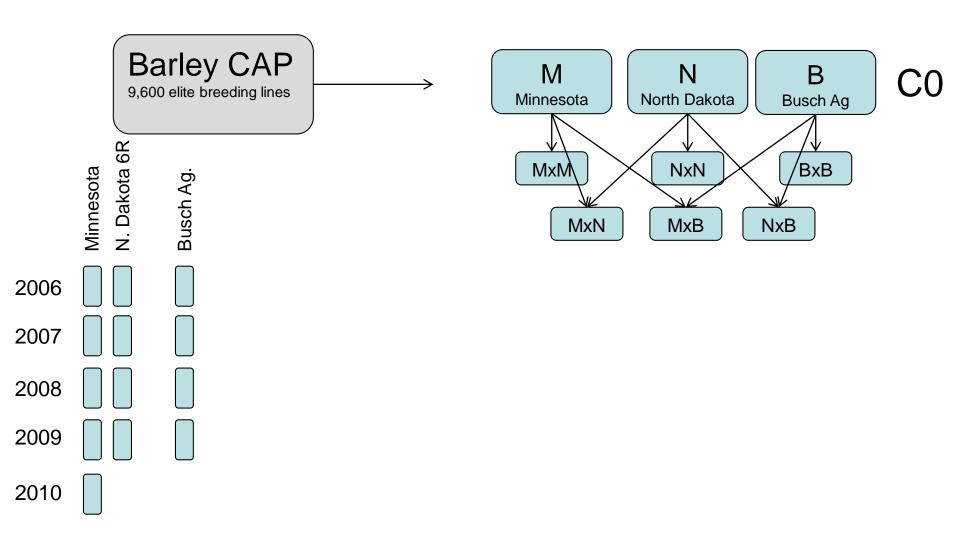


Genomic

Selection Method	FHB severity (%)	DON (ppm)
Sel. Candidates	19.6a	29.1a
Genomic	17.1 b	22.3 b
Phenotypic	17.5 b	23.5 b

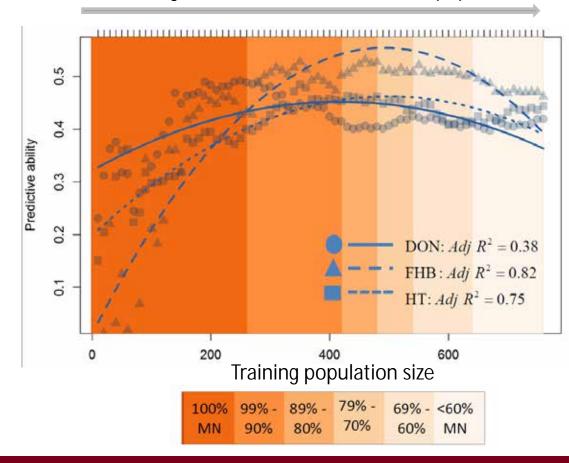
Sallam et al., 2016 (in review)

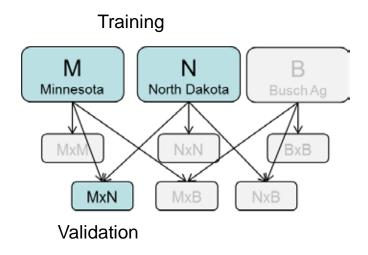




#### **Training Population Composition**

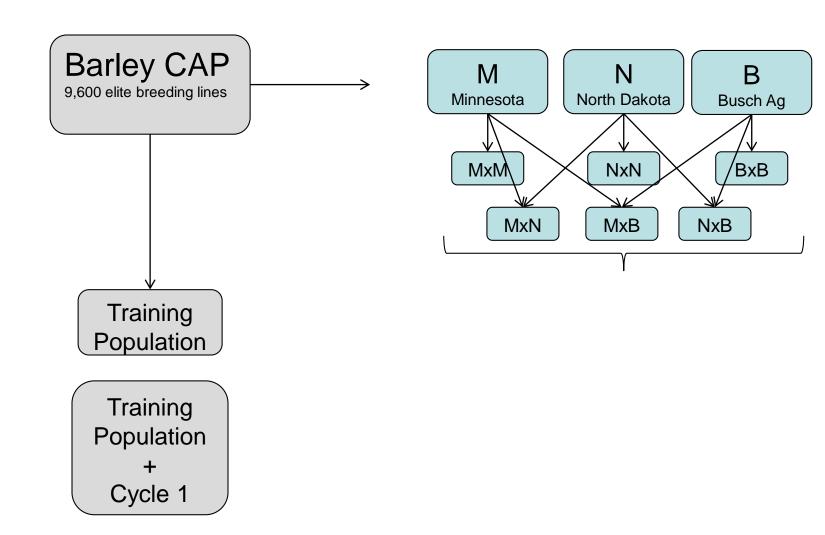
#### Decreasing relatedness to the validation population



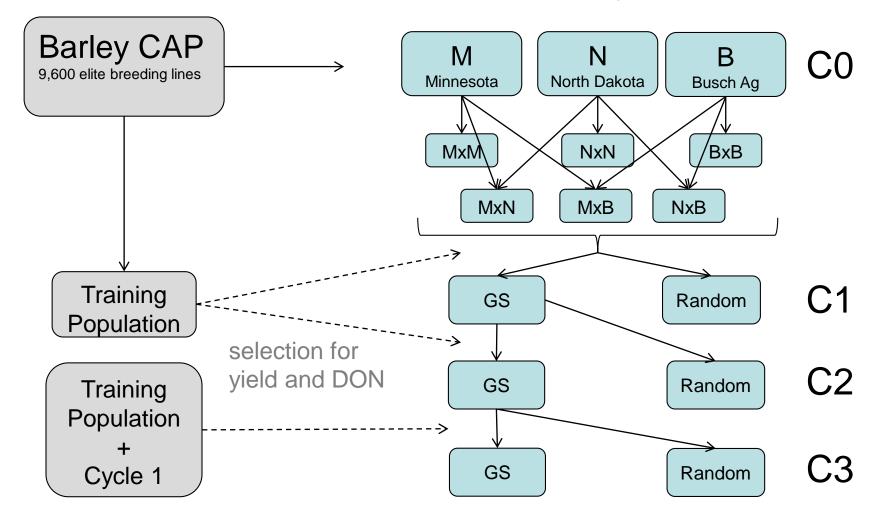


Lorenz and Smith, 2015 Crop Sci

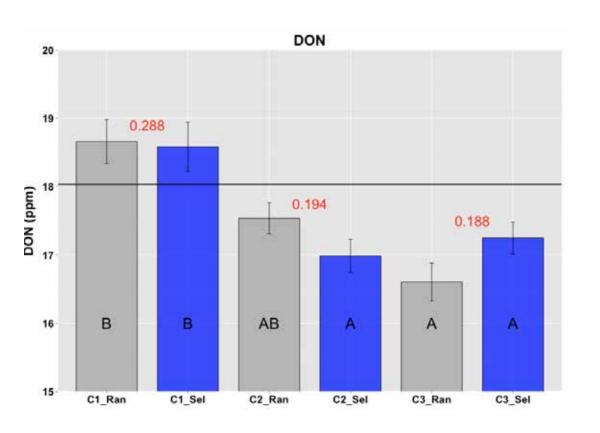


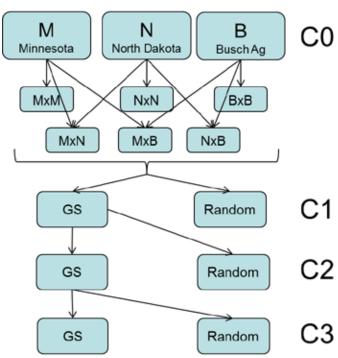


### Gain From Selection (3 cycles)

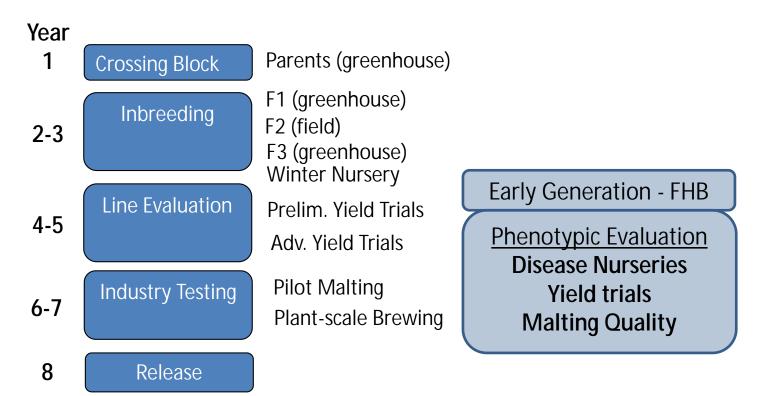


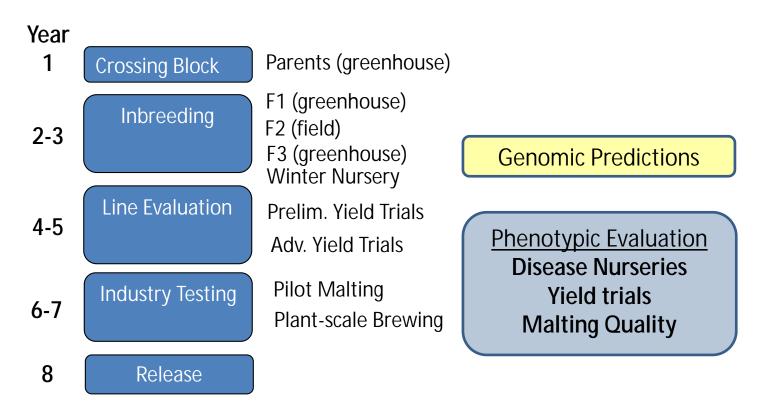
### Gain From Selection for DON (3 cycles)

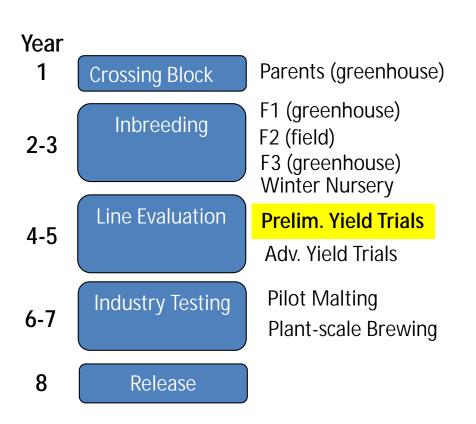


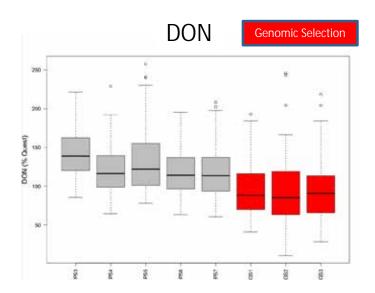


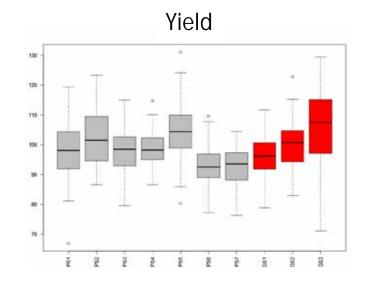
Tiede et al, (in prep)

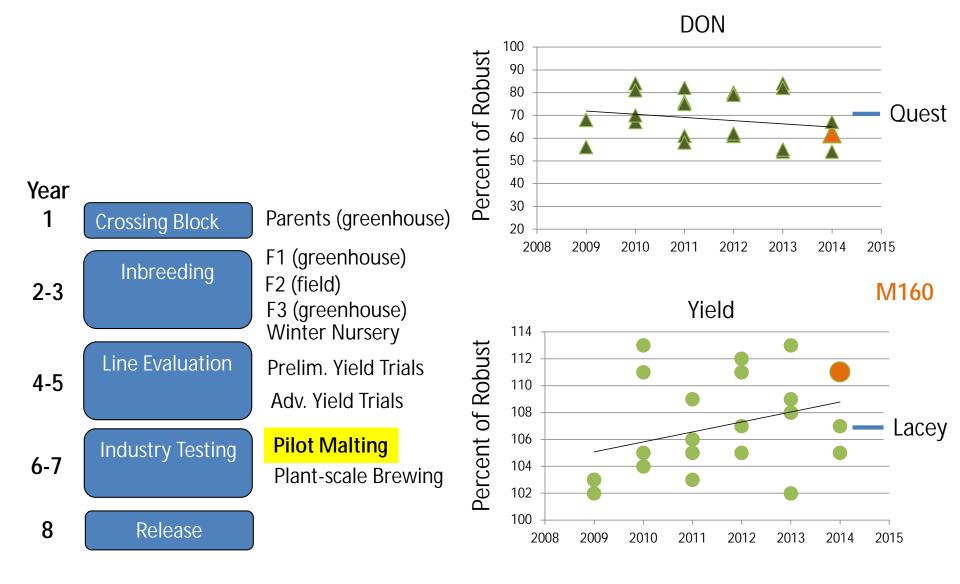




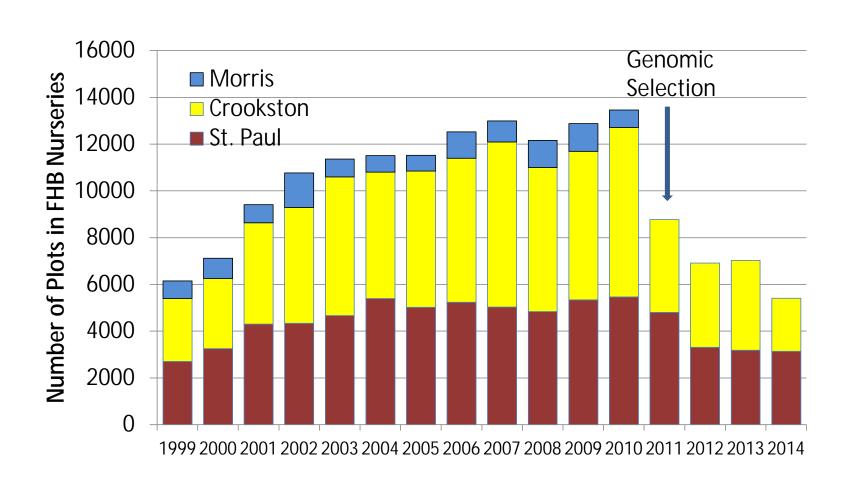




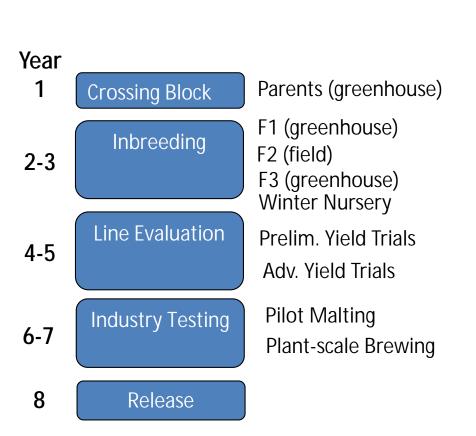


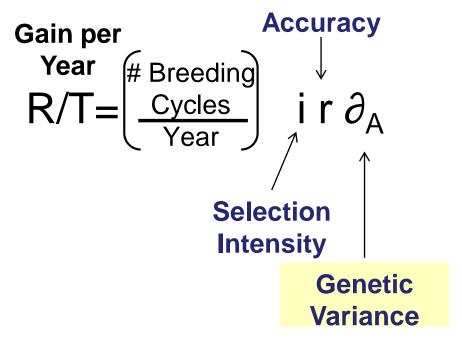


# Genomic selection has reduced the field resources for disease screening

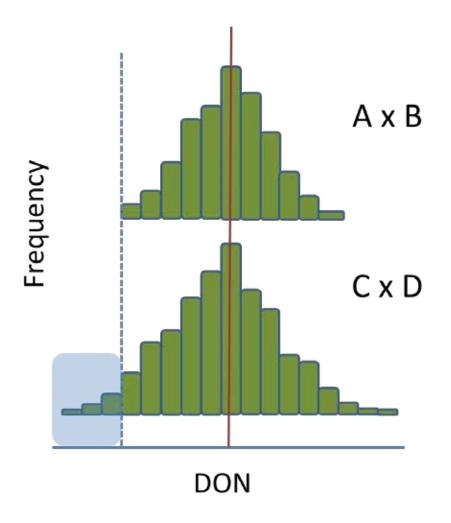


### **Genomic Selection**





#### Maximizing Genetic Variance



Both crosses have the same mid-parent and population mean

.... but different variances

### Maximizing Genetic Variance

# PopVar: R package that predicts the genetic variance of crosses between two parents

- Training Population -> estimate marker effects
- Genotype Parents
- Simulate RIL progeny for each cross combination
- Obtain GEBV for each progeny
- Calculate predicted variance & mean of best 10%
- Calculates correlation among GEBVs for multiple traits

Mohammadi et al., 2015 Crop Sci 55:2068–2077.

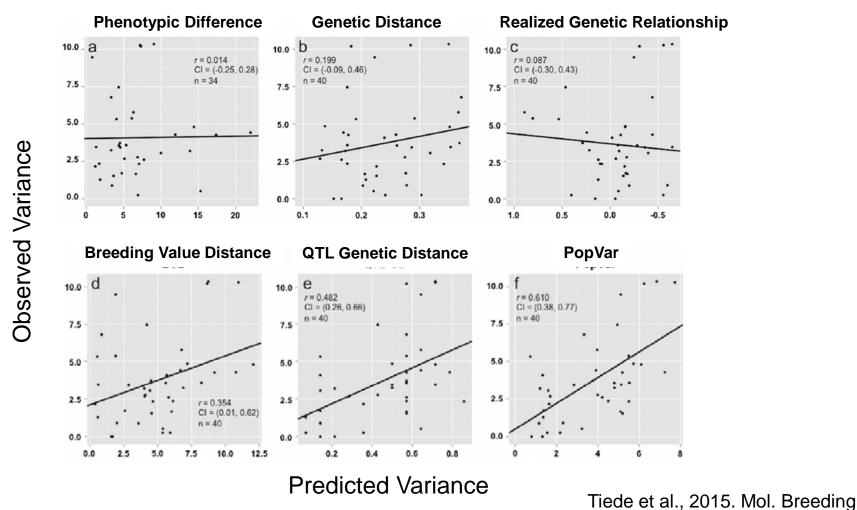
#### Predicting the Genetic Variance of FHB Severity

40 bi-parental breeding populations evaluated 2003–2010 FHB severity data collected in misted and inoculated nurseries 2 Locations, 2 reps per location, RCB design



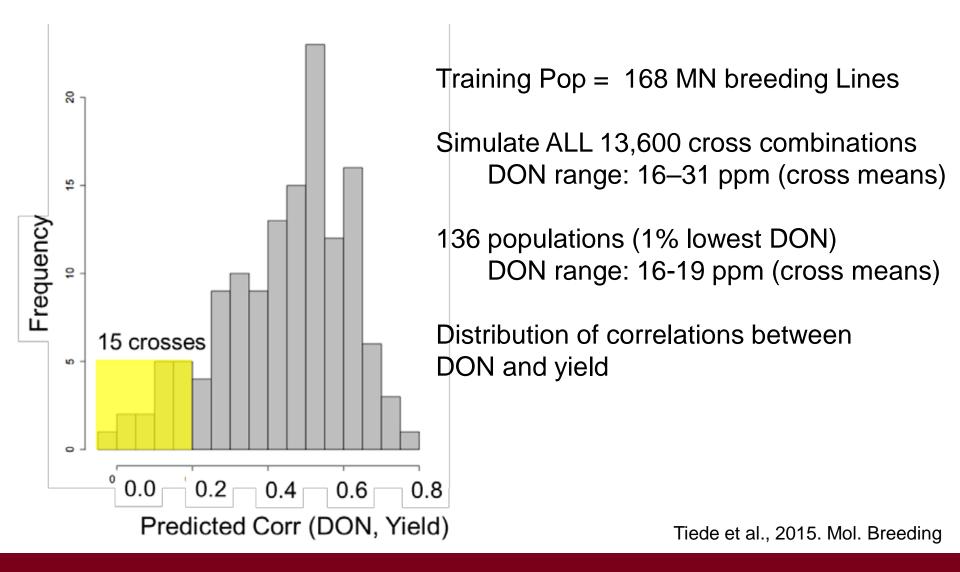


#### Predicting the Genetic Variance of FHB Severity





#### Using PopVar to reduce negative trait correlation



#### Take Home

- GS is about as effective as PS for selecting for FHB severity and DON
- Optimizing the training population greatest opportunity to improve selection accuracy
- Implementing GS allows reallocation of resources
- Potential to use genome-wide marker effects to predict favorable parent combinations.
- Future expanding the number of traits that are predicted

#### **Project Members / Key Collaborators**

#### **Barley Project**

Ed Schiefelbein
Guillermo Velasquez
Karen Beaubien
Jerry Franckowiak
Celeste Falcon
Tyler Tiede
Alex Olhoff
Lu Yin
Jeffrey Neyhart
John Price
Elijah Rasmusson
Thomasina Roth-Krosnoski
Erik Vavra

#### **University of Minnesota**

Gary Muehlbauer Aaron Lorenz Ruth Dill-Macky Brian Steffenson, Yanhong Dong,



#### **Other Institutions**

Richard Horsley, NDSU; Shiaoman Chao, USDA Fargo; Joshua Butler, Busch Ag. Res., Inc Jean-Luc Jannink, USDA



# **Funding**

U.S. Wheat & Barley Scab Initiative









#### **SMALL GRAINS INITIATIVE**



Minnesota Agricultural Experiment Station



