





## Cap Dev Lectures series: Breeding autogamous cereals a complete lecture from Parents to Farms

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International Center for Agricultural Research in the Dry Areas

## Lecture 5 – April 21<sup>st</sup> 2022

- Improving parental selection by genomics (L1)
  - GWAS to combine QTLs
- Improving inbreeding advancement by genomics (L2)
  - MAS for untestable traits
  - GS for recycling time
  - Risk of loosing intensity
- Improving Stage 1-2 yield trials by genomics (L3)
  - Higher accuracy with genomic GxE models
- Improving Release and certification by genomics (L4)
  - Selection for homogeneity by markers
  - PVP by genetic passports
- Improving Genetic gain by genomics
  - The equation and how it changes



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- Parental selection can not be blind
  - How would you use genomic to improve this step?

• Select 2 parents for a new top yielding var with resistance to 2 diseases and good quality

Trait	P1	P2	P3	P4	New var
Yield	5	5	6	6	>5
Disease 1	R	S	R	S	R
Disease 2	S	R	R	S	R
Quality	Υ	Y	Ν	Ν	Υ

#### Phenotypic data

• Select 2 parents for a new *top yielding var with resistance to 2 diseases and good quality* 

New var

>5

R

R

Y

• Now you have GWAS data: which one would you combine?

#### Phenotypic data

r	Trait	P1	P2	P3	P4
	Yield QTL 1	Ν	Y	Y	Y
	Yield QTL 2	Y	Ν	Y	Y
	Disease 1 QTL 1	Y	Ν	Y	Ν
	Disease 2 QTL 2	Ν	Y	Y	Ν
	Quality QTL 1	Y	Ν	Ν	Ν
	Quality QTL 2	Ν	Y	Ν	Ν

**Genomic data** 

Trait	(P1)	P2	(P3)	P4
Yield	5	5	6	6
Disease 1	R	S	R	S
Disease 2	S	R	R	S
Quality	Y	Υ	Ν	Ν

- Parental selection can be improved via GWAS and also based on *genetic diversity*
- Ideally we also want to know the linkage between traits (neg or pos correlation)
  - PopVar is a R script that can be used for this scope



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## Progress to inbreeding: genomics augmentation

• How do you think molecular markers can improve this step?

# Progress to inbreeding: MAS

• Marker Assisted Selection:

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- Traits controlled by 1 or few genes can be selected by MAS
- That is why is very important to "validate" GWAS discoveries (example of Hessian fly)



Allele	Туре	HF response	N	Ratio
Suscept.	TN	Susceptible	610	93%
	FN	Resistant	45	7%
Resistant	FP	Susceptible	0	0%
Nesistant	ТР	Resistant	32	100%

*Bassi et al. 2019 Doi:* 10.1007/s11032-019-0927-1 8

## Progress to inbreeding: MAS

- Marker Assisted Selection:
  - Keep in mind the issue of *intensity vs inbreeding* 
    - Screening 12 progenies in F4 is ok, while at F6 it would fail
      - Why not testing 192 then?



# Progress to inbreeding: GS

• Genomic models are used to predict the performances of untested material

- It can improve recycling time
- It can improve accuracy
- It certainly cost **intensity**: KASP with 96-192 markers: 5 USD per sample 100 populations x 48 progenies + 200 parents as TP = **25K USD per year**



icarda.org Zaim, et al. 2021. Combining QTL analysis and genomic predictions for four durum wheat populations under drought 10 conditions. Frontiers.

## GS Approach: keeping intensity is difficult

- Using 48 progenies it is possible to select up to **10 loci** per selection cycle in F2 to F4
  - A *recurrent mating scheme* allows to select 10 loci each cycle, hence **30 in 3** cycles



# Training population: Relatedness is *accuracy*

A training population can only model those alleles for which it is <u>segregating</u> "Far-related" training populations can not predict the allelic interactions of the breeding pop

• Training pop in full-sib relationship with breeding pop is ideal to ensure high accuracy



icarda.org Zaim, et al. 2021. Combining QTL analysis and genomic predictions for four durum wheat populations under drought 12 conditions. Frontiers.

## Breeding GS schemes: cost vs genetic gain



Bassi, et al. 2014. Breeding schemes for the implementation of genomic selection in wheat (Triticum spp.) Plant Science.



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Shuttle **F5 F3** 0.5 0.5 Accuracy 0.2 Intensity 1 1.5 1 Heritability Time 5 3 <u>2</u> GG (U) 6x 7x 4x P. positive 2% 3% 6%

Bassi, et al. 2014. Breeding schemes for the implementation of genomic selection in wheat (Triticum spp.) Plant Science.

# • What elem

# Stage 1 and 2 yield trials: Genomic augmentation

• What elements could be improved?

## Stage 1 and 2 yield trials: Genomic augmentation

- What elements could be improved?
  - Assessment of complex traits with high accuracy: index
  - GxE at Stage 1



Phenotypic predictions of one site (TPE) to another are very poor: GEBV are better at controlling GxE A Stage 1 training pop has limited ability to predict real performances

## Stage 1 sparse testing: Genomic augmentation

- Stage 1 is composed of thousands of lines
  - Increasing intensity results in genetic gain
  - But it comes at extra cost
    - What if we could plant few and predict many?

Ratio (%)	Stage 1	Stage 2	STD differential	Genetic gain
10.0%	2 400	240	1.750	1.00
5.0%	2 400	120	2.063	1.18
1.0%	2 400	24	2.665	1.52
0.5%	2 400	12	2.892	1.65
0.1%	2 400	2-3	3.253	1.86



## Stage 1 sparse testing: Genomic augmentation

- Stage 1 is composed of thousands of lines
  - Increasing intensity results in genetic gain
  - But it comes at extra cost
    - What if we could plant few and predict many?
- If genotyping cost is less than plot cost, it is a good approach
  - Otherwise, only gain will be accuracy, not intensity
  - Keep in mind the need to produce seeds for Stage 2

Ratio (%)	Stage 1	Stage 2	STD differential	Genetic gain
10.0%	2 400	240	1.750	1.00
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# Release and certification: Genomic augmentation

• What elements could be improved?

# Release and certification: Genomic augmentation

- What elements could be improved?
  - Homogeneity
    - Use markers to accelerate homogeneity
    - Use markers for DUS testing
  - Intellectual property protection
    - Use genetic passports to investigate use of germplasm

Use of genetic markers for the detection of off-types for DUS phenotypic traits in the inbreeding crop, barley

Benedetta Saccomanno • Margaret Wallace • Donal M. O'Sullivan • James Cockram

Var name	Origin	Karim	Duilio	Meridiano
		Jori69/Anhinga/Flamingo	Cap.lli//Anhinga/Flamingo	Sim.to/WB881/Duilio/F21
Karim	CIMMYT		<u>0.92</u>	0.62
Dulio	Italy	<u>0.92</u>		0.62
Nassira	Morocco	0.64	<u>0.66</u>	0.42
Don Pedro	Spain	0.63	<u>0.64</u>	0.42
Wollaroi	Australia	0.61	<u>0.63</u>	0.41
Meridiano	Italy	0.60	<u>0.62</u>	
Flaminio	Italy	0.57	0.56	<u>0.58</u>

# Connecting the dots: molecular breeding



• 25K array is used to genotype Stage 2 and MP



# Connecting the dots: molecular breeding



- 25K array is used to genotype Stage 2 and MP
- F4 are genotyped with 196 KASP
- KASP match 25K array probes
- New KASP are added/removed



## How can genomic augmentations help us?

Genetic gain =

Accuracy × Heritability × Selection intensity

**Recycling time** 

Accuracy: Heritability: Intensity: Time:

reduce experimental error
markers are not affected by GxE
predict more entries without testing them
predict early and then recombine

## **Conclusion lecture 5**

- The genetic gain equation drives all breeding decisions
- Markers help selecting parents via GWAS and genetic diversity
- MAS/GS assure GG in F2>F5, but need to find balance: relatedness vs time vs accuracy vs costs
- GS in Stage 1 can increase accuracy (GxE), helps deploy index, and can increase intensity
- Markers can simplify the progress to homogeneity
- Genetic passports are good records to protect IP
- Genotyping is a **cost** and should be treated as such
- Molecular breeding has its own **logistics** to be considered:
  - Few seeds in plants grown in the GH
  - Crossing 1 plant is hard, 10 plants is better
  - Time gap between genotyping and modeling has to be accounted for

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