

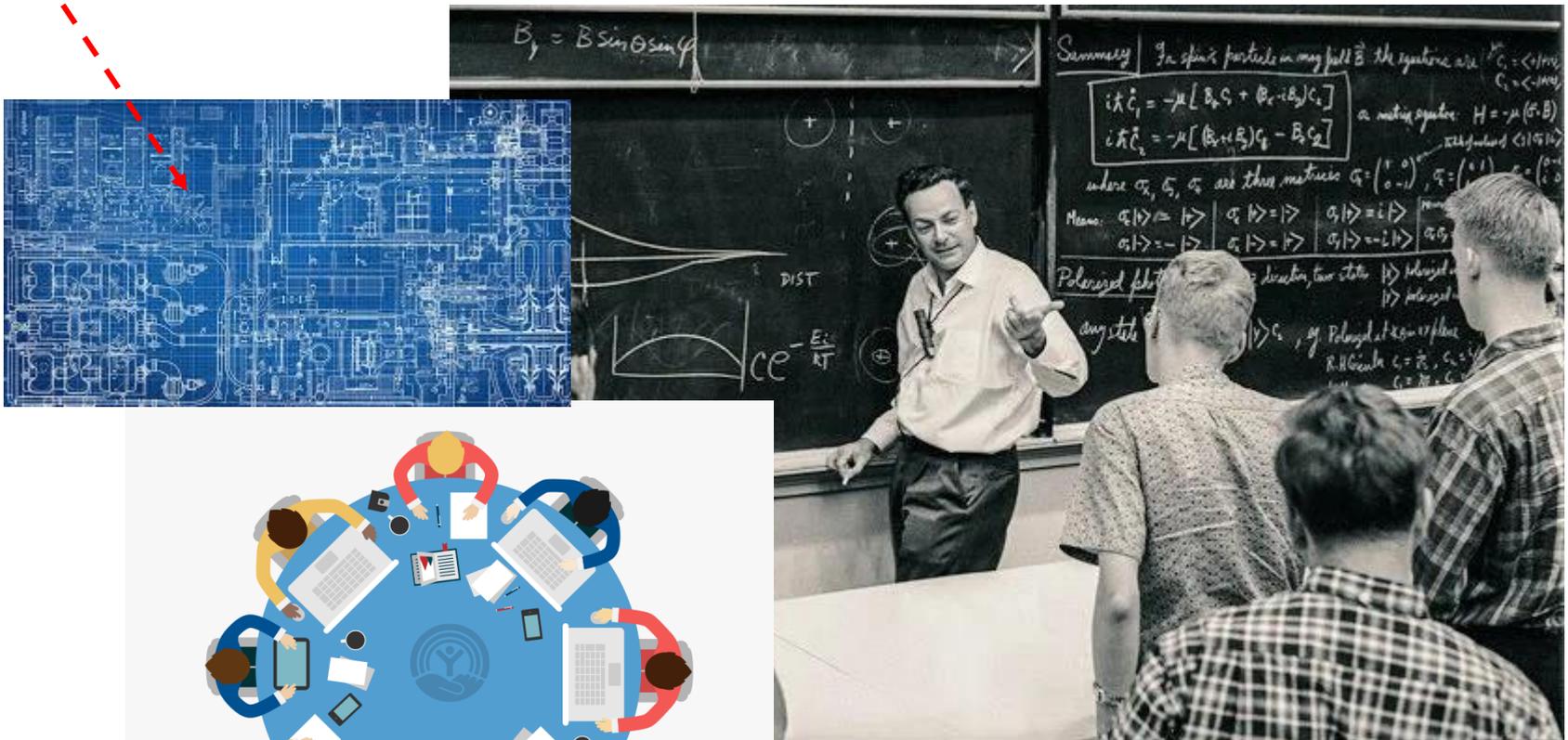
Rate of response to selection (genetic gain): concept, methods and recommendations

November 18th, 2020

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Breeding scheme optimization lead



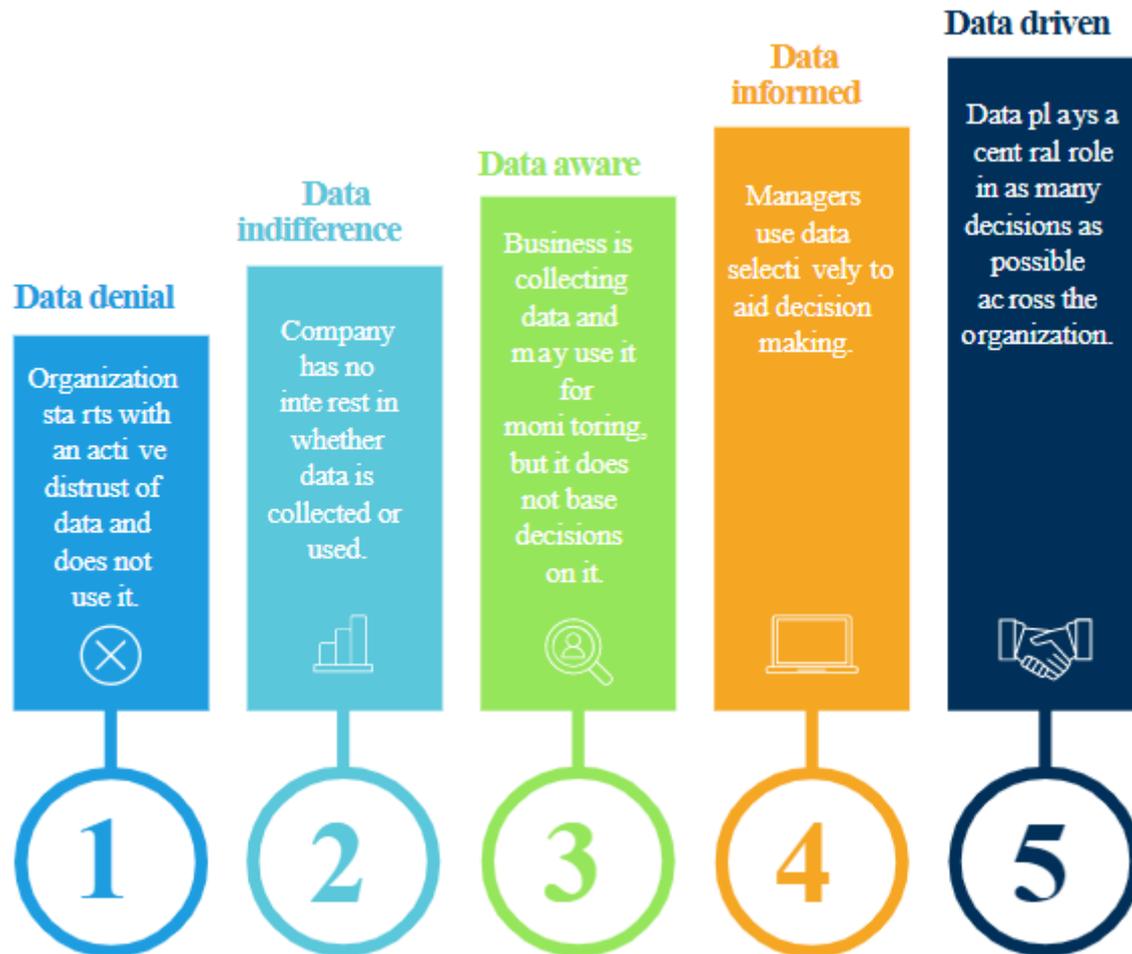
You better ask if there's doubts



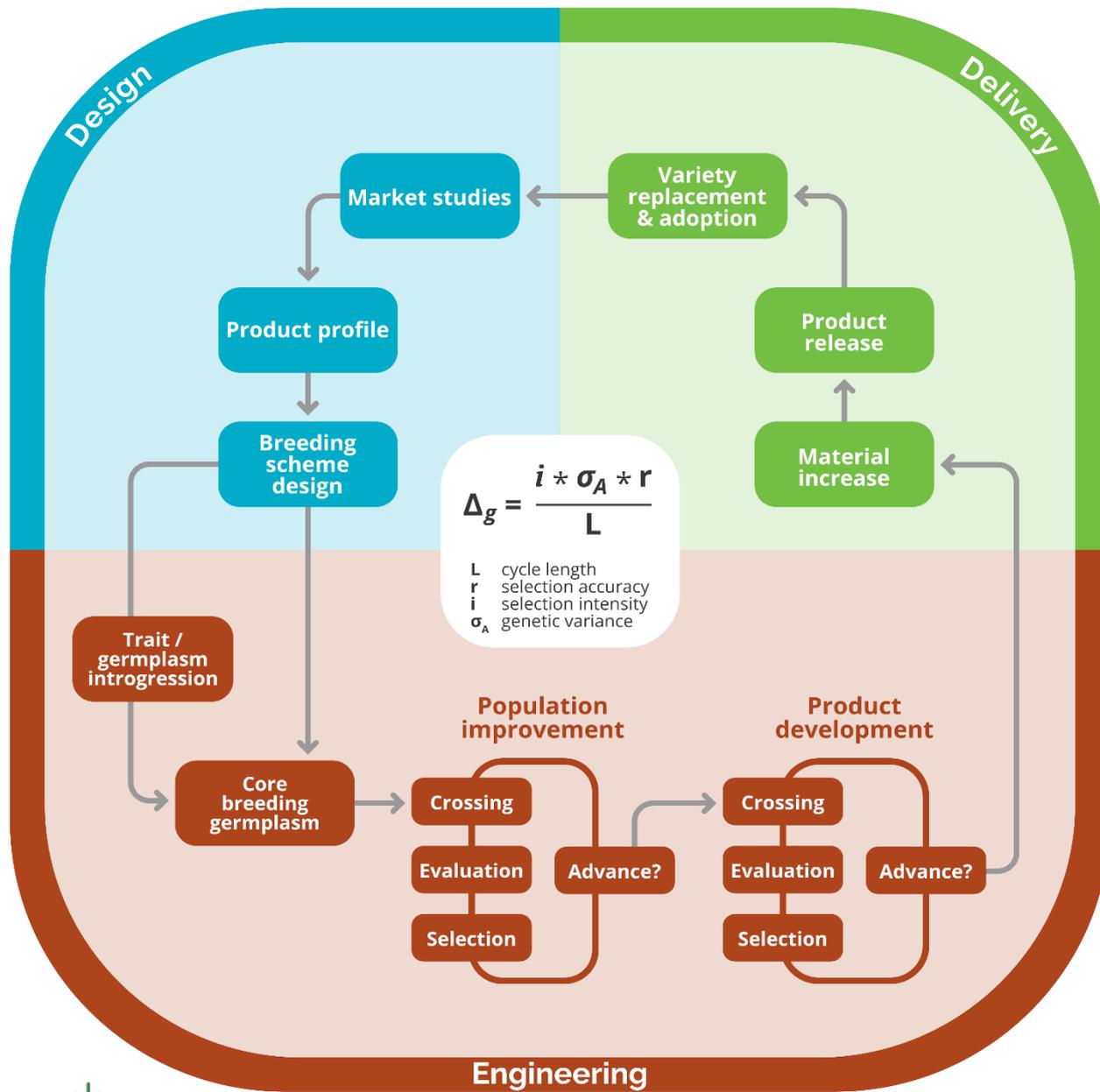
Outlook

- Breeding as a process
- What is genetic gain?
- Decisions for estimating genetic gain
- Guidelines for calculation

Why we want to estimate the rate of genetic gain?



Five stages towards data drive culture



Breeding as a process

Any process requires KPIs

Breeding as a process

Engineering KPIs

Process level



Design

Market-driven product design specifications



Engineering

The optimum way to develop the product



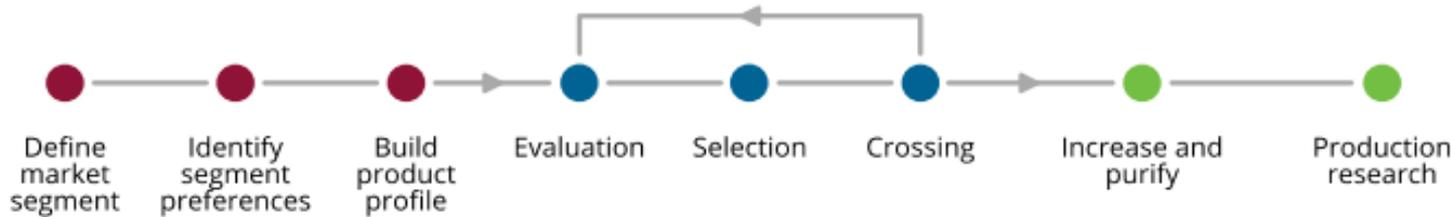
Production

Commercial quality and scale

Genetic gain at process level

- Δ_g in program (R&E)
- Δ_g in farmers' fields

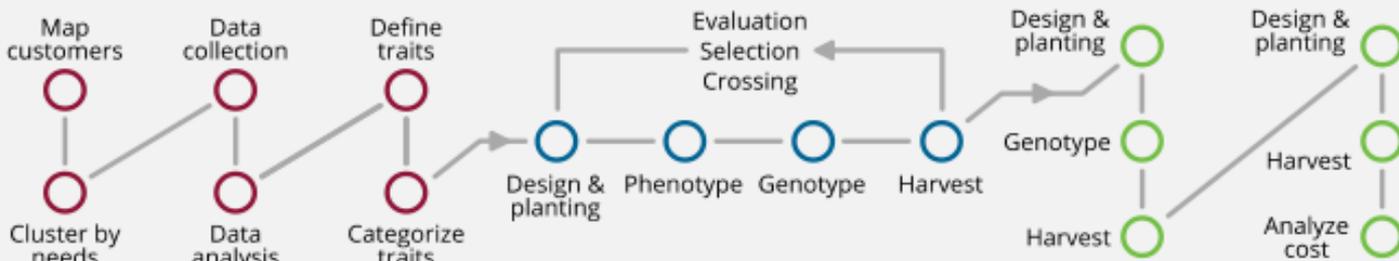
Sub-process level



Genetic gain at sub-process level

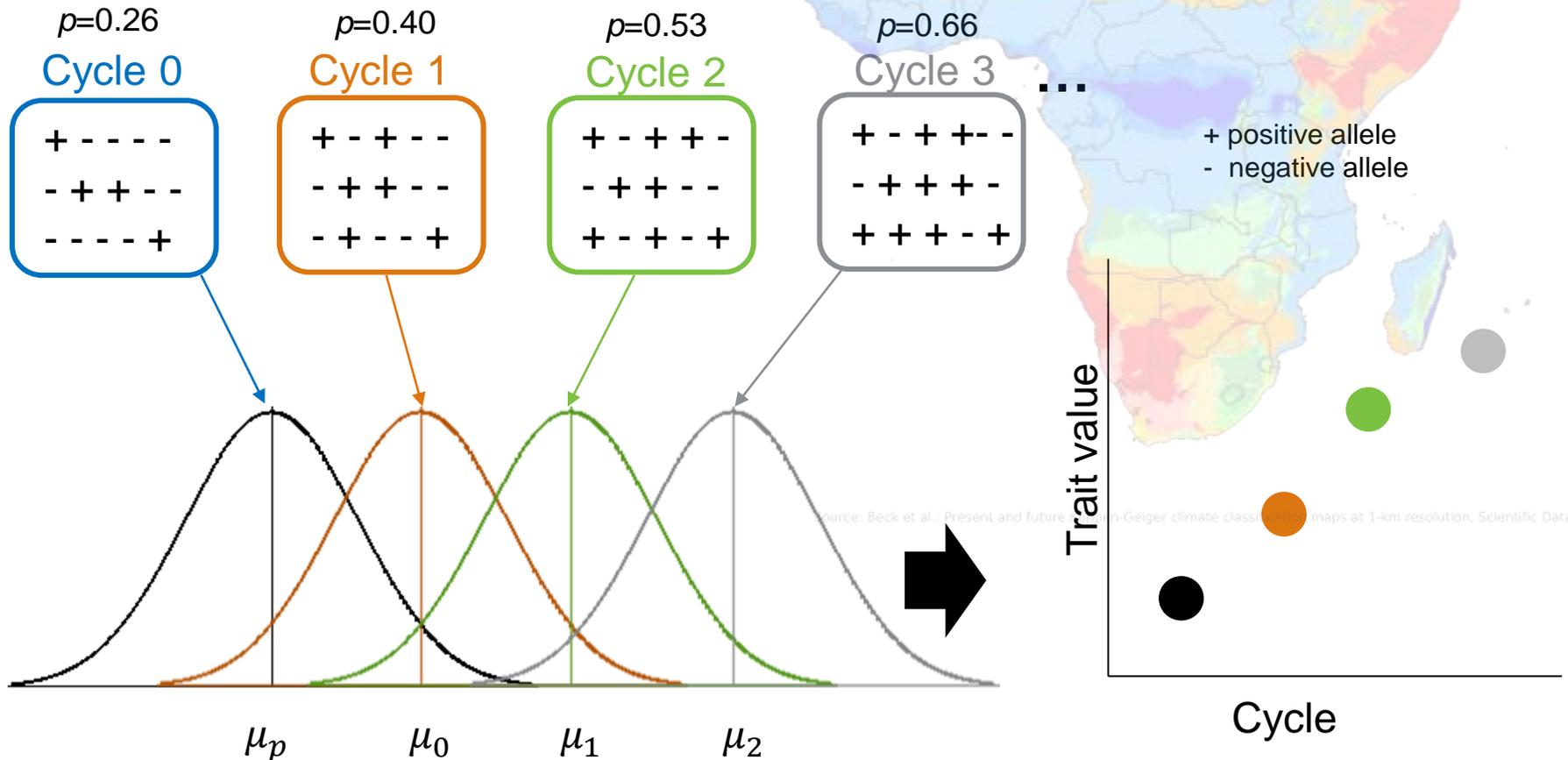
- r_g in program (R&E)
- h^2 in farmers' fields

Sub-subprocess level



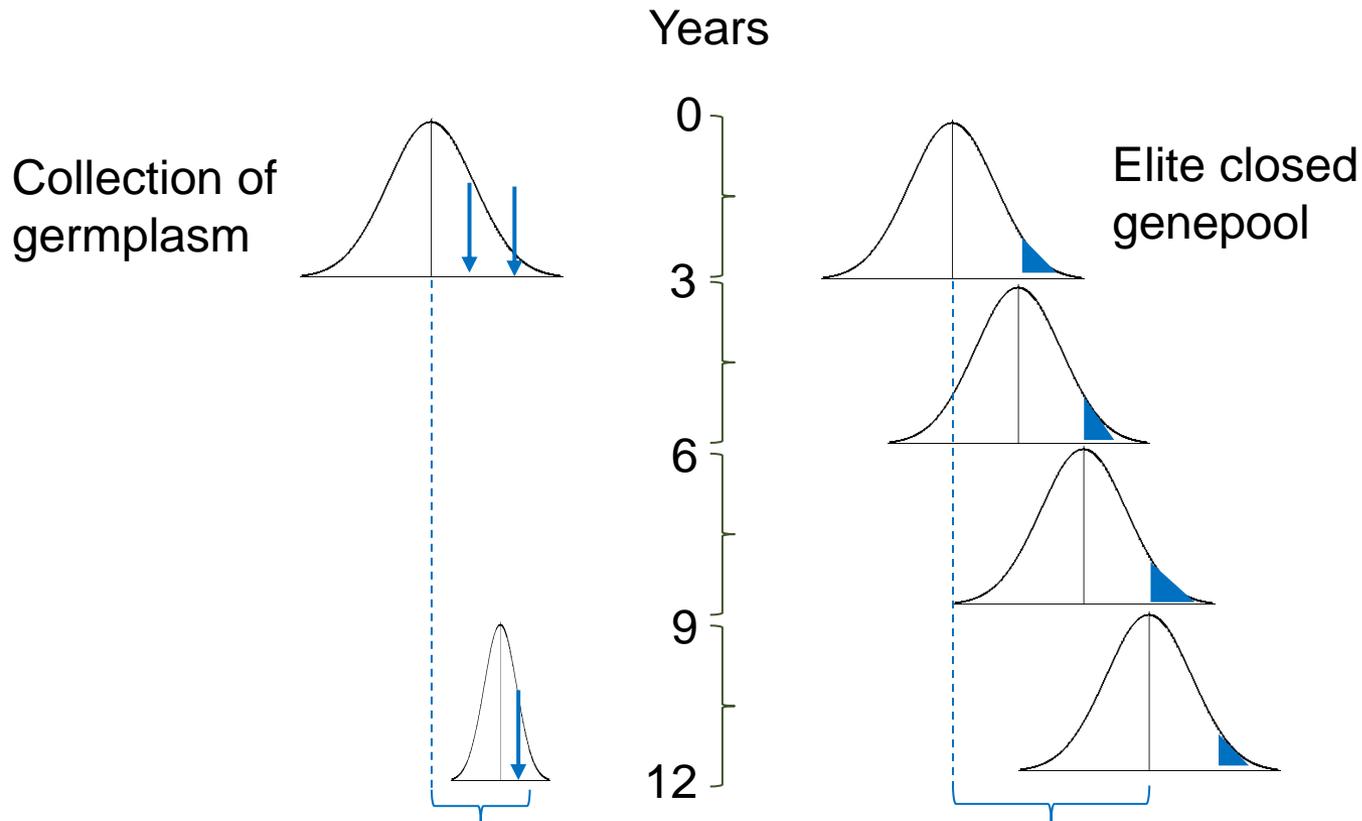
Genetic gain at sub-subprocess level

Where genetic gain comes from?



- 1) Understood in a population basis
- 2) Comes from a recurrent selection system (relatively closed system)
- 3) Comes from the increase in the frequency of alleles responding positively to natural and/or simulated environmental conditions.

What is not “rate of response to selection” or “genetic gain”?





OR



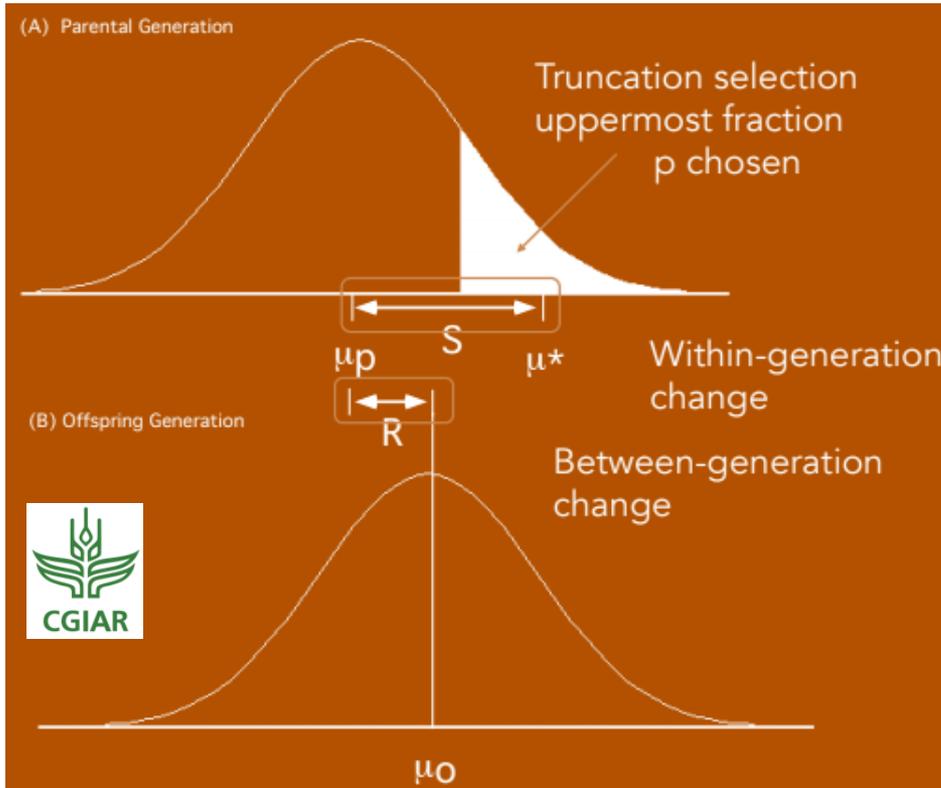
How to derive the expected genetic gain?

1a) $y = \mu + g + e$

1b) $g = b(y - \mu)$

Assuming no dominance, $g = a$ (the additive component), remember that:

1c)
$$b = \frac{cov(a, y)}{var(y)} = \frac{cov(a, a + e)}{var(y)} = \frac{cov(a, a)}{var(y)} = \frac{var(a)}{var(y)} = h^2$$



If we apply selection (*) in the parental generation based in 1b):

1d) $g_* = \mu_p + b(y_* - \mu_p)$

Take the expectation for individuals selected in 1d)

1e) $\mu_0 = \mu_p + b(\mu_* - \mu_p)$

Re-arrange

1f) $\mu_0 - \mu_p = b(\mu_* - \mu_p)$

Remember 1c) and:
 $S = \mu_* - \mu_p$

1g)

1h) $R = h^2 S$

Now remember:

1i)
$$h^2 = \frac{\sigma_a^2}{\sigma_y^2} = \frac{\sigma_a \sigma_a}{\sigma_y \sigma_y} \quad \& \quad i = \frac{S}{\sigma_y}$$

1j)
$$R = \frac{\sigma_a \sigma_a}{\sigma_y \sigma_y} S = \frac{\sigma_a \sigma_a}{\sigma_y} i = h \sigma_a i = r \sigma_a i$$

How to derive the realized genetic gain?

	Input	Method	Output																														
 <p>Data generation</p>	Target segment information + Resources + Breeding scheme	Breeding trials method OR Era trial method	<table border="0"> <tr> <td></td> <td>STG1</td> <td>STG2</td> <td>...</td> <td>On-farm</td> </tr> <tr> <td>Y1</td> <td></td> <td></td> <td>...</td> <td></td> </tr> <tr> <td>...</td> <td></td> <td></td> <td></td> <td></td> </tr> <tr> <td>Yn-1</td> <td></td> <td></td> <td>...</td> <td></td> </tr> <tr> <td>...</td> <td></td> <td></td> <td></td> <td></td> </tr> <tr> <td>Yn</td> <td></td> <td></td> <td>...</td> <td></td> </tr> </table>		STG1	STG2	...	On-farm	Y1			...		...					Yn-1			...		...					Yn			...	
		STG1	STG2	...	On-farm																												
	Y1			...																													
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Yn			...																														
Data reduction	Phenotypes	Trait(s) + Stage + Time	<table border="0"> <tr> <td></td> <td>Y1</td> <td>...</td> <td>Y10</td> </tr> <tr> <td>On-farm</td> <td></td> <td>...</td> <td></td> </tr> </table>		Y1	...	Y10	On-farm		...																							
	Y1	...	Y10																														
On-farm		...																															
Data analysis	Phenotypes subset	Linear models $y = X\beta + Zu + e$	<table border="0"> <tr> <td></td> <td>Across-years</td> </tr> <tr> <td>On-farm</td> <td></td> </tr> </table>		Across-years	On-farm																											
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Realized gain calc	Adjusted means	Linear models $y = X\beta$	<table border="0"> <tr> <td>ΔG</td> <td>  </td> </tr> </table>	ΔG																													
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Methods for calculating the rate of genetic gain

Table 1. Summary of methods to estimate the response to selection and rate of genetic gain.

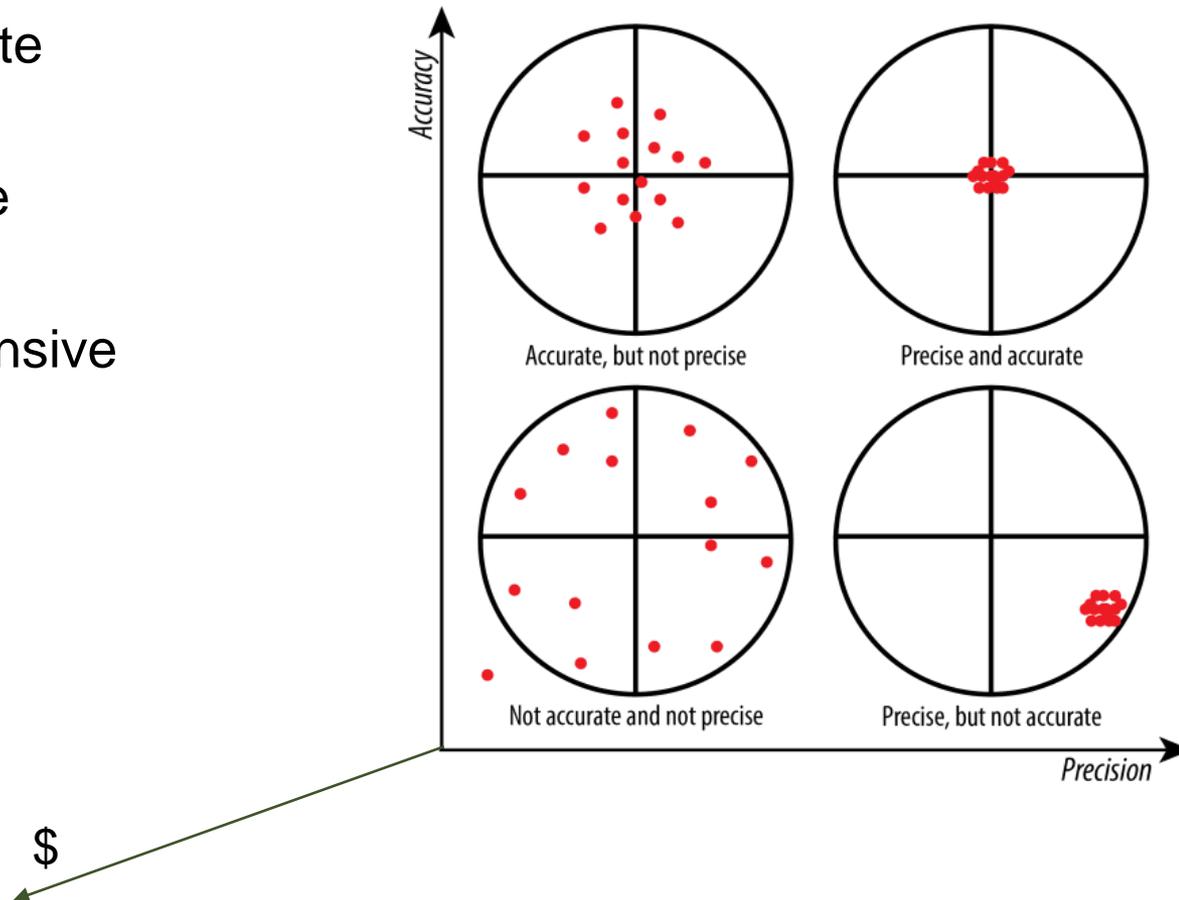
Method	Formula used	Data required	Sample	Factors to be considered	Connectivity** / TPE* coverage	Recommendations
Expected	$R = h^2 S = i r \sigma_g$ Lush (1942), Burrows (1972), Walsh (2004)	Any trial information	Any generation material	The heritability used will have an important effect in under- or over-estimating the metric.	Low after first selection cycle.	Use across-trials heritability. Do not use to take complex decisions
Realized	$y = X\beta + Z_d u_d + Z_g u_g + \varepsilon$ β : vector of fixed effects. u_d, u_g, ε : vector of random non-genetic, genetic and error effects. X, Z_d, Z_g : incidence matrices connecting observations with vectors of fixed and random effects	Era trial information	Early generation material Advanced material Released varieties On-farm	1) TPE* coverage is low (usually some locations & a couple of years). 2) Connectivity** among entries is maximum (cohorts evaluated at the same time). 3) Sample can overestimate the metric.	High / Low	1) Evaluate the material in representative environments for more than one year. 2) Use a replicated design. 3) Take a representative sample from each cohort if an estimate of evolution of genetic variance is required.
	Laidig <i>et al.</i> (2014), Piepho <i>et al.</i> (2014), Mackay (2011), Garrick (2010)	Historical trial information	Early generation material Advanced material Released varieties On farm	1) TPE* coverage can be low (early), intermediate (advanced) or high (varieties). 2) Connectivity among entries depends on checks and the use of methods like EBV. 3) Sample can overestimate the metric.	Variable / Low Variable / Intermediate Variable / High Variable / High	1) Use 4-10 checks depending on the stage*** to increase the connectivity of the data. 2) Use early generation trials for better estimate of evolution of genetic variance and advanced material for better estimates of the rate of genetic gain.

Decisions for measuring Δ_G

1. Associated target (Market segment: TPE + product description)
2. Trait of interest [productivity (yield), other trait or an index; BV or genetic value?]
3. Germplasm stage sample (PYT, IYT, AYT, On-Farm, ... trials)
4. Time period of interest (i.e. last 10 years)
5. Method to produce the data (explicit era trial or historical information)
6. Methodology to calculate Δ_G (mixed models; 1-step, 2-step, etc.)

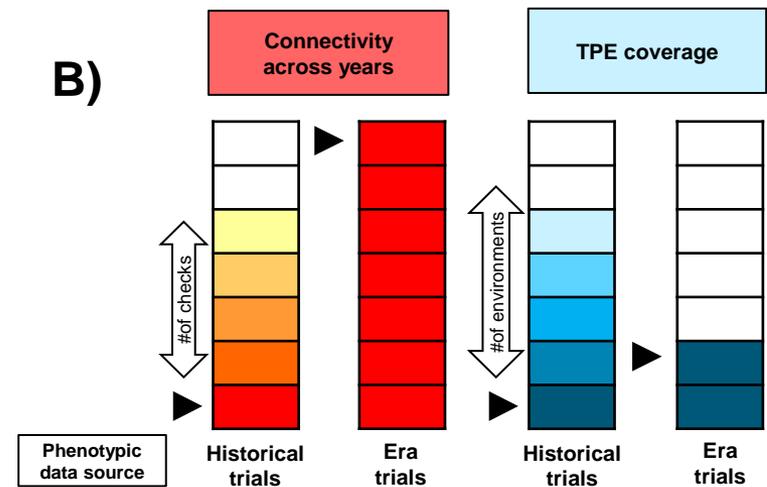
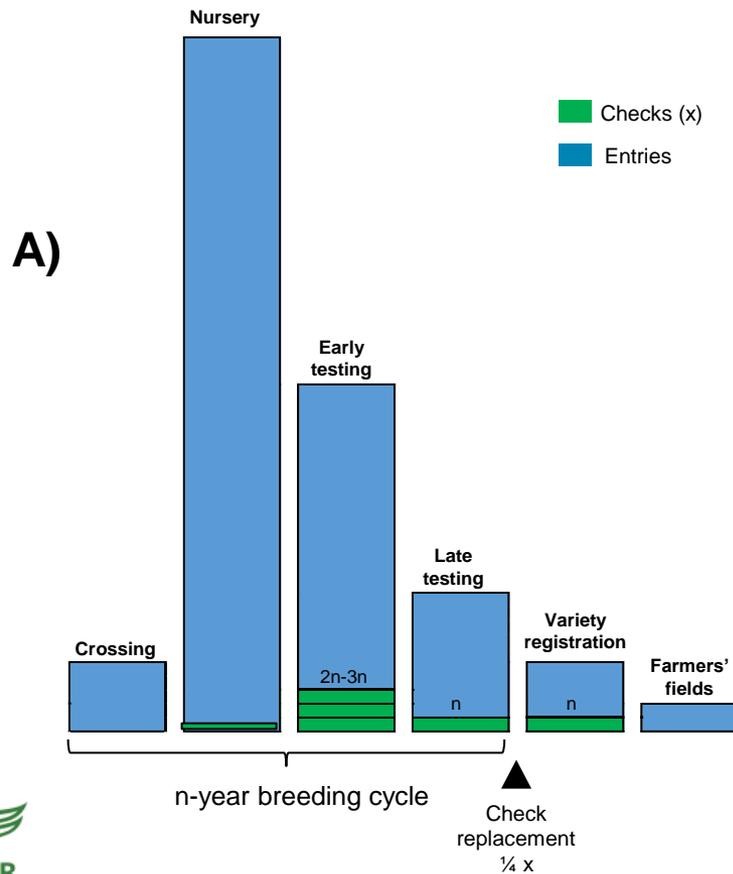
What makes a good estimate of rate of genetic gain?

- Accurate
- Precise
- Inexpensive



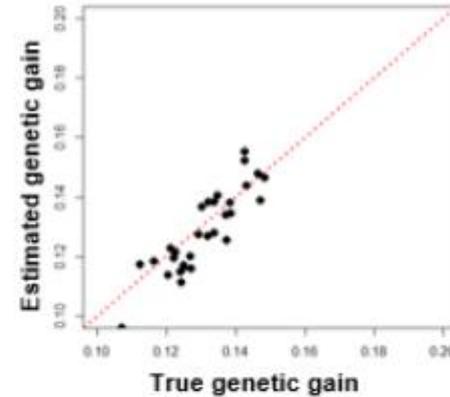
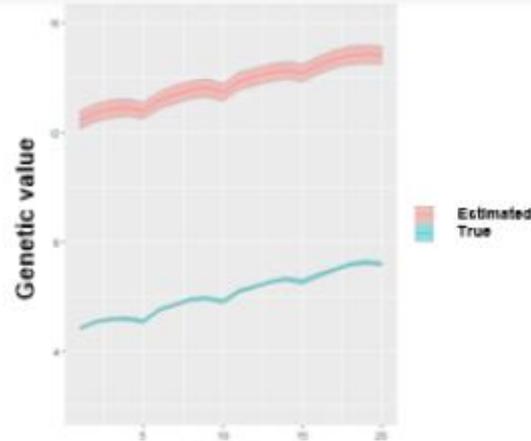
Guidelines to maximize accuracy and precision of Δ_G

- Make sure you can **connect your data (A)** to remove year effects. Make sure you can **cover the target (i.e. TPE) properly (B)**.

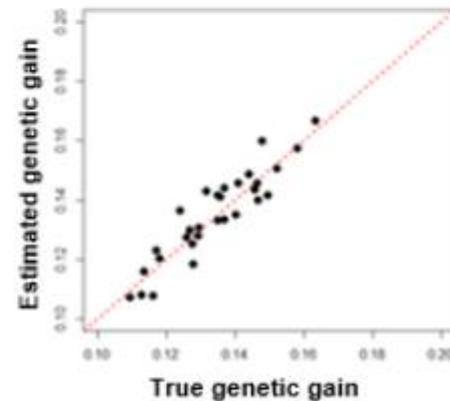
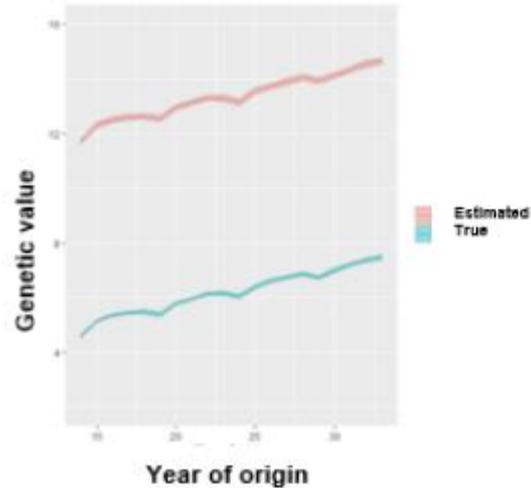


Guidelines to minimize cost of estimating the Δ_G

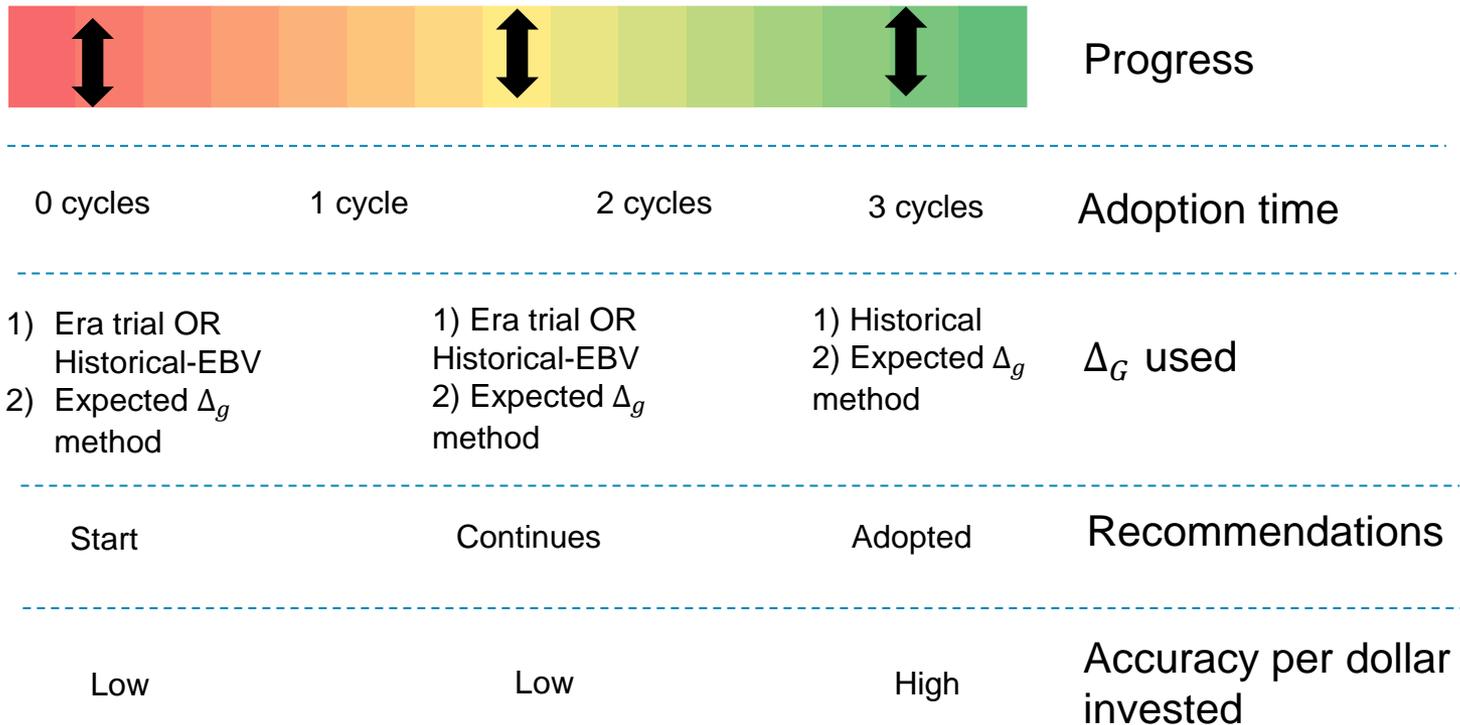
Era trial information
(Advanced)



Historical trial information
(Varieties)



Adoption of Δ_G KPI estimation



**Thank you for your
interest!**



Recipe to calculate Δ_G

1. Define target associated (i.e. Drought conditions environments).
2. Define time period of interest (i.e. last 20 years).
3. Define trait of interest (i.e. yield).
4. Define germplasm stage sample (PYT & AYT).
5. Produce field data:
 1. Run an era trial with PYT&AYT materials from last 20 years under drought environments and record yield.
 2. Collect historical information from PYT&AYT materials from last 20 years under drought environments and extract yield)
6. Fit a MET analysis with a mixed model and extract adjusted means for genotypes.
7. Merge year of origin to the adjusted means.
8. Fit a linear model of the form *adjusted.mean~year.origin*
9. The slope is the rate of genetic gain.

Typical structure of a breeding program

