



TOWARDS IMPROVEMENT OF **RUMINANT** BREEDING
THROUGH **GENOMIC** AND EPIGENOMIC APPROACHES

Genomic selection strategies and their potential to maintain **rare alleles** and **de-novo mutations**

M. Schrauf, J. Vandenplas, H. Mulder & Y. Wientjes



EAAP 2023 – Lyon, France

28th August 2023

Motivation

Rare alleles and de-novo mutations have...

- low correlation with phenotypes at the population level
- usually weak linkage with SNP markers

Led to think that

- genomic selection may not use favorable rare alleles effectively
- could lose rare alleles at a higher rate than pedigree selection

Previous works

Compared **mass** selection, **pedigree** selection and **genomic** selection

Some conclusions about genomic selection:

- inclusion of **own phenotypes** is a main factor in the conservation of rare alleles
- doesn't have to be worse than pedigree selection at this
- but is much more prone, specifically, to **hitch-hiking** than pedigree selection

Mulder et al., (2019) Genetics

Wientjes et al., (2022) GSE

Wientjes et al., (2023) Genetics

Current work

- Assessment of different genomic selection strategies

Not *if* genomic selection but *how* genomic selection may be implemented

Selection Strategies:

Truncation selection (TS) -----> Maximize average EBVs from selected candidates

Optimal contributions (OCS) -----> with a constraint on the candidates' coancestry

Meuwissen et al., (2020) *Frontiers*

Alleles re-weighting (ARW) -----> with favorable rare alleles up-weighted in EBVs

Liu et al., (2015) *GSE*

(2 versions: *fixed* and *moving* time horizon)

Constrained allele loss (CAL) -----> with a constraint on the reduction in frequency

novel strategy

of rare favourable alleles

* plus Random selection (RS) for reference

The Simulation

The Population:

50 discrete generations

1000 individuals

100 sires + 100 dams selected

- selected without own phenotypes
- using marker effects learnt from the 3 prior generations

Genome:

20k SNP marker panel

- MAFs 0.5 to 0.1
- neutral loci

2k starting causal loci

mutations rate $3.8 \times 10^{-5} (\text{loci.ind})^{-1}$

The Traits

Additive

Normally distributed additive effects, with a common variance.

Dominant

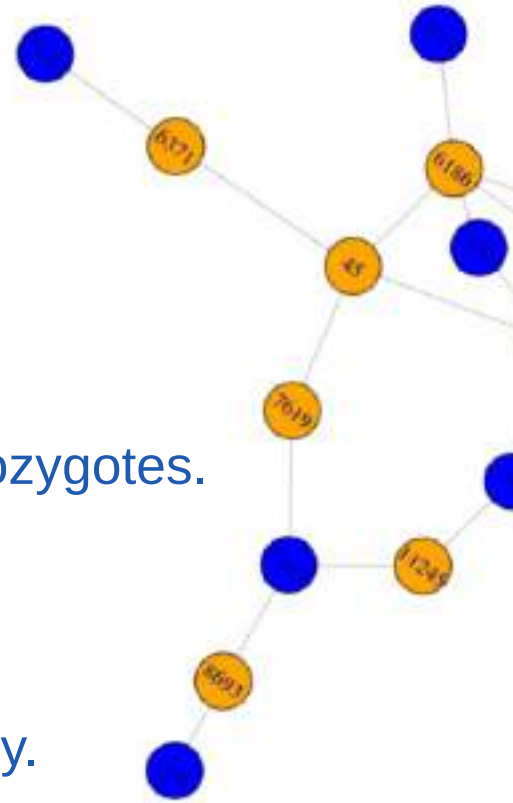
Includes dominance effects, with a small positive bias for heterozygotes.

Epistatic

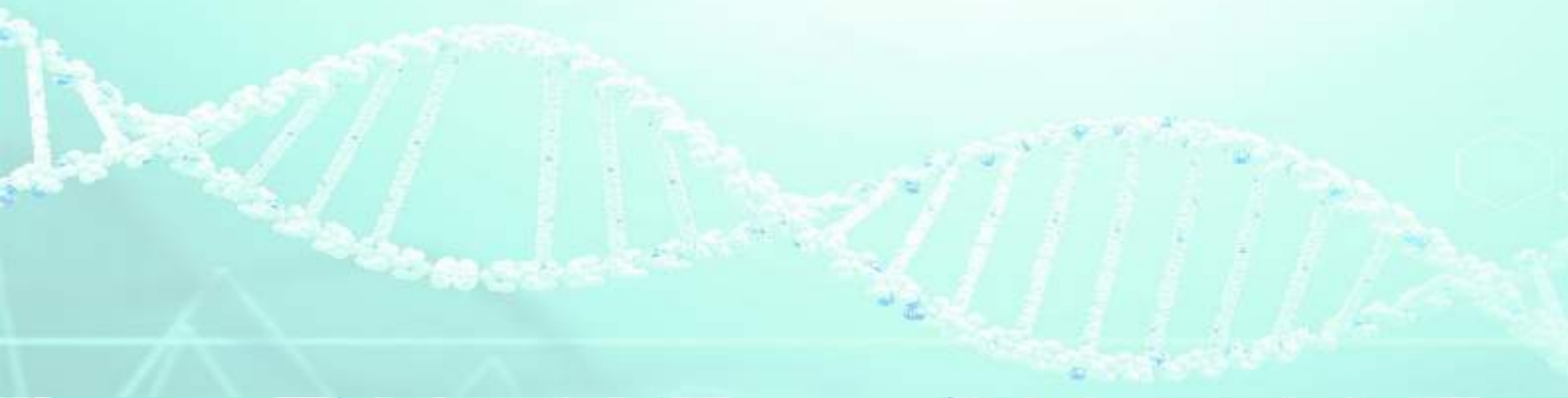
Includes pairwise interactions, with connectivity pattern taken from a yeast study.

Traits specifications taken from Wientjes, et al. 2022

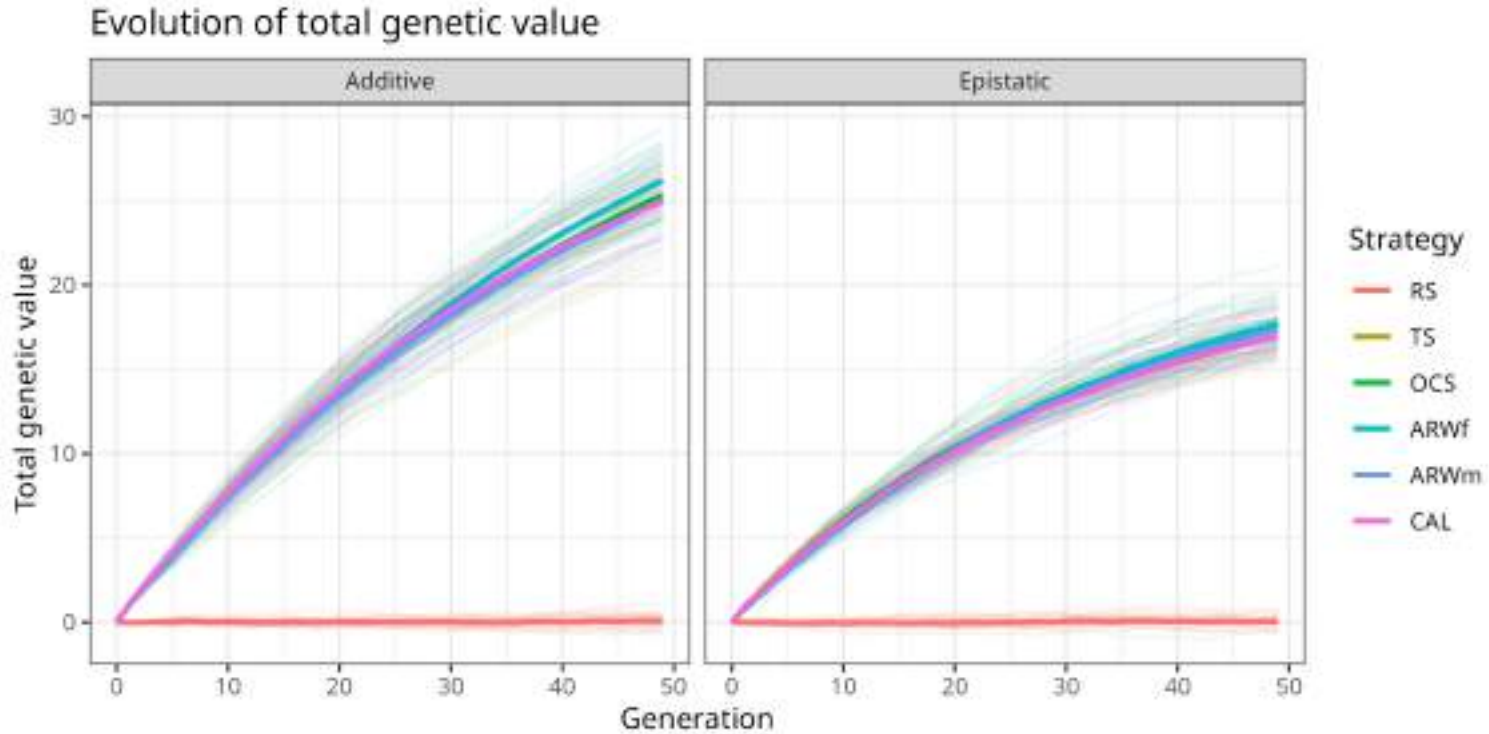
Yeast study in Costanzo et al., 2016



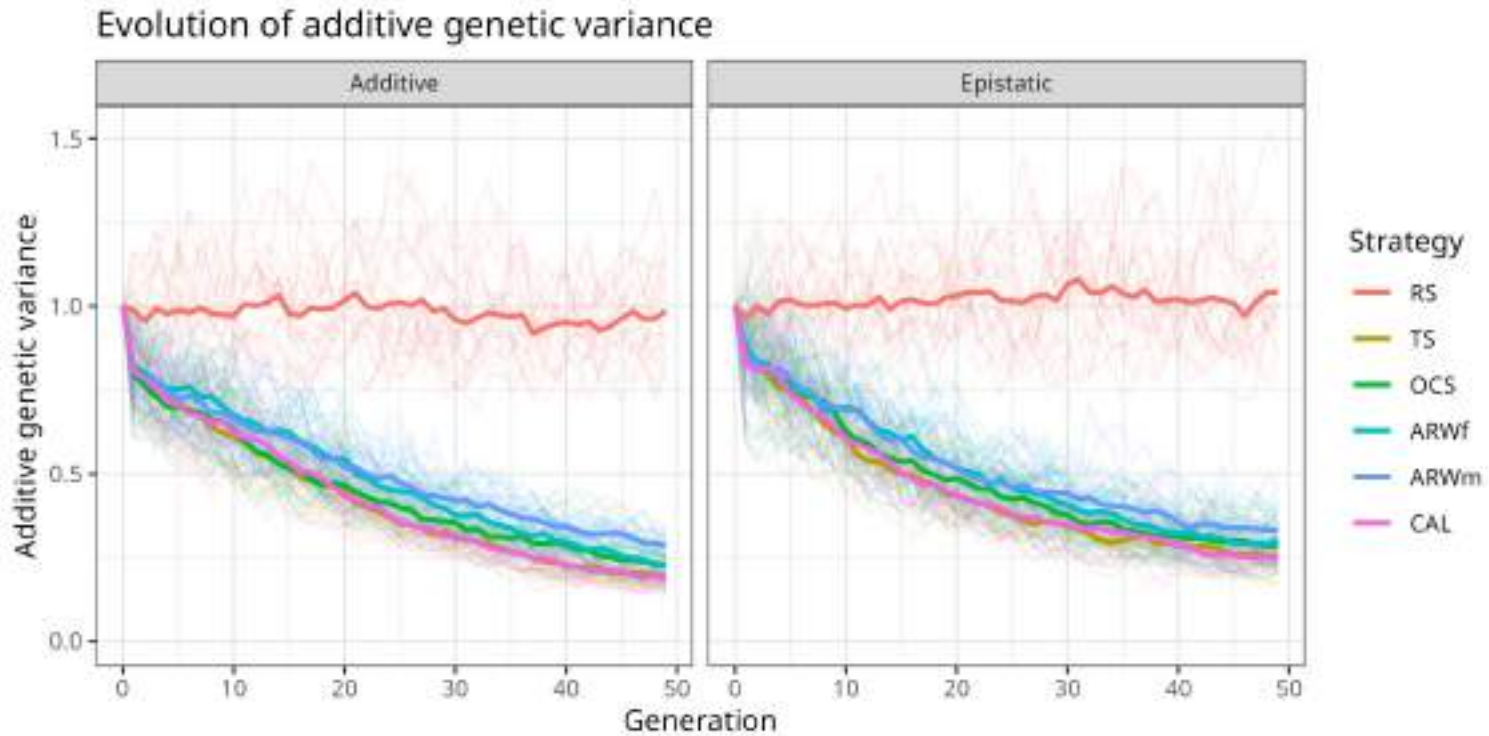
Results & Discussion



Results & Discussion

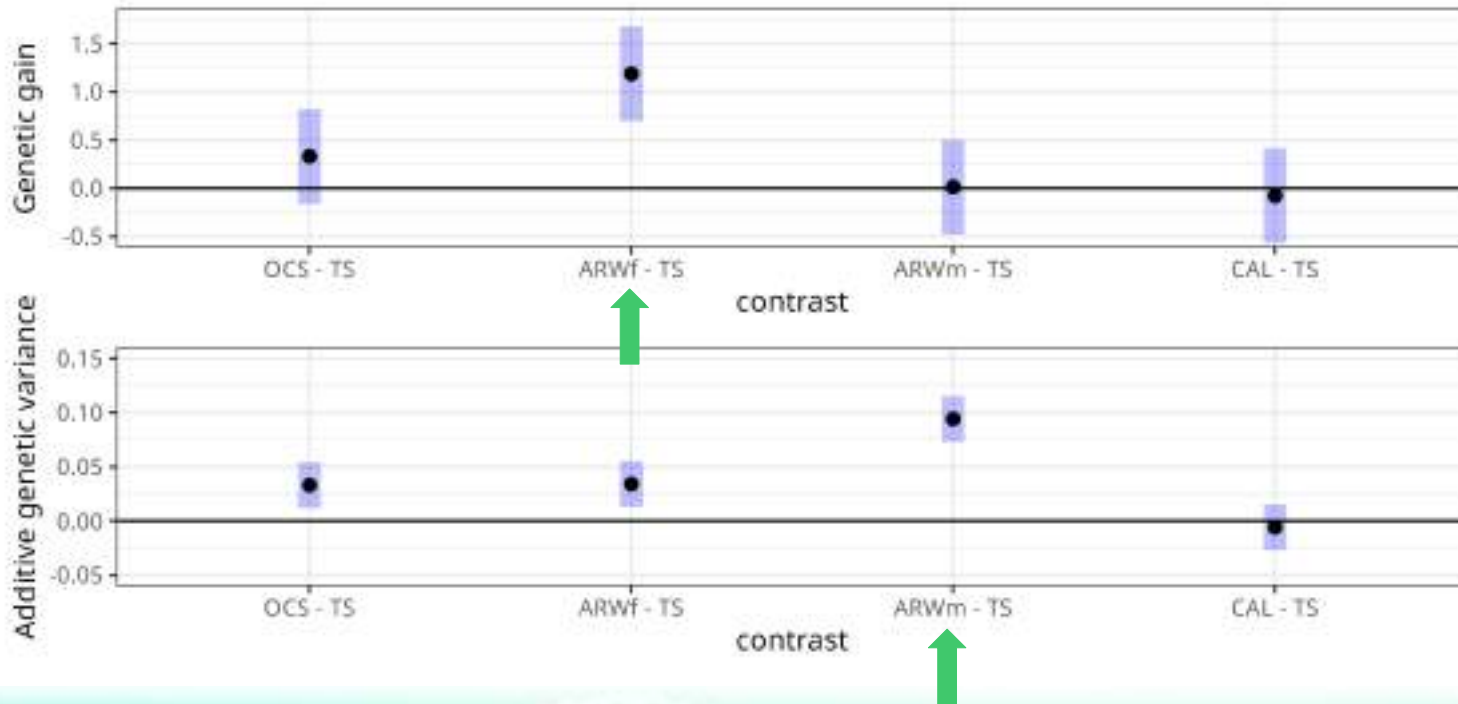


Results & Discussion



Genetic gain vs. genetic variance

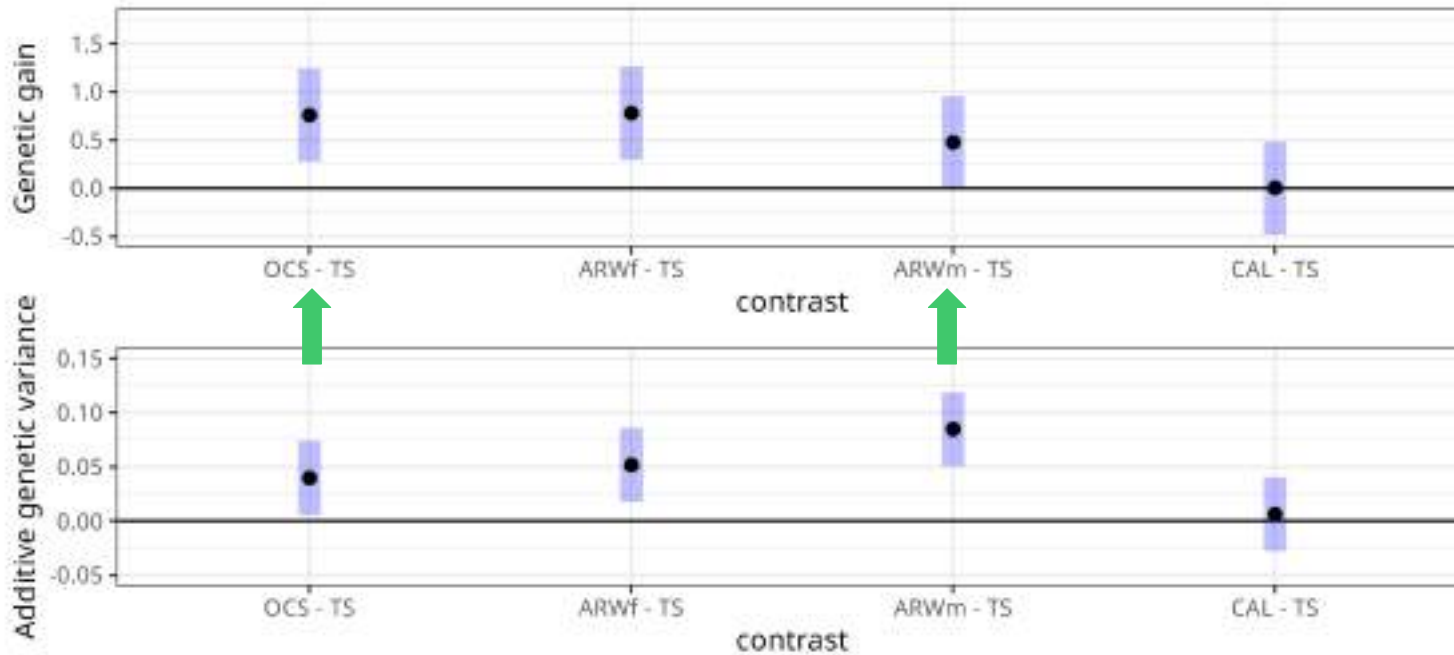
Alternative strategies compared with truncation selection (Additive)



ARW strategies allow effective trade-off between increased genetic gain and conservation of genetic variance

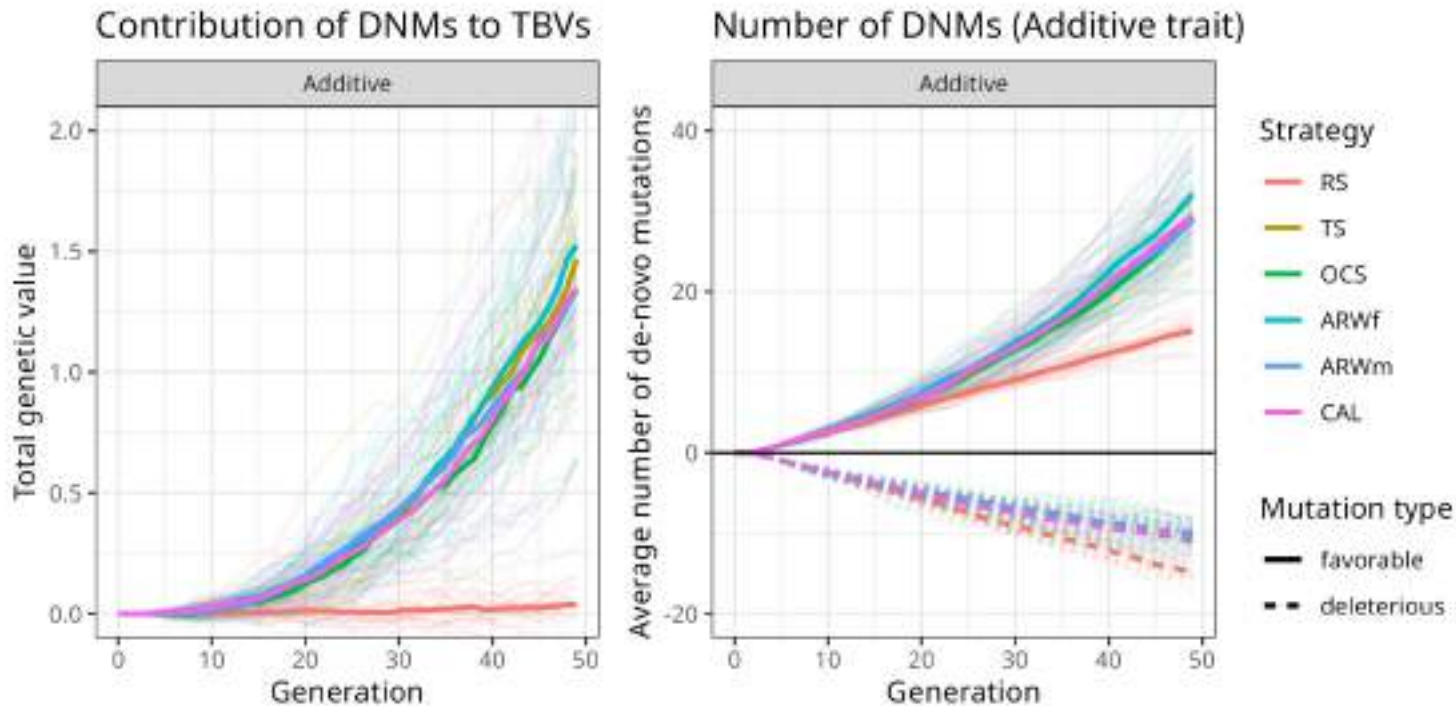
Genetic gain vs. genetic variance

Alternative strategies compared with truncation selection (Epistatic)



Considering traits with non-additive effects improves the assessments of OCS and ARWm for genetic gain

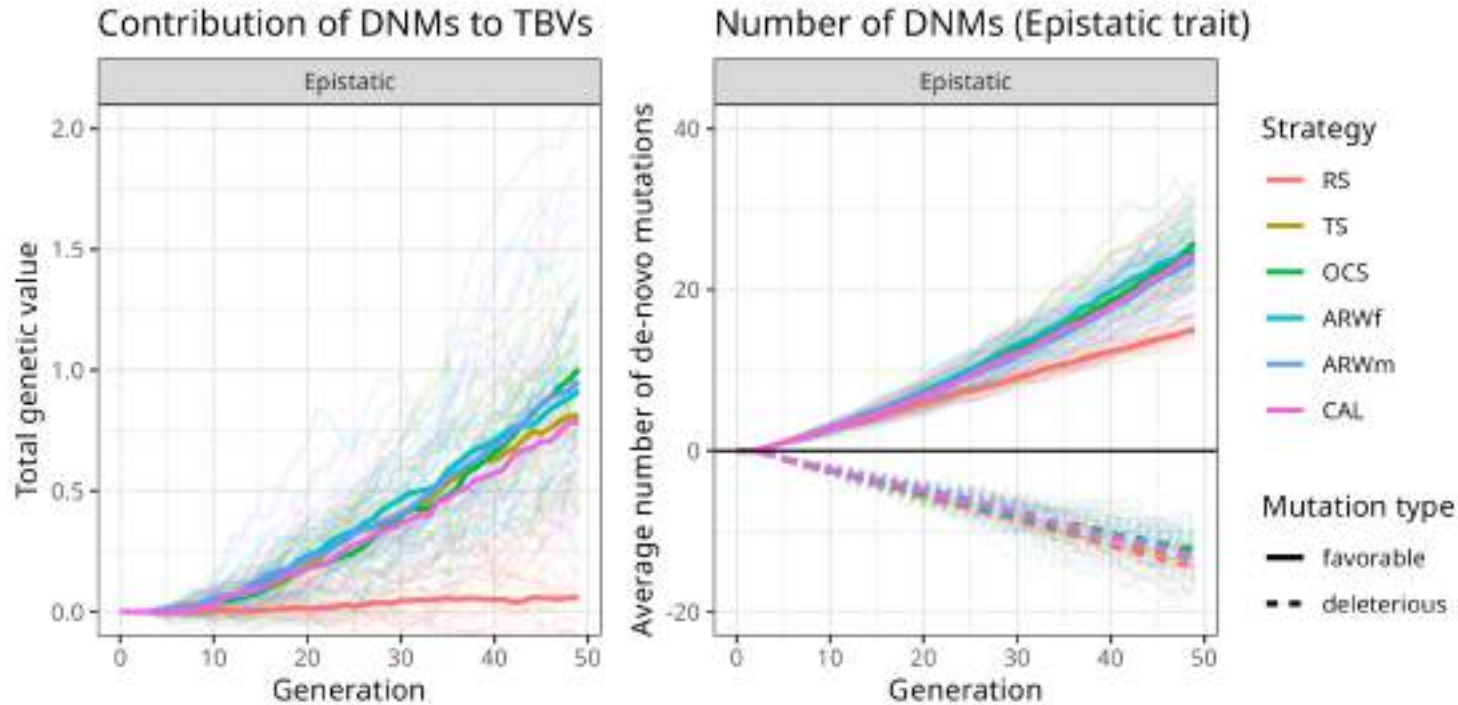
Selection of de-novo mutations



No strategy outperforms truncation selection on these metrics

All selection strategies are applying pressure on the mutations

Selection of de-novo mutations



Considering traits with non-additive effects, selection of DNMs becomes more challenging

CAL selection has the lowest and OCS the highest contribution of DNMs to TBVs

For the fully additive trait

- Truncation selection starts with higher gains,
 - Saturates earlier and gain is surpassed by a reweighting strategy.
- Allelic reweighting is an effective strategy for long term selection,
 - Even if working with markers rather than causal loci.
- No strategy is significantly more effective at keeping favourable de-novo mutations segregating,
 - Although they are all slowly purging the deleterious mutational load.

For the trait with epistasis

- Allelic reweighting remains an effective strategy for long term selection,
 - Even while favorable alleles change through generations.
- Optimal contribution outperforms truncation's long term genetic gain,
 - Which didn't happen for the fully additive trait.
- Purging deleterious mutations becomes more challenging for all the selection criteria explored,
 - Possibly due to a combination of lower narrow-sense heritability and changes in which rare alleles are estimated to be favorable.



Danmarks
Tekniske
Universitet



Matias Schrauf

Thank you for your attention



Jeremie Vandenplas



Yvonne Wientjes

www.rumigen.eu



Han Mulder



This project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No 101000226

Disclaimer: the sole responsibility of this presentation lies with the authors. The Research Executive Agency is not responsible for any use that may be made of the information contained therein.

Extra Slides



Truncation selection (TS)

Maximizes average EBVs from selected candidates without any consideration of diversity management

We estimated SNP effects (β) with the phenotypes of the 3 previous generations (by means of a SNPBLUP model)

And selected the 100 top sires and 100 top dams for:

$$\text{GEBVs} = \mathbf{x}\beta$$

Optimal contribution selection (OCS)

Maximize average EBVs from selected candidates
with a constraint on the candidates' coancestry

Maximize $\mathbf{g} = \mathbf{c}'\mathbf{X}\boldsymbol{\beta}$

$K_t \geq \frac{1}{2} \mathbf{c}'\mathbf{G}\mathbf{c}$

$\mathbf{Qc} = [\frac{1}{2} \ \frac{1}{2}]'$

$\mathbf{c} \geq 0$

where $K_t = K_{t-1} + (1 - K_{t-1}) / (2Ne)$, using $Ne=60$

From Meuwissen, et al. (2020) "Management of genetic diversity in the era of genomics."

Allele re-weighting (ARW)

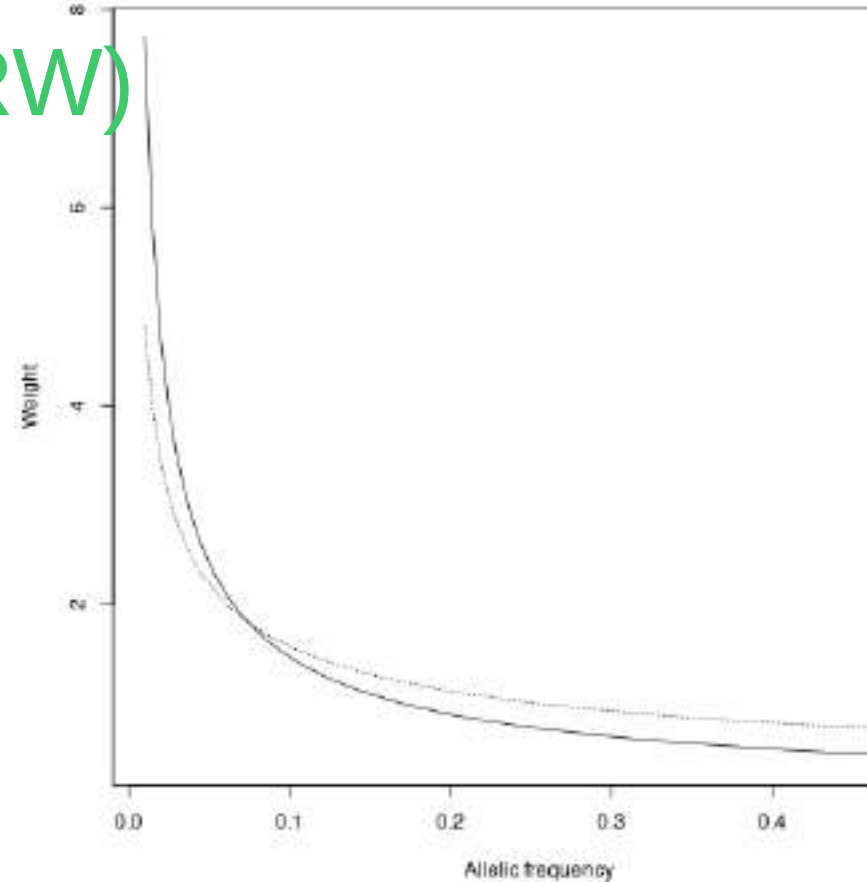
Marker effects of rare alleles re-weighted according to Liu et al., 2015

$$W_{jj} \propto 1/p_j^{c(t)}$$

where $c(0) = 0.5$ and $c(T) = 0.0$
and p_j is the freq of the favourable allele.

$$w\text{GEBVs} = \mathbf{XW}\boldsymbol{\beta}$$

(years to horizon; dotted line: 5 years , solid line: 20 years)



Allele re-weighting (ARW)

Included two variants of this strategy, using different definitions for the time horizons:

- ARWf (fixed): using the full length of the simulation of 50 generations, as the time horizon.
- ARWm (moving): using a moving horizon, always 5 generations ahead.

Constrained allele loss (CAL)

Maximize average EBVs from selected candidates with a constraint on the loss of rare (favourable) alleles.

Maximize $\mathbf{g} = \mathbf{c}'\mathbf{X}\boldsymbol{\beta}$

$\mathbf{L} \geq \mathbf{c}'\mathbf{X}\boldsymbol{\alpha}$

$\mathbf{Q}\mathbf{c} = [\frac{1}{2} \ \frac{1}{2}]'$

$\mathbf{c} \geq 0$

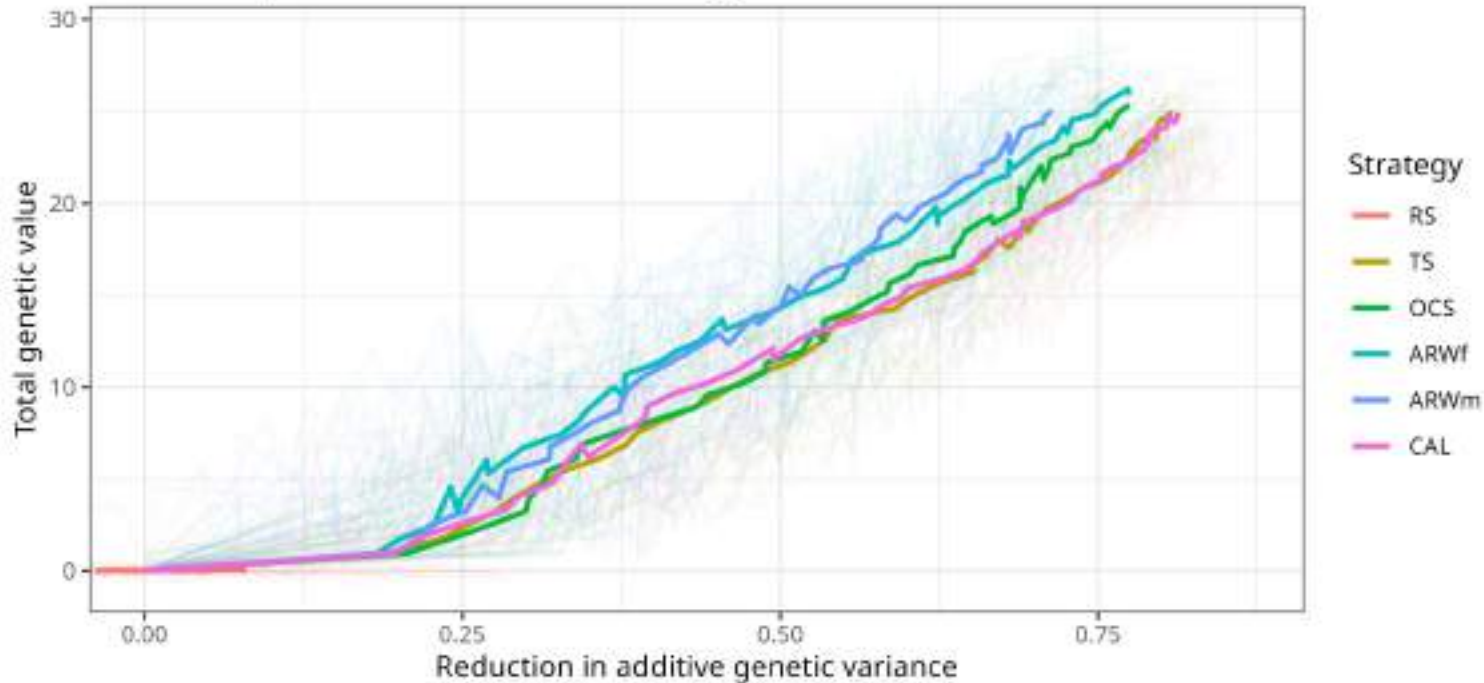
(logarithm with offset $\log(1/n + x)$)

where $\alpha_j = -\log(1/n * (1 + (\mathbf{J}'\mathbf{X})_j))$ [if $\beta_j \geq 0$],

$\mathbf{L} = 1.10 * 1/n * (\mathbf{J}'\mathbf{X}\boldsymbol{\alpha})$, and \mathbf{J} is an n-length vector of ones.

Genetic gain vs. genetic variance

Genetic improvement vs. reduction in genetic variance



ARW strategies allow effective trade-off between increased genetic gain and conservation of genetic variance