

Interpretation of dominant and additive variances from genomic models

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1



Introduction

- Genomic evaluation models typically fit only additive effects.
 - Dominance is of theoretical and practical interest: crosses, mating allocations.

Dominance effects have rarely been included in pedigree-based genetic evaluations. Genomic evaluations have renewed the interest in dominance (e.g., Toro and Varona, 2010; Wellmann and Bennewitz, 2012; Su et al., 2012).



Objectives

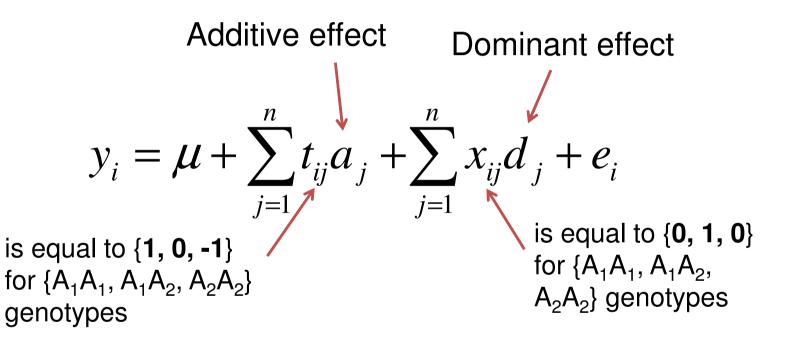
To show the equivalences between additive and dominant effects at the marker and the population levels.

To present how to compute from genotypes the covariances between individuals due to dominant deviations: **D**, Dominant genomic relationship matrix.



Theory

A model with additive and dominant **SNP effects**:



In matrix form for a set of individuals,

 $\mathbf{y} = \mathbf{1}\boldsymbol{\mu} + \mathbf{T}\mathbf{a} + \mathbf{X}\mathbf{d} + \mathbf{e}$



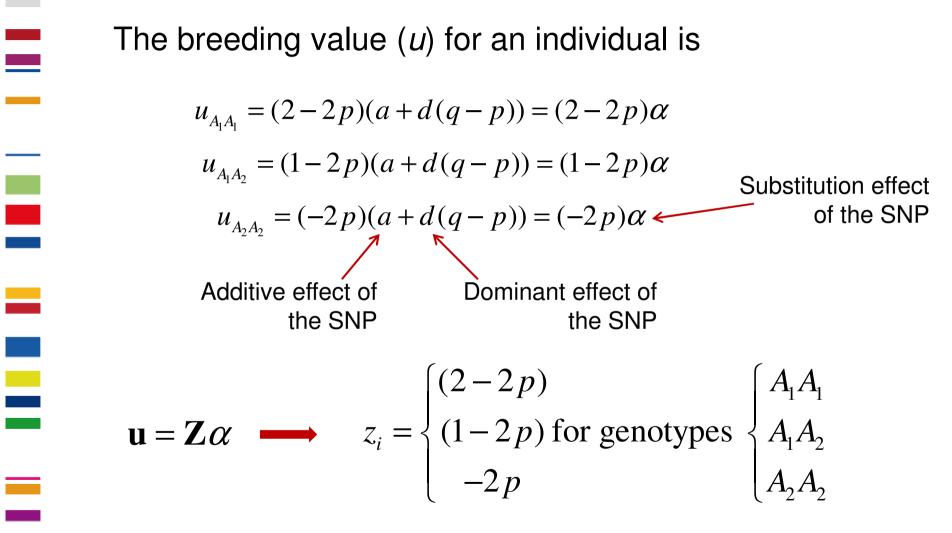
Theory

- Additive or "breeding" values (*u*) of individuals are generated by substitution effects (*α*) (Falconer, 1981)
 The *α* involve both "biological" additive (*a*) and
 - The α involve both "biological" additive (*a*) and dominant (*d*) effects of the markers and the allelic frequency *p*

$$\alpha = a + d(q - p)$$

Dominance deviations (v) only include part of the biological dominant effects of the markers

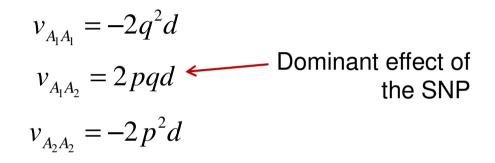
Breeding values



VanRaden, 2008

Dominant deviations

Also, the dominant deviation (\mathbf{v}) of an individual is



So, the dominant deviations of a set of individuals are

$$\mathbf{v} = \mathbf{W}d \qquad \longrightarrow \qquad w_i = \begin{cases} -2q^2 \\ 2pq \text{ for genotypes} \\ -2p^2 \end{cases} \begin{cases} A_1A_1 \\ A_1A_2 \\ A_2A_2 \end{cases}$$

Additive and dominance variances

The partition of the total variance

$$\sigma_G^2 = \sigma_A^2 + \sigma_D^2$$

$$\sigma_A^2 = 2pq\alpha^2 \quad \sigma_D^2 = [2pqd]^2$$

If *a* and *d* are considered random, the covariance of additive individual effects, *u*, is

$$Cov(\mathbf{u}) = \frac{\mathbf{Z}\mathbf{Z}'}{2\sum p_i q_i} \sigma_A^2 = \mathbf{G}\sigma_A^2$$

G is the genomic additive relationship matrix

VanRaden, 2008

with $\sigma_A^2 = 2\sum p_i q_i \sigma_a^2 + 2\sum p_i q_i (q_i - p_i)^2 \sigma_d^2$

Genomic dominant relationship matrix

Also, the covariances across dominant deviations (\boldsymbol{v}) are

$$Cov(\mathbf{v}) = \frac{\mathbf{W}\mathbf{W}'}{\sum (2p_i q_i)^2} \sigma_D^2 = \mathbf{D}\sigma_D^2$$

$$\mathbf{D} \text{ is the dominant genomic} relationship matrix$$
As we have $\sigma_D^2 = \sum (2p_i q_i)^2 \sigma_d^2$

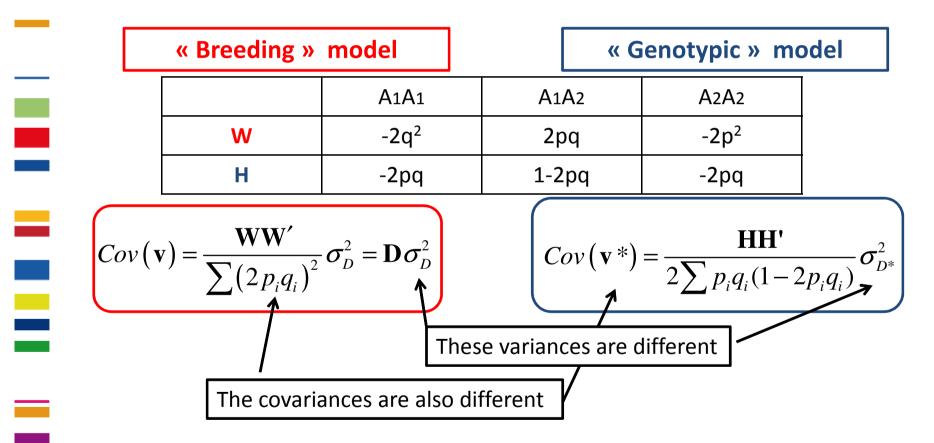
$$\mathbf{SNP variance for dominant component}$$

Use in Mixed Model: DBLUP, GDBLUP



The models

Su et al. (2012) presented an alternative parameterization based on genotypic values of the individuals. We call this "genotypic" model.



It's possible to go from genotypic to breeding model using the total genetic variance



The models

The "**breeding**" (or classical) and the "**genotypic**" models are equivalent models to explain the data (y) but their interpretation is different.

The "breeding" model is expressed in termed of <u>breeding values</u> and <u>dominant deviations</u>; The "genotypic" model, in <u>additive and dominant</u> <u>genotypic values</u>.

 σ_A^2 from the "**breeding**" model is the variance useful in selection and comparable with pedigree-based estimates.



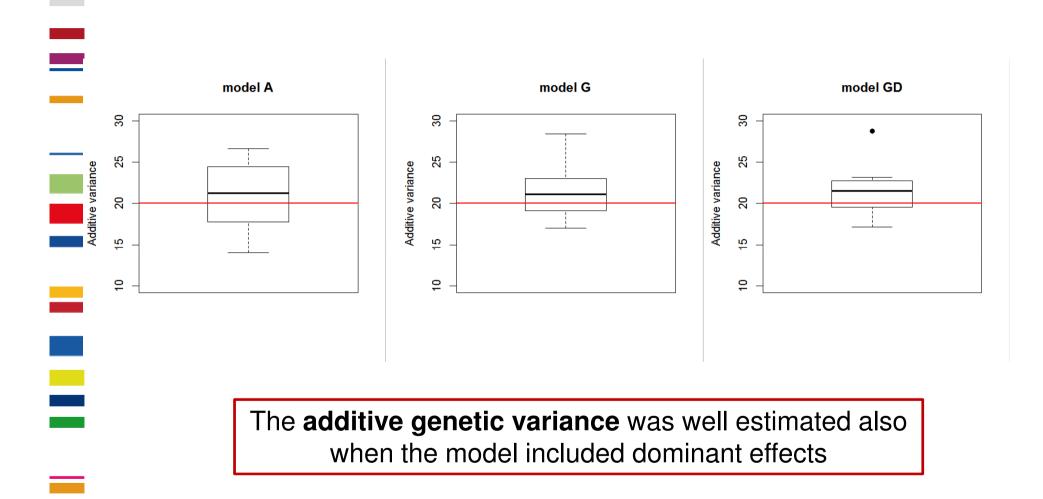
Simulation data

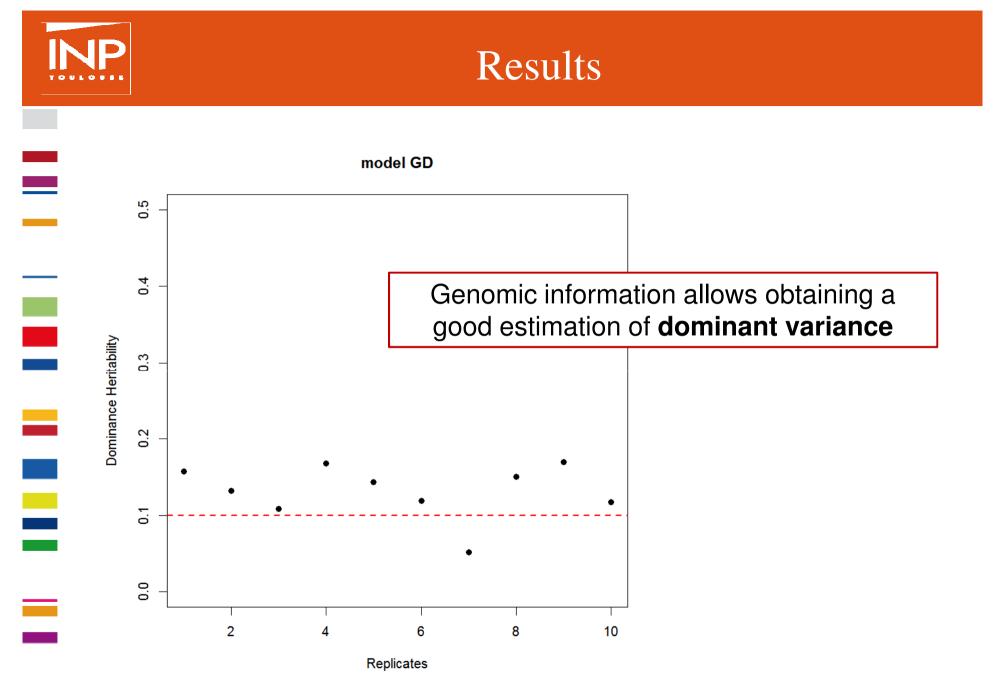
- As in Toro and Varona, 2010 (Ne=100)
- 9,000 markers + 1,000 QTLs
- 2,100 individuals
- Var(A)=20, Var(D)=10, total phenotypic variance=100
- Results were the mean of 10 replicates

Estimation of variance components by REML (remlf90) Model A : pedigree relationships Model G: Genomic additive relationships Model GD: Genomic additive and Dominant relationships









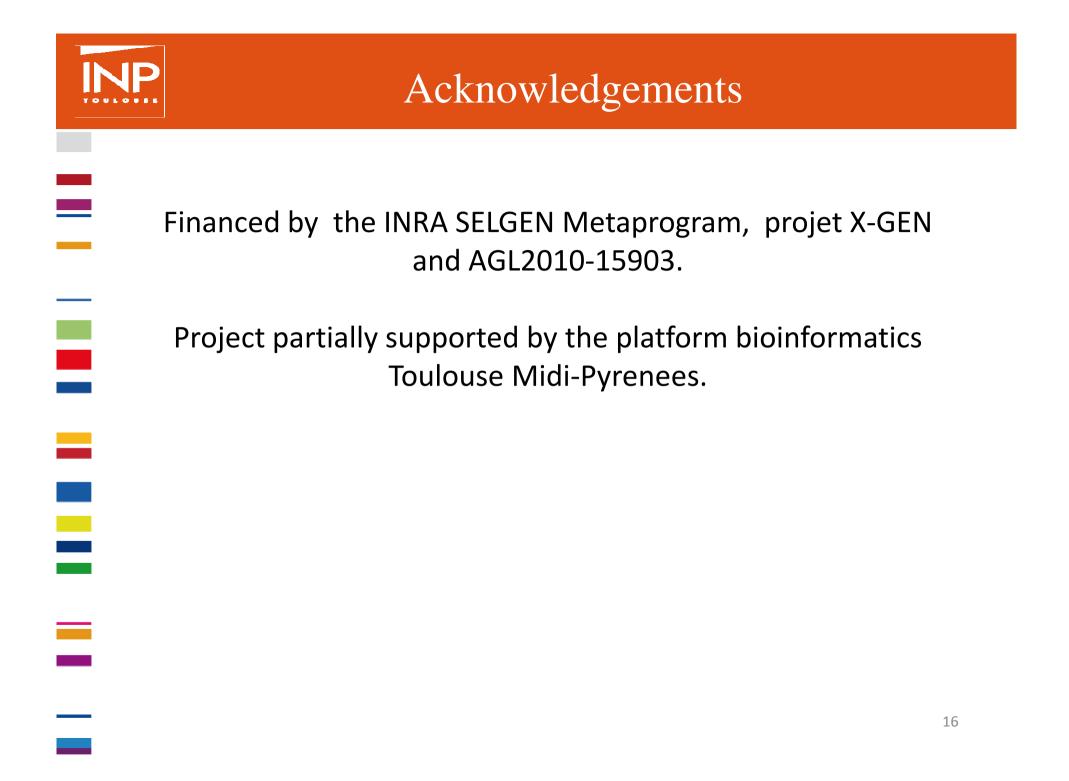


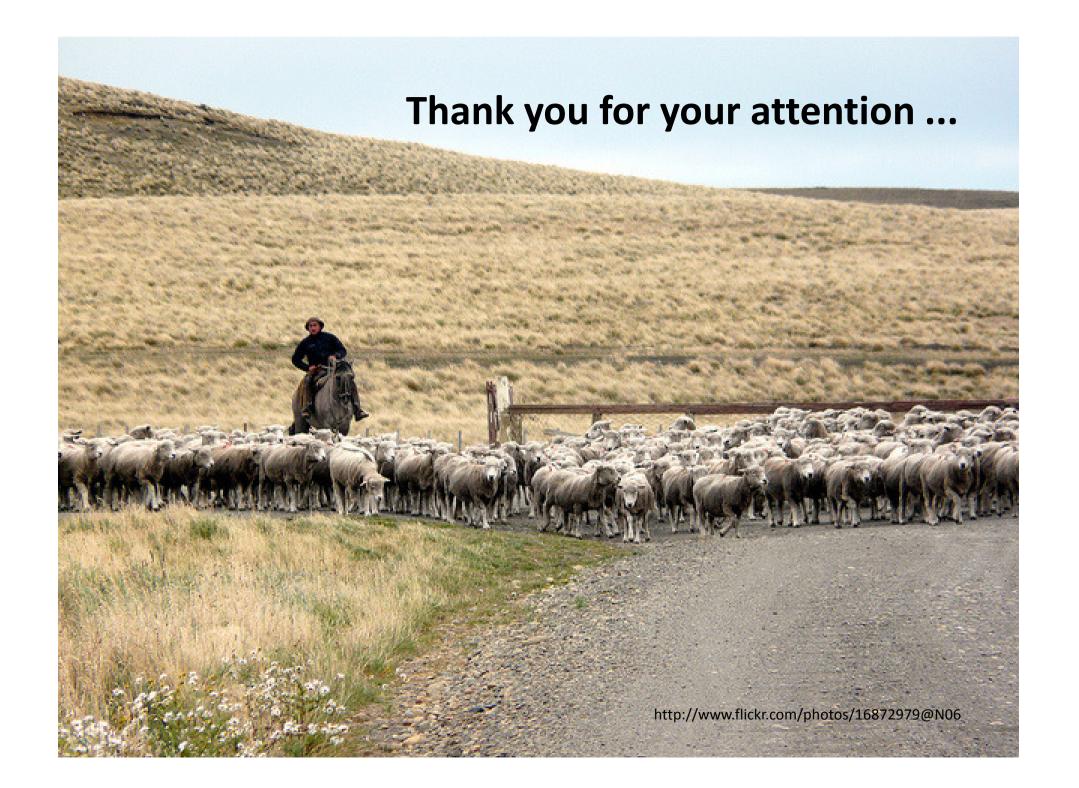
Conclusions

The parameterization in terms of breeding values and substitution effects is more adequate than other parameterization for selection

Models using genomic additive and dominant relationships can estimate variance components correctly

Genomic models including dominance have the advantage that they provide a simple framework, compared with pedigree models







Theory

Remember (as in Falconer 1981)

Consider one locus. Following model (1) the genotypic value G of an individual is as follows:

$$G_{A_1A_1} = a$$
 $G_{A_1A_2} = d$ $G_{A_2A_2} = -a$

where the values *a* and *d* are deviations from the midpoint of the two homozygous.

The genetic mean is therefore

$$E(G) = (p - q)a + 2pqd$$

where *p* is the frequency of A₁ and q = 1-p.

The substitution effect of the gene/marker is

$$\alpha = a + d(q - p)$$

Results

Table 1. Accuracies (SDs) and inflations (SDs) computed from true and estimated breeding values for different effects and prediction models

Effect	Accuracy	Inflation
Additive		
Model A	0.58 (0.04)	0.98 (0.15)
Model G	0.68 (0.02)	0.96 (0.08)
Model GD	0.69 (0.02)	0.96 (0.08)
Model ADped	0.58 (0.03)	0.99 (0.15)
Dominance		
Model GD	0.44 (0.03)	0.91 (0.26)
Model ADped	0.32 (0.03)	1.23 (1.28)

Compared with A and ADped model, all genomic prediction methods (models G and GD) increased **accuracy** by about 18 % and 32% for additive and dominance effects, respectively.