Applied Statistical Methods In Animal Science

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Administration

- ► Course: 2 hours of lecture (2 V)
- ▶ Plan: $2 \text{ V} \rightarrow 1 \text{ U} + 1 \text{ V}$ (i.e., 1 hour of lecture intersperced with time to do exercises)
- Exercises: Work on problems in R
- Material: course notes, slides, solution to exercises
- Exam: written, date: 25.05.2020, 08:15-09:00

Objectives

The students

- are familiar with the properties of fixed linear effects models
- are able to analyse simple data sets
- know why least squares cannot be used for genomic selection.
- know the statistical methods used in genomic selection, such as
 - BLUP-based approaches,
 - Bayesian procedures and
 - LASSO.
- ▶ are able to solve simple exercise problems using the statistical framework R.

Program

Woche	Datum	Thema
1	17.02	Introduction
2	24.02	Linear Fixed Effect Models
3	02.03	GBLUP - Marker-Effects Models
4	09.03	GBLUP - Breeding Value Models
5	16.03	Lasso
6	23.03	Bayesian Approaches

Information

- ▶ Website: https://charlotte-ngs.github.io/GELASMSS2020/
- ► Topics for master thesis
- ► Exam: 25.05.2020 08:15 09:00

This Course

- Use dataset that is used to predict genomic breeding values and introduce four methods
- 1. Fixed Linear Effects Models Least Squares
- 2. GBLUP genomic version of BLUP
- 3. LASSO still fixed linear effects model, but modified parameter estimation
- 4. Bayesian approach to estimate unknown parameter

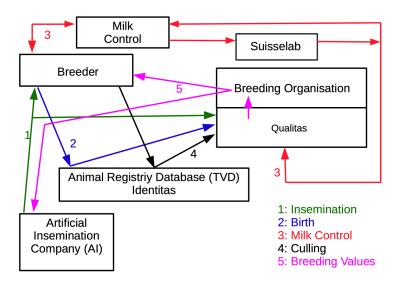
Significance

- Why is this important?
- Is this only relevant for animal breeding?
- ▶ What about the rest of animal science?
- General trend of collecting data has led to development of Big Data
- Examples
 - Presidential campains in the US
 - Health care
 - Face recognition
 - Agriculture: Smart Farming
 - Animal Science

Traditional Animal Breeding

- ▶ Before 2006
- Data collected for other purposes were used to predict breeding values
- Predicted breeding values as side-product

Data Logistics

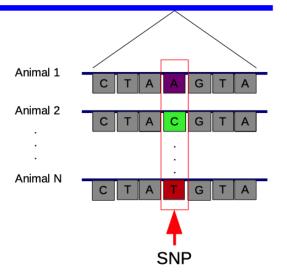


Genomic Selection

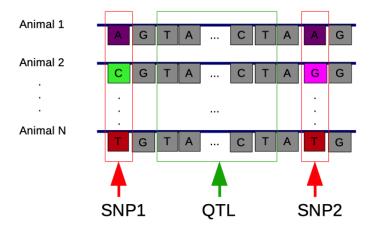
- ► Same goal as in traditional breeding: Find animals with best genetic potential as parents of next generation
- ▶ New: use additional source of information
- **▶ Genomic** information
 - spread accross whole genome
 - single nucleotide polymorphisms (SNP)
- ► Introduction:
- "> Meuwissen THE, Hayes BJ, Goddard ME (2001) Prediction of total genetic value using genome-wide dense marker maps. Genetics 157:1819–1829"
 - ► Popularisation:
- "> L. R. Schaeffer. Strategy for applying genome-wide selection in dairy cat- tle. Journal of Animal Breeding and Genetics, 123(4):218-223, 2006. ISSN 09312668. doi: 10.1111/j.1439-0388.2006.00595.x."

SNP

DNA Molecule



QTL



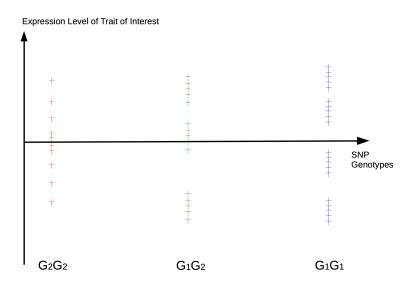
Linkage

- Flanking SNPs and QTL not independent passed on from parents to progeny
- ► Favorable QTL-allele linked with a given SNP-allele
- QTL is unknown, but use SNPs close to QTL as information for selection

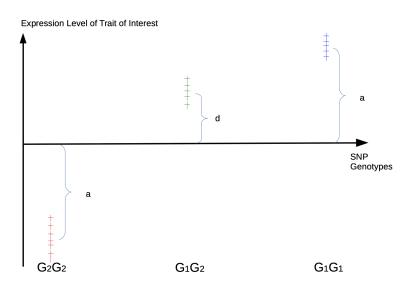
Monogenic Model

- Assume quantitative trait is influenced by one locus only
- ▶ Locus is bi-allelic \rightarrow two alleles (G_1 and G_2) and three genotypes
- Look at Distribution of trait values for three different genotypes

Distribution No Effect



Distribution With Effect



Breeding Value

- Definition: Two times deviation from large number of offspring from population mean
- Assume: Hardy-Weinberg equilibrium
- Compute population mean as expected value of genotypic values
- Compute expected genotypic value of offspring for each of the three parental genotypes
- Assume purely additive loci, hence d = 0

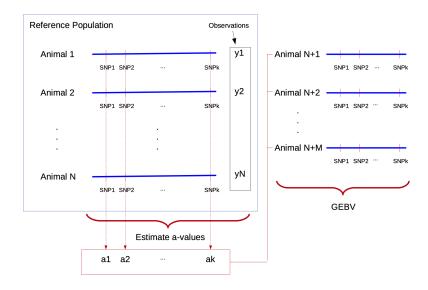
Genomic Breeding Value

- ► Take into account many loci
- Approximate unknown QTL with linked SNP
- Estimate a-effects from monogenic model
- Compute genomic breeding values for all loci based on a effects

Two Approaches

- 1. Two Step Procedure (used currently in Swiss Dairy Cattle)
- 2. Single Step

Two Step



Single Step

- ▶ Combine all information into one single BLUP-based analysis
- ► Problem: Determine covariance between animals with and without genomic information

Summary: Traditional versus Genomic Selection

