Applied Statistical Methods In Animal Science

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22.02.2021

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 Exercise platform, w02
- Material: course notes, slides, solution to exercises
- Exam: written, date: 31.05.2021, 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

Week		Date	Topic
	(1	22.02	Introduction to Applied Statistical Methods
Applied Statistics	2	01.03	Linear Fixed Effect Models
	3	08.03	GBLUP - Marker-Effects Models
	4	15.03	GBLUP - Breeding Value Models
	5	22.03	Lasso
	6	29.03	Bayesian Approaches
7		05.04	Easter Monday
	8	12.04	Introduction to Genetic Evaluation of Livestock
	9	19.04	Model Selection
/	10	26.04	Variance Components
	11	03.05	Genetic Groups and Longitudinal Data
	12	10.05	Genomic Selection
L	13	17.05	Questions, Test Exam
	14	24.05	Pfingstmontag for both courses
	15	31.05	Exams ————————————————————————————————————

Information

- Website: https://charlotte-ngs.github.io/gelasmss2021/
- ► Topics for master thesis: https://charlotte-ngs.github.io/gelasmss2021/misc/MasterThesisTopics_SS2021.html
- ► Exam: 31.05.2021 08:15 09:00

This Course

Bachelor Statistics: Multiple Linear Regression (MLR)
Applied Statistics: Aim: Further develop the concepts started in MLR

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

Methods 2, 3 and 4 are solving problems found with method 1

Assumption: Population of livestock animals. From animals of this population, we have a dataset of observations, and genomic information

Significance

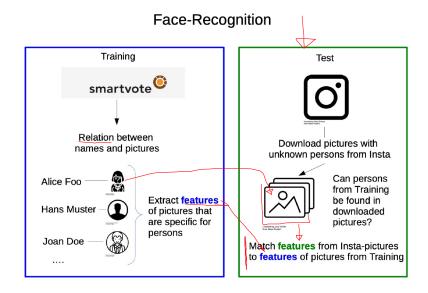
Ex: Corona Pandemic:

Goverments: Develop measures and rules of behavior based on the number of infections, R-value which the reproduction number

- 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
- Examples
 - ▶ Presidential campains in the US
 - ► Health care
 - ► Face recognition
 - Agriculture: Smart Farming
 - Animal Science

Face Recognition

Swiss TV (SRF): 10 vor 10 in 2019

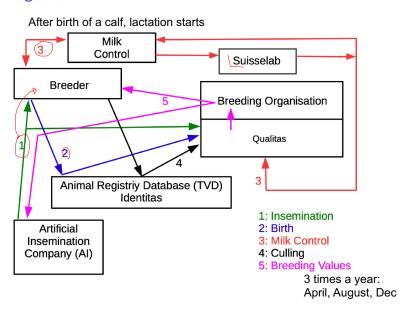


Traditional Animal Breeding

traditional or "pre-genomic" era (before 2006) ==> breeding values are predicted only based on phenotypic information and pedigree data

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

Data Logistics

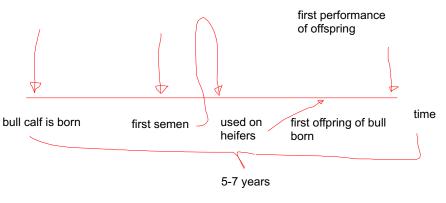


Genomic Selection

- 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: Use genomic selection to save about 90% of the total costs of cattle breeding program
- "> 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."

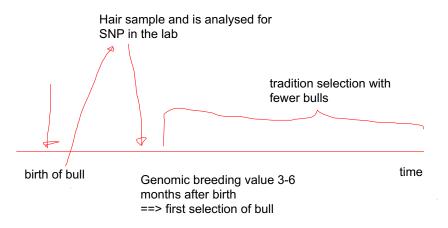
Tradtional breeding programs in dairy cattle:

- selection of bulls is based on evaluation of daughter performance
- most important traits can only be observed in cows

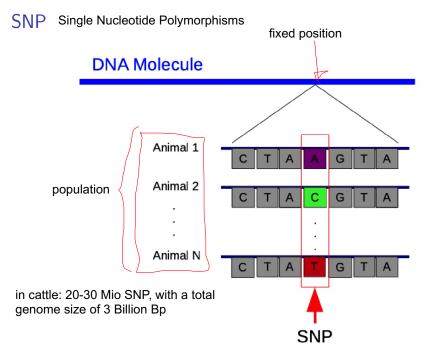


Progeny tests of bulls: start with 300-400 bulls in test, kept 15-20

Breeding program with Genomic Selection

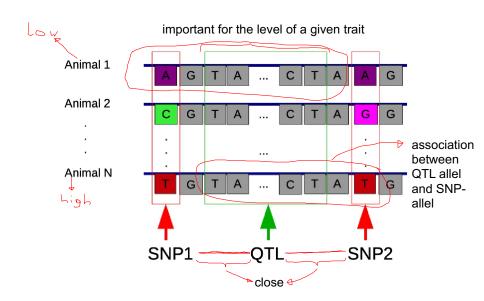


Cost saving: Reduction of time until the first selection decision from 7 years to 6 months.

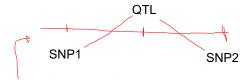




Quantitative Trait Locus, with unknown positions



Linkage

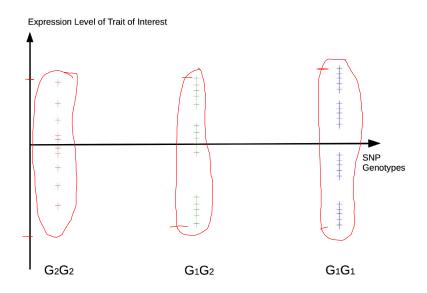


- ► Flanking SNPs and QTL not independent passed on from parents to progeny
- ► Favorable QTL-allele linked with a given SNP-allele
- QTL is unknown, but use SNPs close to QTL as information for selection

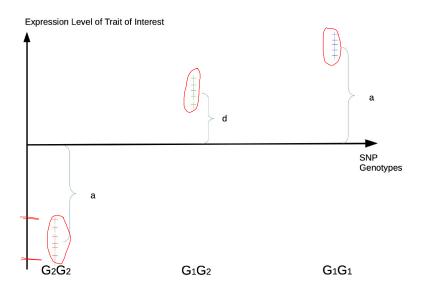
Monogenic Model

- 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





- ▶ 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

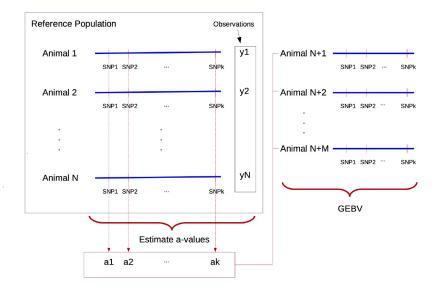
Genomic Breeding Value

- Take into account many loci
- Approximate unknown QTL with linked SNP
- Estimate <u>a-effects</u> 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

