

Smith, E. J., Henshall, J.M. (2009). Variability in the Distributions of Single Nucleotide Polymorphism Effects in Livestock Populations. *Proceedings of the 18th Conference of the Association for the Advancement of Animal Breeding and Genetics*, 18, 64-67.

Background

- ❖ SNP is a DNA sequence variation when a *single nucleotide* (A,T,C,G) in the genome differs between members of a biological species or paired chromosomes in humans.
- ❖ Compare two sequenced DNA fragments from different individuals: AAGCCTA, AAGCTTA
 - Differs by a single nucleotide and is an example of two alleles.
 - *Allele* is an alternative form of the same gene that can result in different observable *phenotypic traits*, such as pigmentation.
 - Almost all common SNPs have only two alleles.
- ❖ SNPs are assigned a *minor allele frequency* (the lesser of the two frequencies) within a population.
 - An SNP allele common in one geographic or ethnic group may be much rarer in another, demonstrating genetic variations between individuals.
 - Such information is most useful in DNA fingerprinting, disease detection and treatment.
- ❖ *SNP density* is affected by the following factors:
 - *Genetic recombination* – new combinations of alleles, encoding a novel set of genetic information
 - *Mutation rate* – measured in units per gamete
- ❖ *Genome size* is total amount of DNA contained within one copy of a single genome.
- ❖ *Quantitative trait loci* (QTL) is a region of DNA associated with a particular phenotypic trait, often found on different chromosomes.
 - *Phenotypes* can be modelled as the sum of genetic and environmental effects.
 - *Heritability* reflects all genetic contributions to a population's phenotypic variance including *additive*, dominant and maternal/paternal effects.
 - *Additive variance* is the variance due to the average (additive) effects of the alleles.
- ❖ Past research on the distribution of QTL effects suggest more rigorous and robust analyses required.
 - Hayes and Goddard (2001) found QTL effects on pig and dairy data displayed a skewed distribution with a few QTL of large effect.
 - Mackay (2004) found that homozygous QTL exhibited an exponential distribution, with most of the variation between parental lines attributable to larger effects.
 - Roff (2007) highlights the need to study the distribution of QTL effects with greater statistical precision.

Outline

- ❖ Aim: To identify factors that influence the distribution of SNP effects.
- ❖ Scope:
 - Bayesian methods used in association studies of dense SNP and phenotype data, rely on assumptions about the distribution of SNP effects.
 - Obtaining reliable estimates for the *true and unknown* distribution of SNP effects is hindered by limited data.
 - Simulation was used to accommodate for this lack of data.
- ❖ Method:
 - Five simulations of livestock populations were performed given the following parameters:
 - #SNPs → number of SNPs
 - SNP / cM → SNP density
 - Dams → number of female parents
 - Sires → number of male parents
 - U → distribution of sampled SNP effects
- ❖ Model:
 - p, q = paired allele frequencies, where $p + q = 1$
 - *mutation rate* = 3.1×10^{-4} per gamete
 - $X_{S1-S4} \sim U(-5, 5)$ $X_{S5} \sim U(-10, 10)$
 - where X is SNP effect size and
 - $U(a, b)$ is the uniform distribution
 - $V(E) = 60.0$ → environmental variance (constant)
 - $V(G) = 20.0$ → genetic variance (target)
 - $\alpha = |X|$ → absolute value of the SNP effect sizes
 - $V(A) = 2pq\alpha^2$ → additive variance
- ❖ Specifications:
 - $E(X)$ was adjusted to account for when SNP was fixed at $p = 1.0$
 - $V(G)$ was not simulated so only the narrow-sense definition of heritability is adopted, meaning only $V(A)$ was modelled.
 - “Assortative” mating system was simulated to account for SNP transmission between animals, mutation and recombination effects.
 - There is algorithm convergence since the results of three repeated runs of $n = 5000$ year periods of simulated data were all similar, indicating stabilised simulations.
- ❖ Analysis:
 - *SNP effects*
 - $E(X) = \frac{1}{2}(a + b) = 0$
 - $V(X_{S1-S4}) = \frac{1}{12}(b - a)^2 = 8.33$ $V(X_{S5}) = \frac{1}{12}(b - a)^2 = 33.33$
 - *QTL effects*
 - $P = G + E$ → phenotype
 - $V(P) = V(G) + V(E) + 2 \text{Cov}(G, E)$
 - $V(P) = 20.0 + 60.0 + 0.0 = 80.0$
 - $\text{Cov}(G, E)$ → controlled and set to zero in a planned experiment.
 - $H^2 = \frac{V(G)}{V(P)}$ → broad definition of heritability
 - $h^2 = \frac{V(A)}{V(P)}$ → narrow definition (additive variance only)
 - $H^2 > h^2 \rightarrow V(A) \in V(G)$

❖ Results:

- Findings do not support the assumption that SNP effect distributions follow an exponential function where $F''(\alpha) > 0$.
- Frequency histograms suggest that $F''(\alpha)$ was not strictly greater than zero, with S_1, S_3, S_4, S_5 containing:
 - inflexion points where $F''(X) = 0$
 - concave down regions where $F''(X) < 0$
- S_2 is clearly not exponential and displays a uniform (rectangular) distribution instead.
- Such results indicate that $F(\alpha)$ may depend on the population parameters used in the simulation, rather than obeying an exponential function.
- $|X| > 6$ or *large effect observations* in S_5 suggest there is an upper limit to the effect size for mutations that can survive in a population.
- Analysis of the *additive variance* ($2pq\alpha^2$) as a function of *allele frequency* (p), suggest that the different (uniform) distributions of *sampled* SNP effects is another factor that influences $F(\alpha)$.
 - S_1 and S_5 have the same simulation parameters except for the width of the sampling interval (a, b) and yet their distributions are significantly different.

❖ Conclusions:

- Use of particular distributions (like the exponential) as priors for Bayesian analyses of SNP effects is invalidated.
- SNP effect distribution was found to be influenced by genome size, SNP density, population size and the distribution of sampled SNP effects.

❖ Comments:

- Are Bayesian prior assumptions of exponentiality based on QTL effects rather than on SNP effects (due to lack of data)? In other words, is QTL a proxy variable for SNP?
- How were the interval values (a, b) for uniform distribution chosen?
- Use of MCMC methods to approximate distribution of realised simulations for more accurate inferences, in addition to histograms.
- Simulate genetic variance in order to model broad-sense definition of heritability, rather than “tuning” the simulation to keep $V(G)$ consistent with observed heritabilities. This is to reduce underestimation bias since $h^2 < H^2$.
- Alternative approaches to genomic simulation explored by more recent studies below.

❖ Recent studies:

- The following two papers are co-authored by UNE Postdoctoral Fellow at the School of Environmental and Rural Sciences, Dr John Hickey.
 - Daetwyler et al. (2013) “Genomic Prediction in Animals and Plants: Simulation of Data, Validation, Reporting, and Benchmarking”
 - Hickey & Gorjanc (2012) “Simulated Data for Genomic Selection and Genome-Wide Association Using a Combination of Coalescent and Gene Drop Methods”