Uncertainty and Imputation of Plant Genotypes

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Primary questions

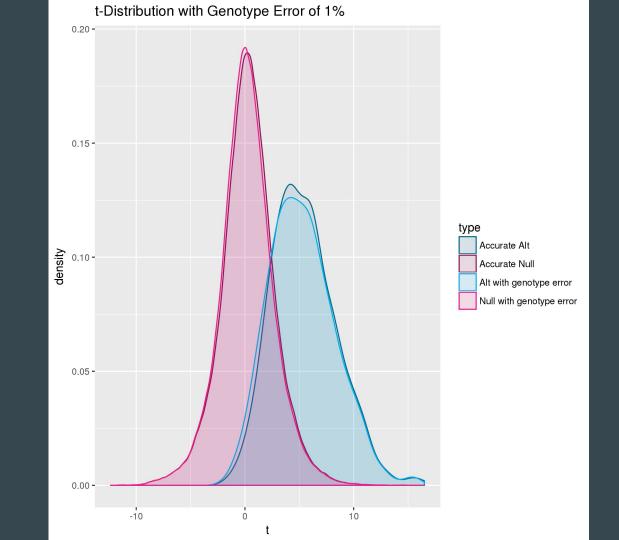
- Can we quantify the uncertainty within a plant database due to genotyping errors?
- Can we adjust summary statistics to account for changes in Type I and Type II error?

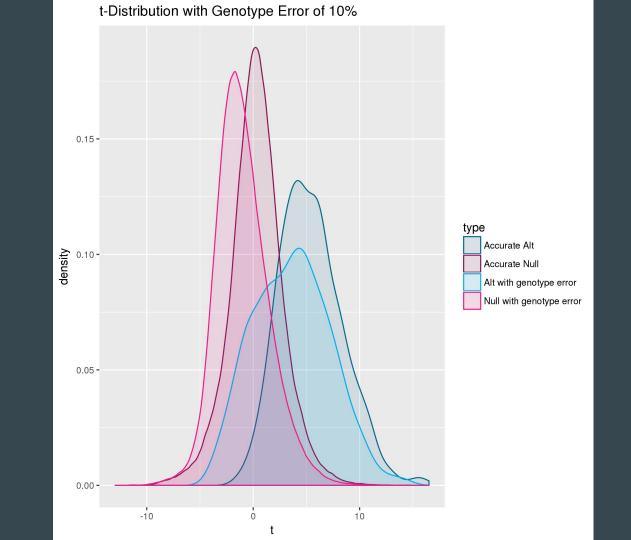
Background

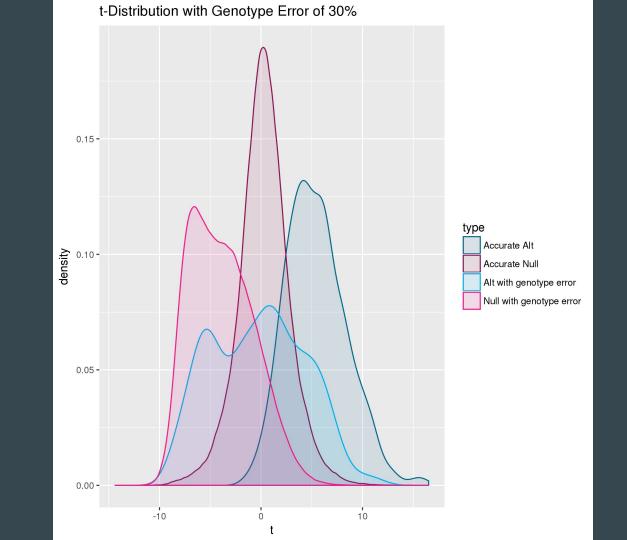
- Plant genetic data has characteristics distinct from human genetic data
 - Lower sequencing coverage
 - Allelic frequencies are dissimilar to humans
 - e.g. with breeding/selfing

Main methods

- Real wheat genetics data simulated phenotypes
 - Minimal correlation between SNPs
- Error distribution
 - Haplotype level
 - \circ Mapped geno \rightarrow pheno \rightarrow error rates \rightarrow erroneous geno
- Parameters
 - Direction of the effect of causal SNPs (+/-)
 - Direction of correlation between phenotype and error (+/-)
 - \circ Maximum error rates (1%, 10%, 30%)
- Test of association
 - Linear models







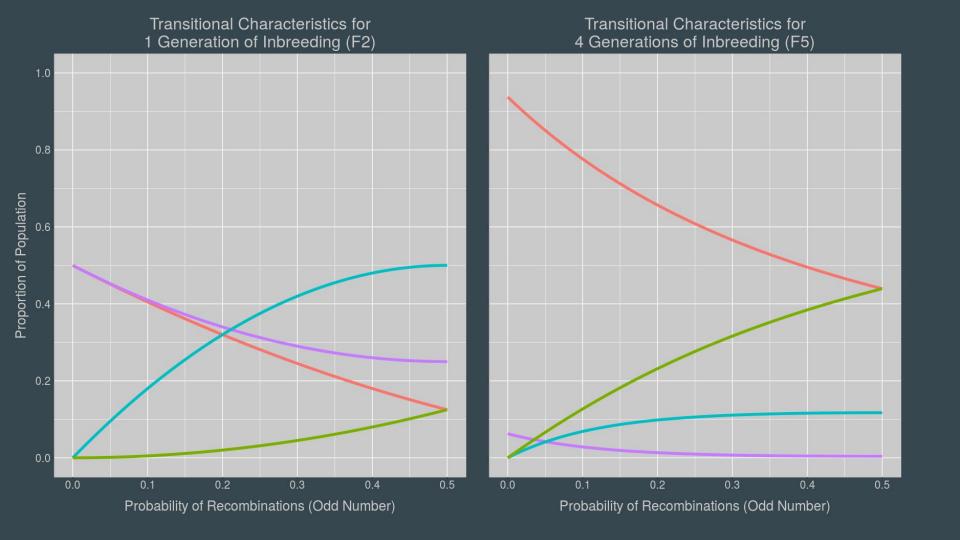
LaByRInth (Low-coverage Biallelic R Imputation)

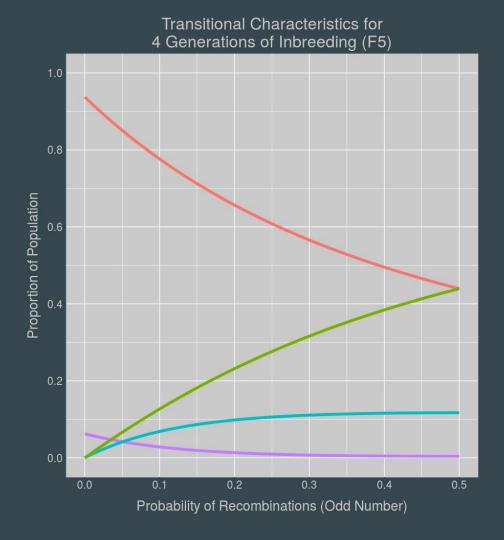
- Accurate imputation can alleviate problems with differential errors
- LB-Impute is one software program to do this
 - Uses a hidden Markov model
- Last summer
 - Port LB-Impute from Java to R for accessibility
- This summer
 - Explore recombinant inbred line specific structure
 - Symbolic equations generated for recombinant lines
 - Transition frequencies
 - Estimate recombination probabilities instead of using physical distance mapping
 - Informed transition probabilities in hidden Markov model

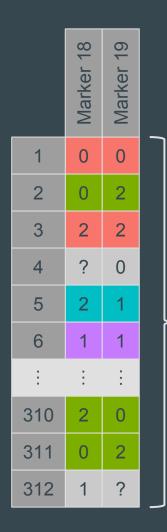
Estimating probability of recombination

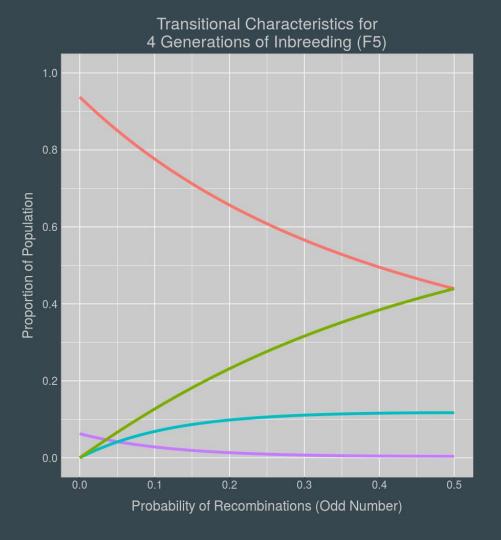
- For a pair of loci, look at frequencies of "transition types" across all members
- Dependent on generation of inbreeding and probability of an odd number of recombinations between the loci

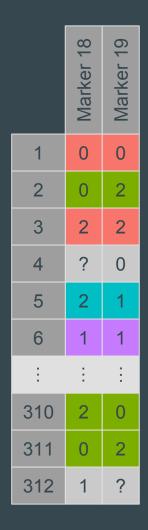
Marker <i>i</i>	Marker <i>i+1</i>
Homozygous	Homozygous for same allele
Homozygous	Homozygous for other allele
Homozygous	Heterozygous
Heterozygous	Homozygous
Heterozygous	Heterozygous

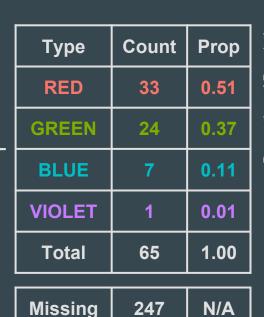


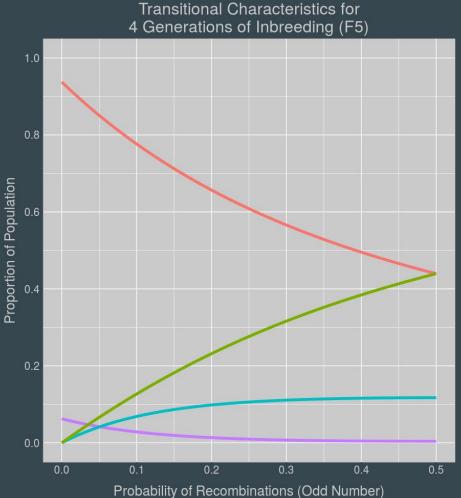


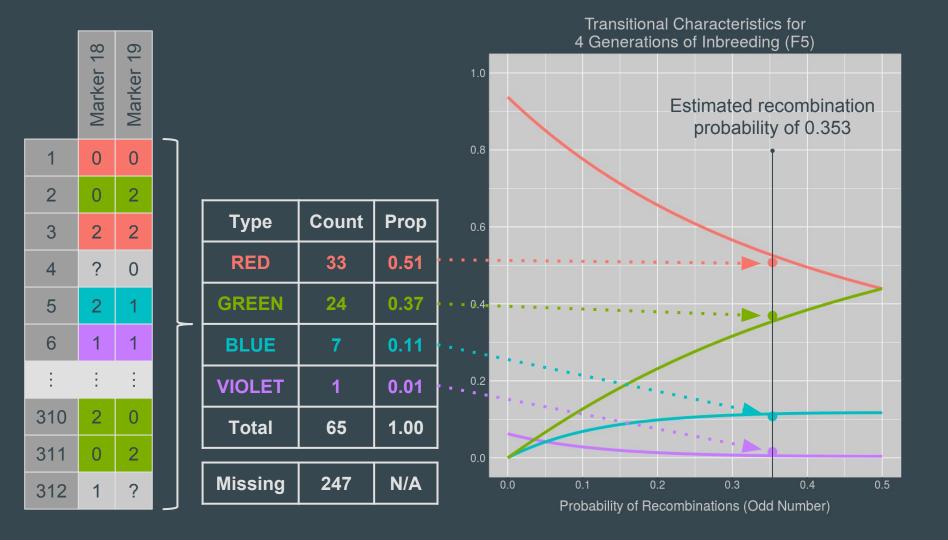


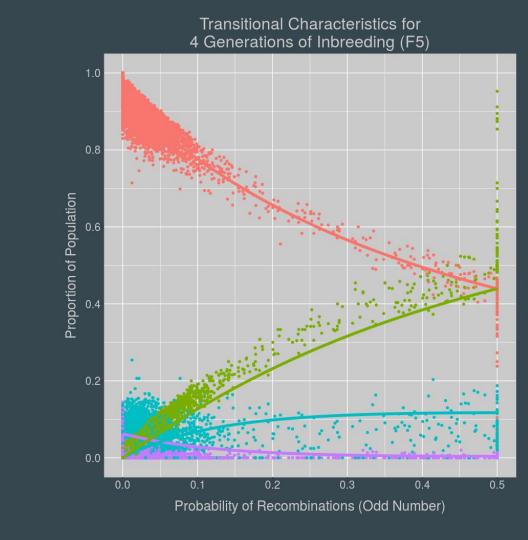


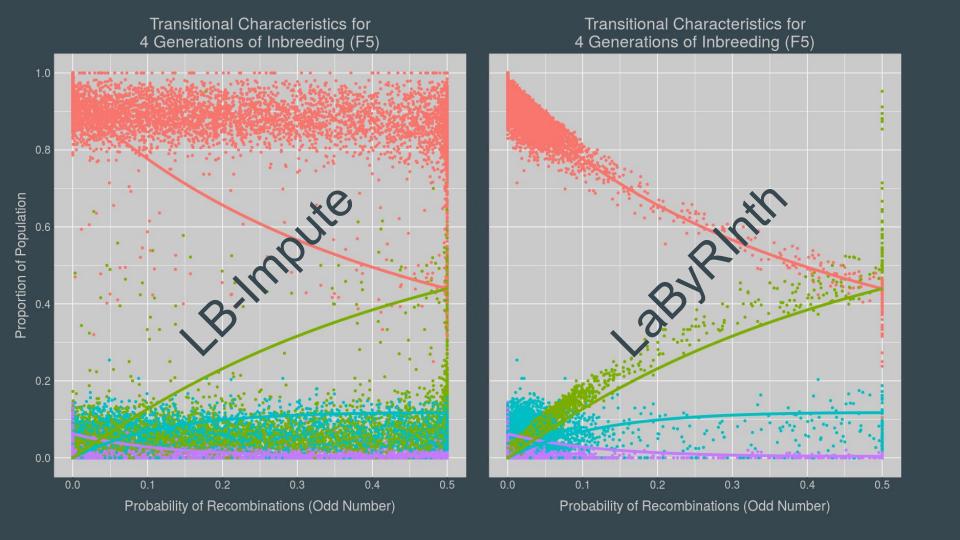












Future Work

- Uncertainty analysis
 - Develop methods
 - Estimate error rates
 - Adjust test statistics to compensate for inflated type I error
 - Extend results to phenotypic error
- Imputation
 - Update Markov transition probabilities using recombination probabilities
 - Release LaByRInth as an R package
 - Publish full LaByRInth paper (recently submitted software note)

References

- Edwards, B., Haynes, C., Levenstien, M., Finch, S., Gordon, D.: Power and sample size calculations in the presence of phenotype errors for case/control genetic association studies. BMC Gen. 6, 18. (2005). https://doi.org/10.1186/1471-2156-6-18
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- Powers, S., Gopalakarishnan, S., Tintle, N.: Assessing the Impact of Non-Differential Genotyping Errors on Rare Variant Tests of Association. Hum. Hered. 152--159. (2011).

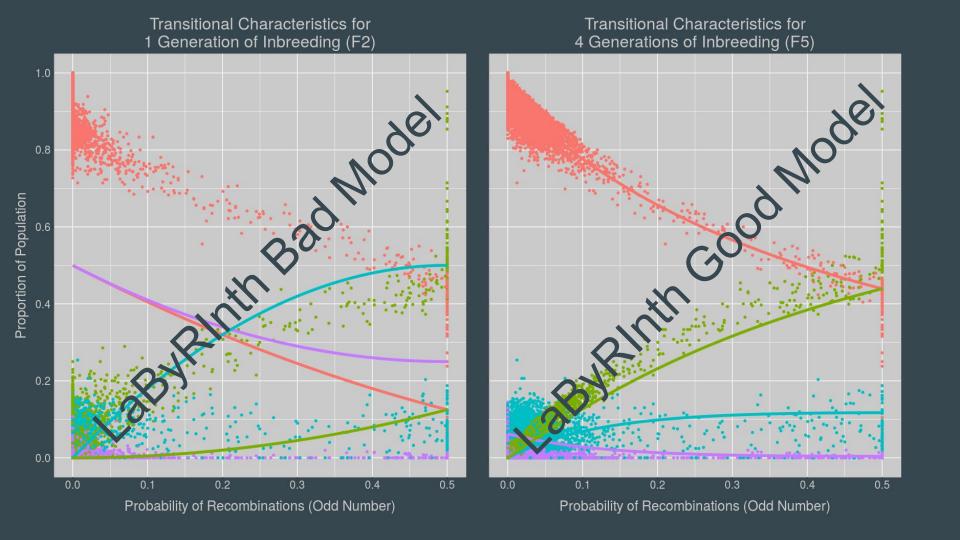
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Does the data fit too well in comparison? Are we fitting data well, but to the wrong model?



Recombinant Inbred Line

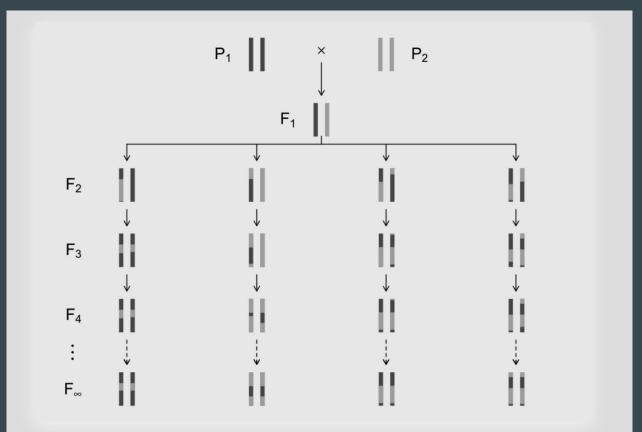


Image from: https://www.biostat.wisc.edu/~kbroman/talks/FunQTL/