

Evaluating the Effects of Trembling Aspen Genetics on Insect Communities

Stat 877 project presentation

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Background

Plant-insect interactions:

- Plants are known to influence insect community structure
- Plant Diversity is positively correlated with insect diversity.
 - effects of interspecific variation are well studied
 - effects of **intraspecific variation** are less well studied

Populus tremuloides (Trembling Aspen):

- the most wide-spread tree species in North America
- one of the fastest-growing tree species
- incredible amounts of intraspecific variation
- food source for an incredible diversity of herbivores

Genome-wide association analysis (GWA)

Question: How do aspen genetics shape insect communities?

- Herbivorous insects are known to differ by aspen clone (genet) and by aspen traits (size, defense chemistry, nutrition, etc.).
- Is there evidence for genetic drivers other than these physical traits?
- **Are there genomic regions that affect multiple insects?**

GWA of insect incidence on aspen common garden:

- 18 common insect species; present/absent (1/0)
- 1,414 trees from 437 genets
- 8 tree trait covariates
- 4 time periods (longitudinal)
- 114,420 SNP regions with 3 possible genotypes {0,1,2}

Statistical Model

$$\text{logit}(p_{ijk_g}) = \beta_0 + \alpha G_g + x_{jk_g}^T \beta + \varepsilon_{g(j)}$$

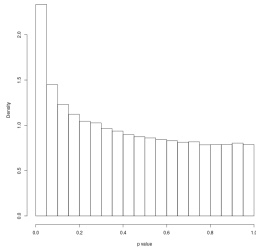
Where:

- $\text{logit}(y) = \log\left(\frac{y}{1-y}\right)$
- p_{ijk_g} : probability of observing > 0 individual insects of species i during survey event j on tree k of genet g
- G_g : SNP-specific genotype of genet g
- x_{jk_g} : vector of observed tree trait covariates for tree k during survey event j
- $\varepsilon_{g(j)}$: nested random effect of genet within survey event
- Storey's q values of coefficient α used to select significant SNPs (.05 cutoff)

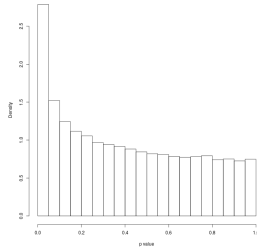
A total of 2,059,560 (18 insects \times 114,420 SNPs) GLMMs were fit.

Diagnostics: p value distributions

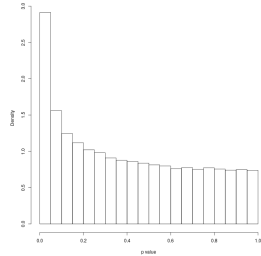
Histogram of Harmandia p values



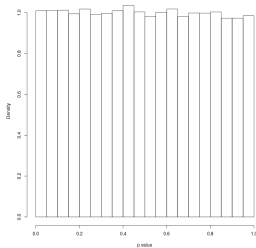
Histogram of Petiole Gall p values



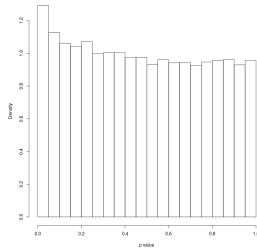
Histogram of Phyllocolpa p values



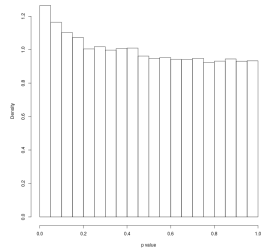
Histogram of Blotch Mine p values



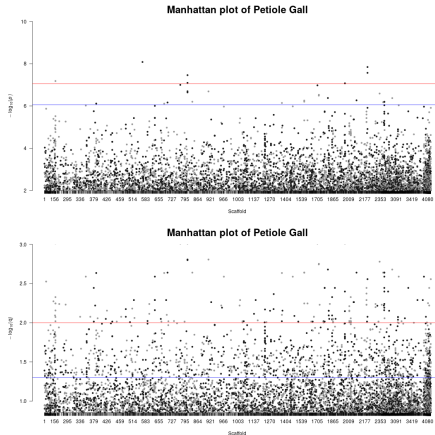
Histogram of Ants p values



Histogram of Green Sawfly p values



Diagnostics: Bonferroni vs. Storey's q



- Using q values (bottom) allows for discovery of many more potential associations than bonferroni p values (top).
- significant q values contain all significant bonferroni p values

Results: Significant SNPs per Insect

	insect description	SNPs	scaffolds
Green Aphids	free-feeding, specialist (salicaceae)	1788	1020
Petiole Gall	leaf-galling, specialist (populus)	1432	788
Phyllocolpa	leaf-rolling, specialist (salicaceae)	1363	814
Harmandia	leaf-galling, specialist (populus)	174	130
Smokey Aphids	free-feeding, specialist (populus)	90	64
Casebearer Moth	case-bearing, generalist	7	3
Lombardy Mine	leaf-mining, specialist (populus)	1	1
Cottonwood Leaf Mine	leaf-mining, specialist (salicaceae)	1	1
Leaf Edge Mine	leaf-mining, specialist (salicaceae)	0	0
Blotch Mine	leaf-mining, specialist (populus)	0	0
Weevil Mine	leaf-mining, specialist (salicaceae)	0	0
Blackmine Beetle	leaf-mining, specialist (populus)	0	0
Leafhoppers	free-feeding, generalist	0	0
Ants	aphid-tending, non-herbivore	0	0
Pale Green Notodontid	free-feeding, specialist (populus)	0	0
Aspen Leaf Beetle	free-feeding, specialist (populus)	0	0
Green Sawfly	free-feeding, specialist (populus)	0	0
Cotton Scale	scale insect, generalist	0	0

There are 4,768 unique significant SNPs.

Results: Common Associations

- Of these 4,768 unique SNPs, There are **84** SNPs with multiple insect associations. These will be our SNPs of interest.
- 4 of these had associations with 3 insects and were contained in 1 gene:

SNP.name	qval.Harmandia	qval.Phyllocolpa	qval.Petiole.Gall
Potra002191:26587	0.0193784	0.0123563	0.0286529
Potra002191:26642	0.0226734	0.0091460	0.0258001
Potra002191:26644	0.0226734	0.0091460	0.0258001
Potra002191:26650	0.0236877	0.0116280	0.0234547

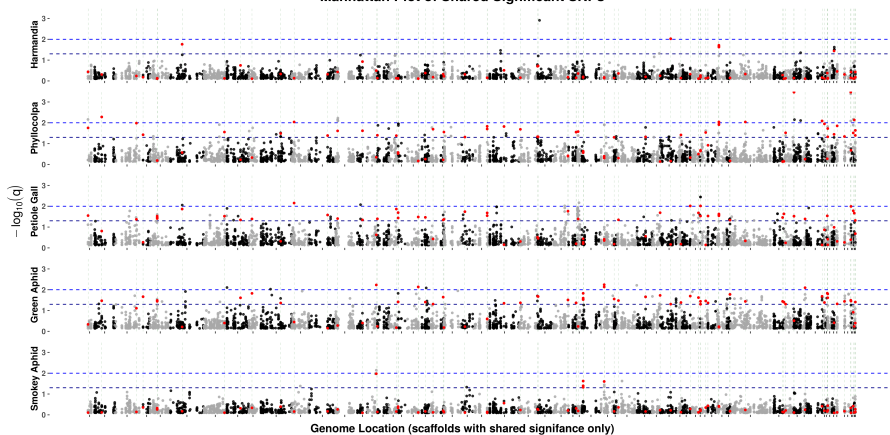
- They are within 63 base pairs (LD/same gene)

Note that these are the 2 **leaf-galling** and 1 **leaf-rolling** species. They are tightly associated with leaves. However, the gene annotation is not very insightful:

Potri Gene	%match	Descr.	avg allele. freq
Potri 011G157300.1	98.97	Uncharacterized oxidoreductase C663	0.136

Results: Multiple-Manhattan Plot

Manhattan Plot of Shared Significant SNPs



Results: Imputed Gene Function

The following is a few of our top (by % gene match) genes with annotations, imputed from congeneric *Populus trichocarpa*.

Potri Gene	%match	Descr.	avg allele. freq
Potri 006G267600.1	99.66	protein phosphatase 2C	0.110
Potri 013G036000.1	99.44	tRNA-specific adenosine deaminase	0.204
Potri 007G058500.1	99.02	1-deoxy-D-xylulose-5-phosphate synthase	0.226
Potri 014G079900.1	99.01	dependent malic enzyme	0.112
Potri 003G045300.1	99.01	multivesicular body protein	0.143
Potri 011G157300.1	98.97	Uncharacterized oxidoreductase C663	0.136
Potri 006G249800.1	98.63	Probable boron transporter	0.139
Potri 016G142100.1	98.60	Unknown protein 1	0.150
Potri 001G369000.1	98.59	Transmembrane emp24 domain-containing protein	0.205
Potri 016G037900.1	98.40	STRICTOSIDINE SYNTHASE-LIKE	0.170
Potri 006G082900.1	98.38	Kinesin-like protein	0.218
Potri 010G014100.1	98.33	repeat-containing protein	0.141
Potri 015G112500.1	98.16	hydrolase domain-containing	0.126
Potri 014G156100.1	98.00	CSC1-like protein	0.137
Potri 001G113500.1	98.00	domain-containing protein	0.403
Potri 005G061600.1	97.99	glucuronate:xylan alpha	0.183
Potri 002G009100.1	97.98	Bromodomain-containing protein	0.230
Potri 010G164400.1	97.93	uncharacterized protein LOC105116966 isoform X1	0.156
Potri 006G149900.1	97.81	RNA-binding protein	0.150
Potri 001G253900.1	97.81	Protein TIC 62, chloroplastic	0.222
Potri 001G253800.1	97.67	Sugar transport protein	0.211
Potri 010G164600.1	97.66	Transmembrane emp24 domain-containing protein	0.171

These annotations are vague and will require more research to determine which

Discussion

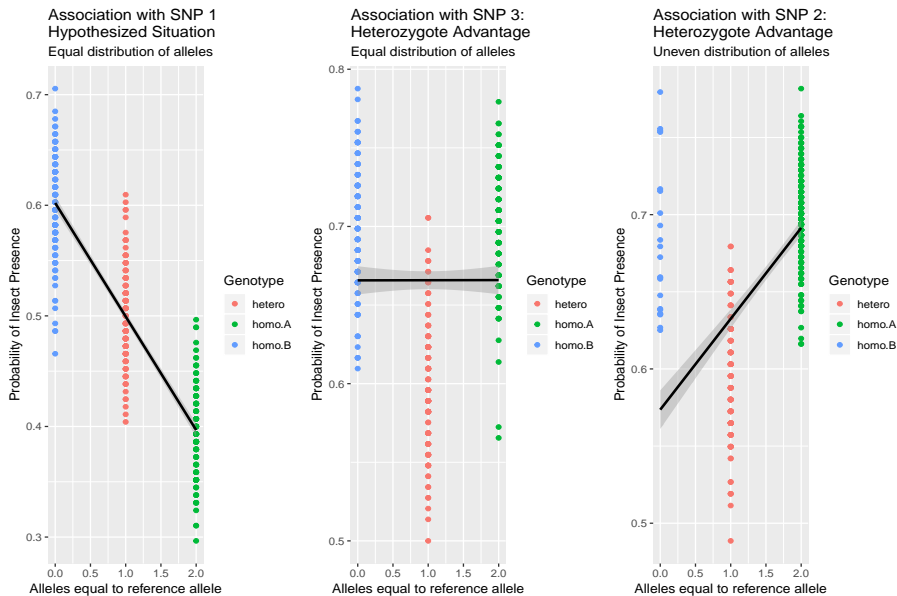
Conclusions

- Evidence of genomic regions that influence individual insects and communities
- These genomic regions are independent of the observed tree traits

Future Directions

- More thoroughly research gene annotations from congeneric *Populus trichocarpa* and model *Arabidopsis thaliana* to check biological function where possible
- Compare with other methods (multi-SNP, BLUP association, GBlups, etc)
 - A similar study on this garden found only 2 insect associations with < 12 total significant SNPs using BLUP association and a more lenient inclusion cutoff ($q < 0.1$) in 2015
- Include more tree traits and environmental variables and interactions

Extra Slides



BLUPs: Linear Mixed Model (LMM) Definition

Consider the following model for some response Y on observation j :

$$Y_j = \mu + X_j\beta + Z_j\tau + \varepsilon$$

- $E(Y_j = X_j\beta)$
- X_j and Z_j are vectors of independent variables for the j th observation
- β is an unknown vector of regression parameters (fixed effects)
- τ is an unknown vector of **random effects**
 - $E(\tau) = 0$ and $Var(\tau) = G$; $G = Cov(\tau_m, \tau_k)$ for all m and k
- ε is an unknown vector of random errors
 - $E(\varepsilon) = 0$ and $Var(\varepsilon) = R$

LMM: Parameter Estimation

$$Y_j = \mu + X_j\beta + Z_j\tau + \varepsilon$$

- $\tau \sim N(0, G)$, $\varepsilon \sim N(0, R)$, and $Cov(\tau, \varepsilon) = 0$
- Find coefficients $\hat{\beta}$ and $\hat{\tau}$ such that they:
 - minimize variance prediction error $Var(Y_j - \hat{Y}_j)$
 - are constrained by $E(Y_j - \hat{Y}_j) = 0$
($Bias(\hat{\theta}) = E(\hat{\theta}) - \theta$)
- $\hat{\beta}$ is a vector of best linear unbiased estimators (BLUEs) of fixed effects
- $\hat{\tau}$ is a vector of best linear unbiased predictors (BLUPs) of fixed effects
- conditional variances (covariance matrices for β and ε) are often unknown and are estimated as a nuisance parameter with Bayesian EM algorithm (nlme, lme4, etc.)

Take-Home Message

BLUPs are constrained coefficient estimates of random effects.

Modern GWA Models

- First fit LMM for a trait of interest (Size) with β_i fixed effects and τ_g random effects:

$$\text{Size} = \mu + \beta_1(\text{PGs}) + \beta_2(\text{CTs}) + \cdots + \beta_n(\text{Trait}_n) + \tau_g(\text{Genet}_g) + \varepsilon$$

- Then conduct a linear association analysis between the size-specific BLUPs τ and the Genotype of at SNP s of each genet g :

$$\tau_g = \mu + \alpha(\text{Genotype}_{s,g}) + \varepsilon$$

