

Establishing the Causal Relationship between Obesity, Insulin Resistance and Autophagy

Introduction & Significance

Type 2 diabetes is a metabolic disorder that hinders the body's ability to process insulin and disrupts metabolic homeostasis. It affected over 382 million people in 2013 (1), and has no known cure. Obesity, a cause of Type 2 diabetes, leads to metabolic stress that activates autophagy (5). Autophagy, a catabolic process that regulates energy homeostasis, correlates with fluctuations in both insulin resistance and obesity (2), but the causal relationship between this process and metabolic disorders is not clear. Using genetic mapping, we have scrutinized the relationships between autophagy related (Atg) genes and the phenotypes exhibited in Type 2 diabetes, and found that Atg5, Atg7, and Atg10 expression in the islet tissue and Insulin levels all are linked to chromosome 2. Using conditional scans and mediation analysis, we discovered that E113 is a likely mediator of Atg5 and Atg10 expression in the islet tissue and insulin levels, and, when downregulated, causes an increase in both autophagy and insulin levels.



Dataset

The data used for this research came from mice from a BTBR x B6 F2 cross. Alan D. Attie's Lab at the University of Wisconsin-Madison recorded this data set. The BTBR set results from a C57BL/6 and BTBR T+ tf strains of mice with Lepob

Specifications

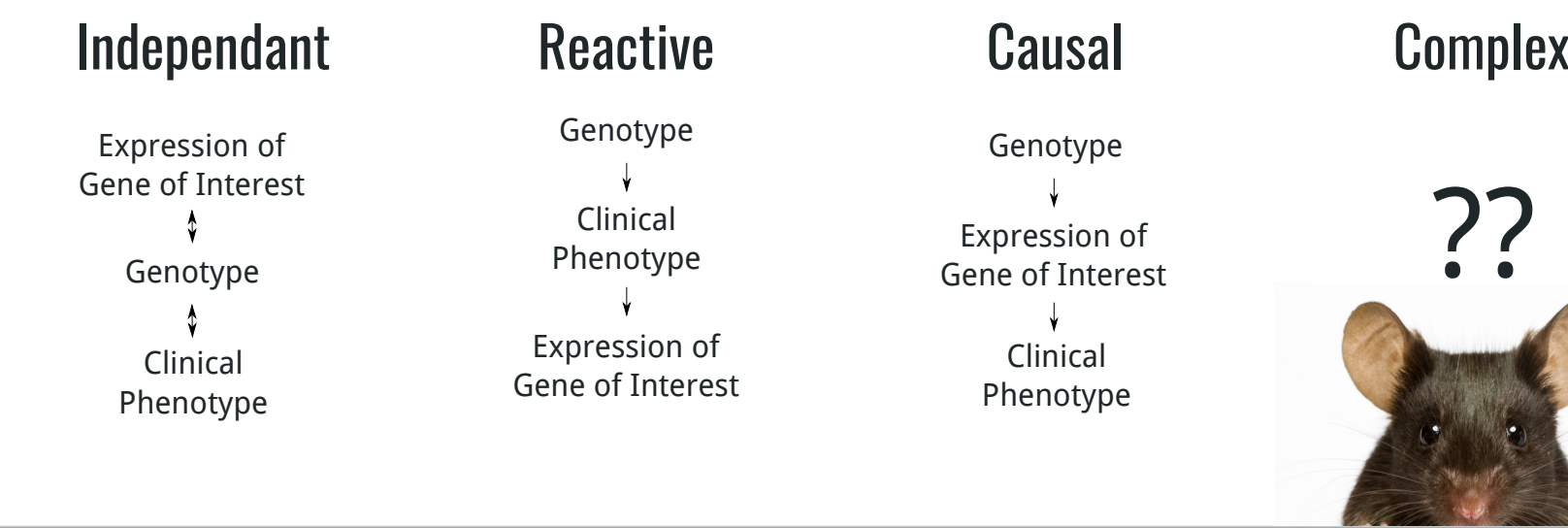
- > 516 mice
- > 144 quantitative phenotypes
- > 16,677 genes
- > 2057 genomic markers



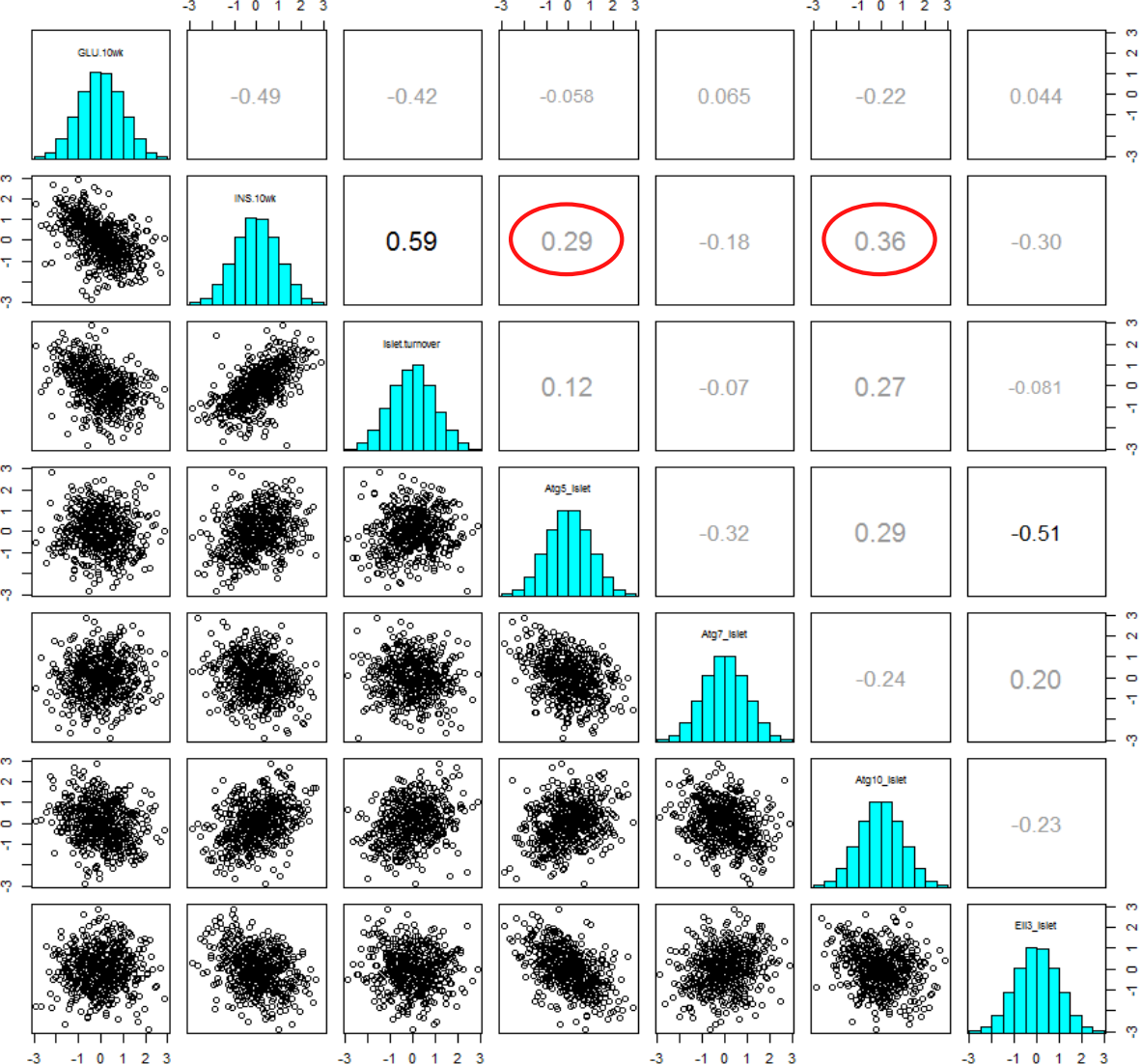
Methods

During this research, our team utilized **QTL analysis**. A QTL is a quantitative trait locus, which is a location on a chromosome that may be linked to a specific phenotype. QTL analysis is used to identify QTLs for phenotypes of interest.

Bayesian Information Criterion is a statistical method for model selection and the lowest BIC score is the preferred model. In our analysis, the models were:



Pairwise Scans



In this example, the correlation value of 0.29 between Atg5 and insulin levels suggests there is a relation between these two traits.

Other correlation values of note include a 0.36 correlation between Atg10 and insulin levels.

A pairwise scan plots each element of a cross object's correlation with every other element.



Mediation Analysis

The **anova** function allows one to determine causality to some extent.

```
> anova(m(atg5.islet ~ sex + Q2, data = F2gpheno))
Analysis of Variance Table

Response: atg5.islet
Df Sum Sq Mean Sq F value Pr(>F)
sex      1    0.06    0.062  0.0737 0.7861
Q2       2 37.21 18.6057 18.1275 2.58e-08 ***
Residuals 487 446.12    0.9161
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
> anova(m(e113.islet ~ sex + Q2, data = F2gpheno))
Analysis of Variance Table

Response: e113.islet
Df Sum Sq Mean Sq F value Pr(>F)
sex      1    0.06    0.062  0.0737 0.7861
Q2       2 70.08 35.0400 42.2251 <2e-16 ***
Residuals 487 409.32    0.840
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
> anova(m(atg5.islet ~ sex + atg5.islet + Q2, data = F2gpheno))
Analysis of Variance Table

Response: atg5.islet
Df Sum Sq Mean Sq F value Pr(>F)
sex      1    1.03    1.031  1.4307 0.2322
atg5.islet 1 125.80 125.802 197.7372 2.26e-16 ***
Q2       2    2.47    1.234  1.7119 0.1816
Residuals 486 350.31    0.721
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
> anova(m(e113.islet ~ sex + atg5.islet + Q2, data = F2gpheno))
Analysis of Variance Table

Response: e113.islet
Df Sum Sq Mean Sq F value Pr(>F)
sex      1    0.06    0.062  0.0937 0.7596
atg5.islet 1 125.80 125.802 197.7372 2.26e-16 ***
Q2       2 32.08 16.039 24.2515 9.130e-11 ***
Residuals 486 351.41    0.661
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

The above anova function output shows that Atg5 and E113 influence each other, and that E113 influences Atg5, not the other way around.

Bayesian Informaiton Criterion

The lowest BIC score is the likely model. If it does not differ by at least 5 from every other score, the test is inconclusive.

```
> with(F2gpheno,
+       triple.fic(e113.islet, INS.10wk, Q2))
independent reactive causal complex
2723.423 2747.453 2685.362 2695.148
> with(F2gpheno,
+       triple.fic(e113.islet, atg5.islet, Q2))
independent reactive causal complex
2701.133 2623.136 2580.950 2589.633
```

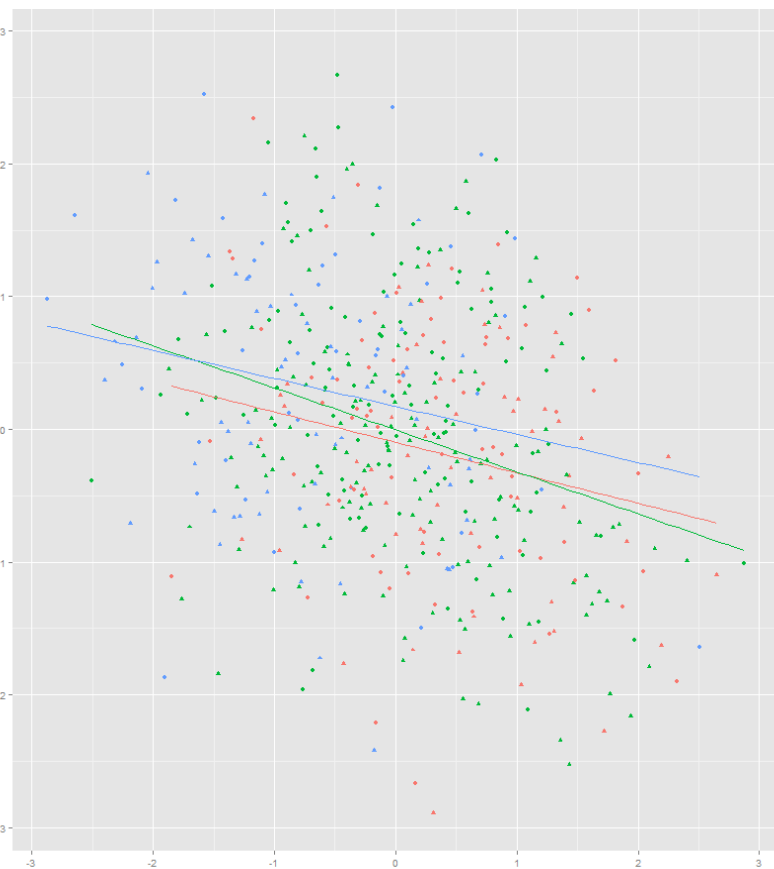
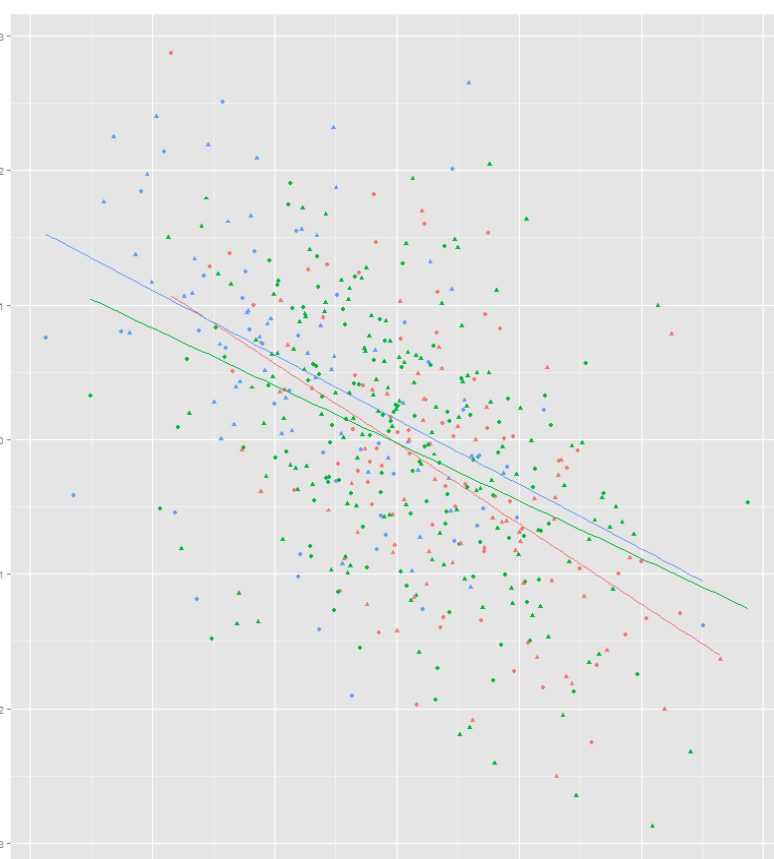
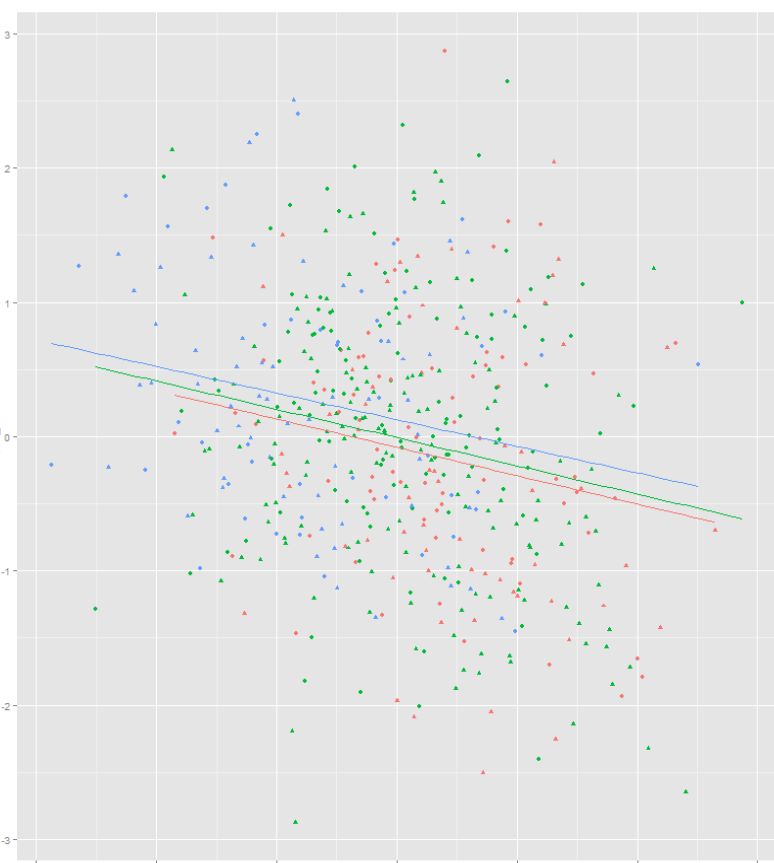
E113 is casual for both Atg5 and insulin, suggesting that it is a strong mediator of the two traits.

```
> with(F2gpheno,
+       triple.fic(e113.islet, atg10.islet, Q2))
independent reactive causal complex
2725.791 2769.810 2702.961 2713.164
```

E113 is also casual for Atg10.

```
> with(F2gpheno,
+       triple.fic(e113.islet, atg7.islet, Q2))
independent reactive causal complex
2717.094 2763.245 2717.764 2715.314
> with(F2gpheno,
+       triple.fic(atg5.islet, INS.10wk, Q2))
independent reactive causal complex
2753.322 2750.169 2722.139 2728.715
```

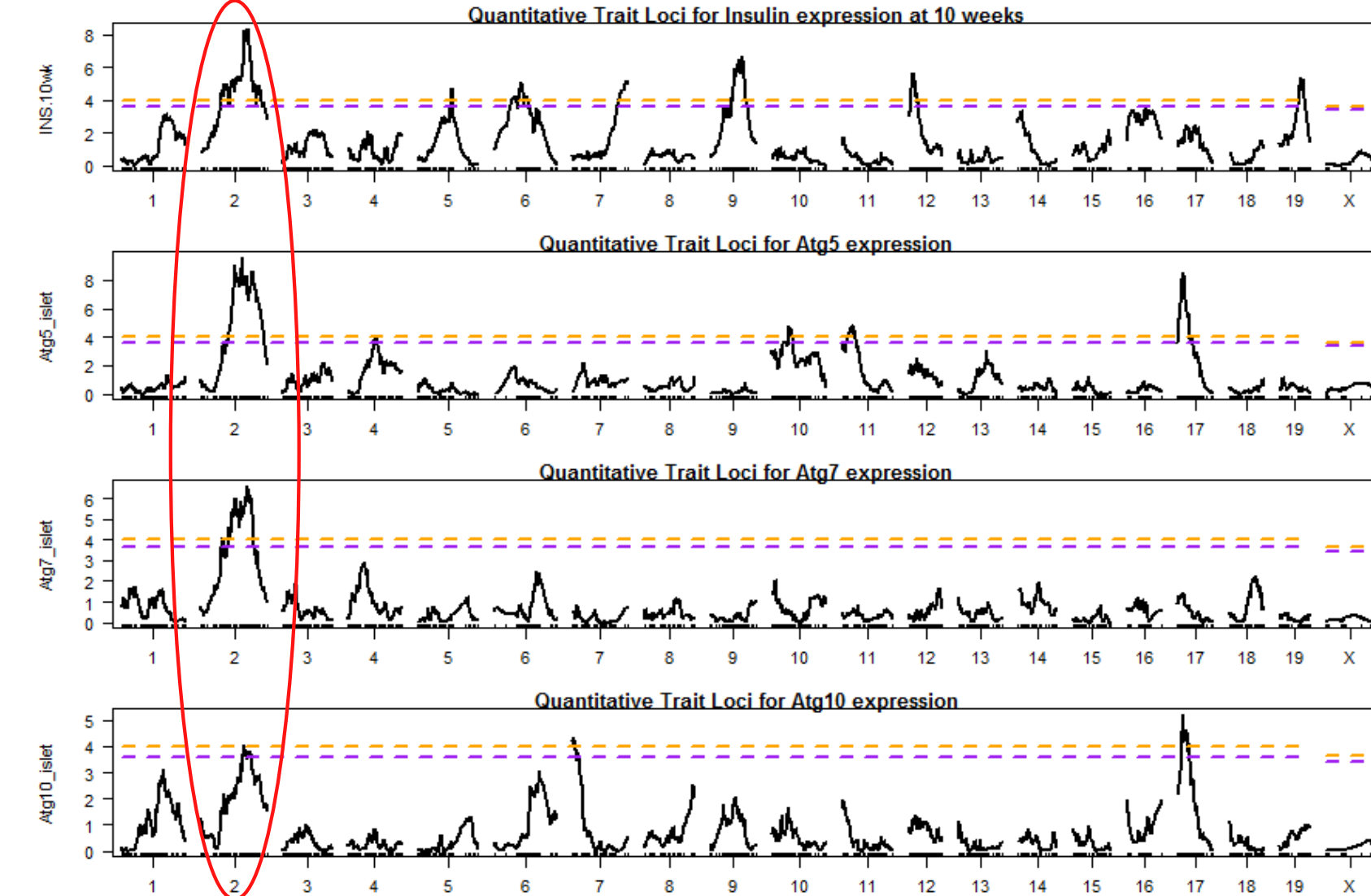
It's relationship with Atg7 is inconclusive.



Genome Scans

A LOD (logarithm of odds) Score is an estimate of the linkage between two genes. For a complex process like genetic trait analysis, 4 is an acceptable threshold.

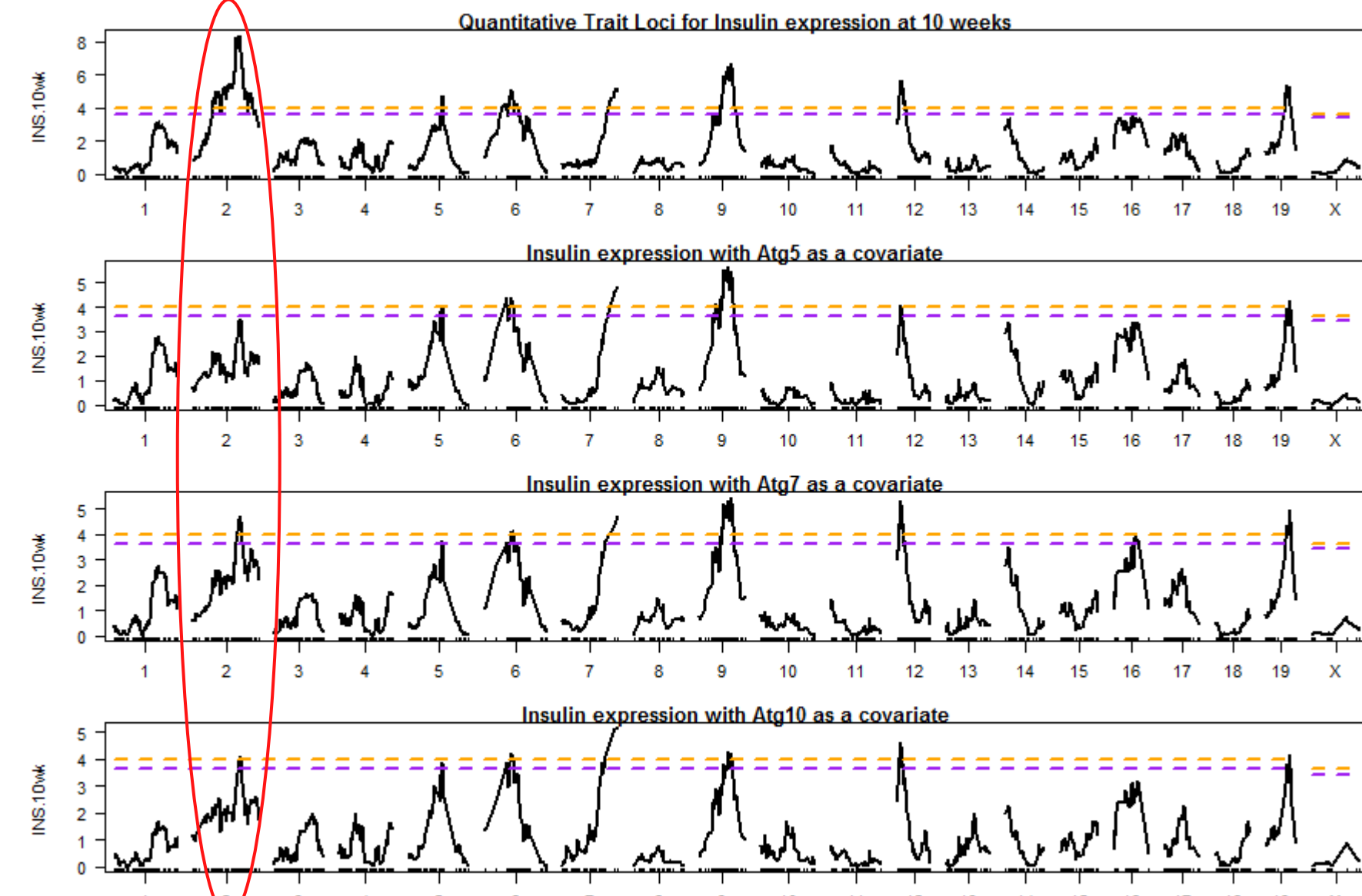
Genome scans like the one below display the LOD Scores for a trait at each point along the genome.



We wanted to discover which Atg proteins shared peaks with either insulin, glucose, or islet turnover, all of which are phenotypes associated with type 2 diabetes. Three different Atg proteins share a peak with insulin on chromosome 2 (seen above), which suggests that there are gene(s) on that chromosome that mediate both autophagy and insulin levels.

Conditional Scans

A conditional scan is a genome scan where a certain trait or gene is run as a covariate. This results in a LOD peak drop in magnitude proportional to the covariate's influence on the gene's expression.

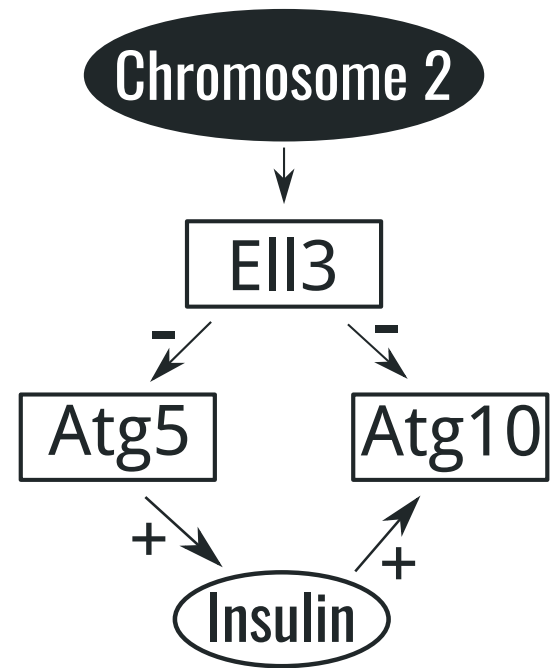
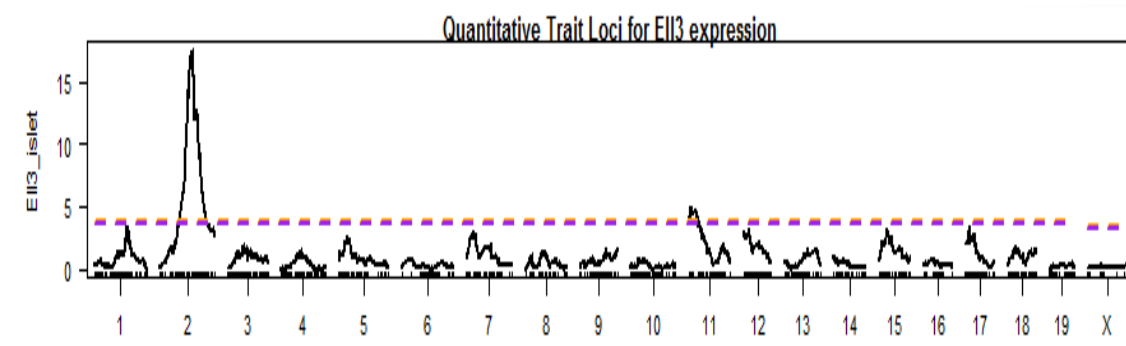


In this example, Insulin's peak on chromosome 2 is dropped below the significance threshold when run with covariates Atg5, Atg7, or Atg10, meaning that there is a gene on chromosome 2 that links Atg5, Atg7, Atg10 and insulin levels.



Conclusion

After using mediation analysis to determine which genes are causal, it is likely that E113 is a strong mediator for Atg5, Atg10, and insulin in islet tissue.

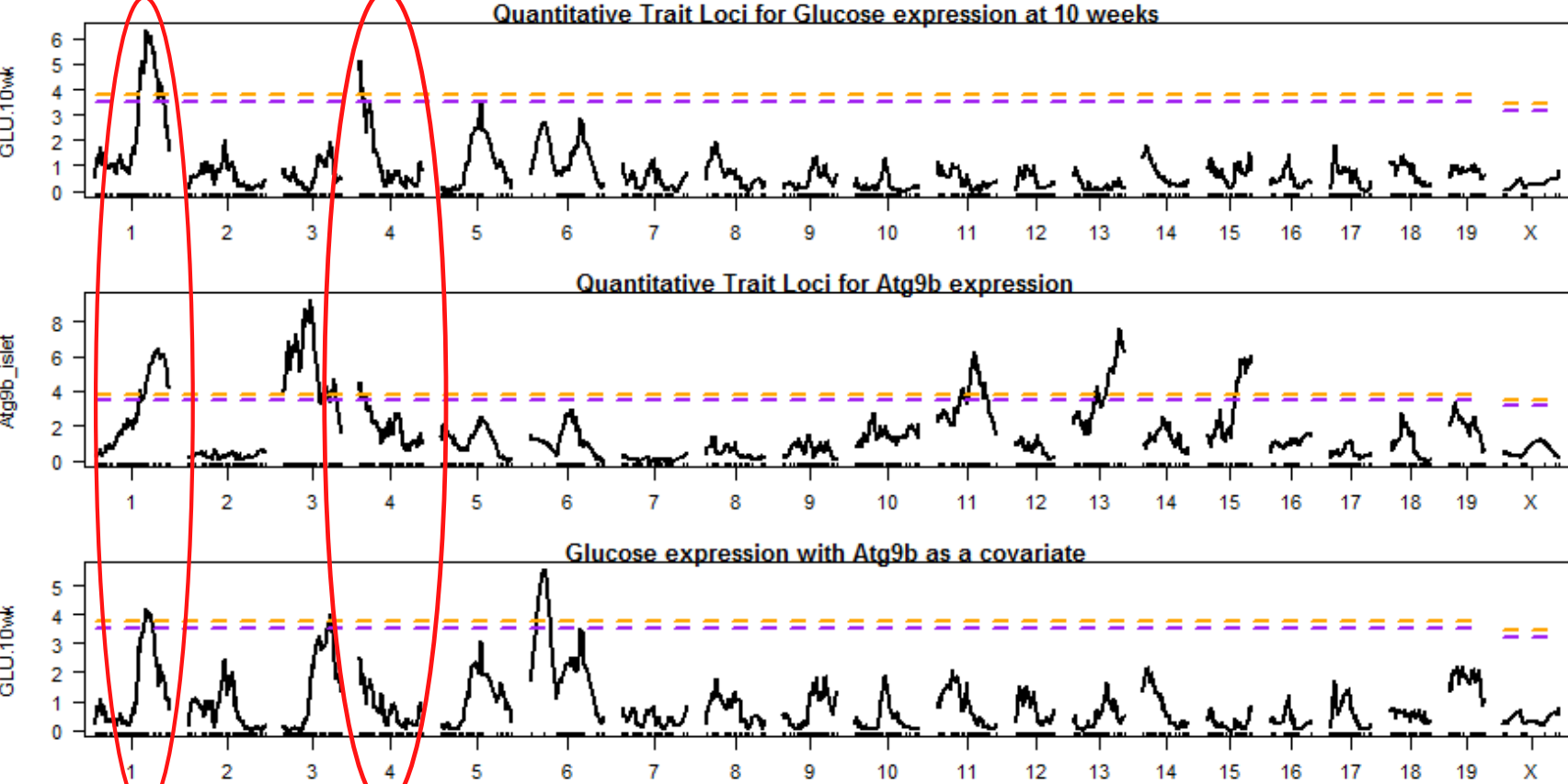


This is the likely QTL pathway, based on our project's results.

E113 is a protein coding gene that is thought to bind enhancers in stem cells, but also behaves like E11 (eleven-nineteen lysine-rich leukemia gene), which acts as a negative regulator of p53 and regulates cell proliferation and survival (3). E113's relation to autophagy also explains the result of a certain study, which showed that autophagy increases cell life and combats cancer formation (4). The latter provides evidence for the underlying idea behind our original hypothesis; that autophagy could help to maintain islet function in the face of Type 2 diabetes by increasing islet cell function and lifespan.

Future Research

Most research surrounding type 2 diabetes focuses on insulin levels and islet turnover, but glucose is another phenotype that is associated with diabetes. Shown by this genome scan, Atg9b causes glucose's LOD peaks on chromosomes 1 and 4 to drop, implying a correlation between Atg9b and glucose. Finding a mediator for both of these would further reveal the relationship between type 2 diabetes and autophagy.



Additionally, E113 was not proven a mediator for Atg5 or Atg10 in the kidney or the liver. This is unsurprising based on the research that backed our original hypothesis. However, it is definitely worth examining activity in other tissues.

Literature Cited

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