

Statistical Analysis of Pleiotropy between Obesity and Substance Dependence

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Cognitive
control

Prefrontal
cortex

Dopamine
reactive system

Ventral tegmental
area
Dorsal /ventral
striatum
Habenula
Thalamus
Amygdala
Hippocampus

Energy
homeostasis

Hypothalamus

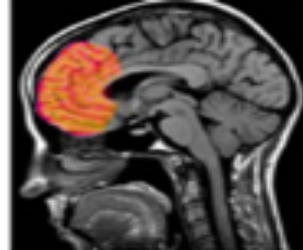
Peripheral and central
messengers



FOOD



DRUGS



Data

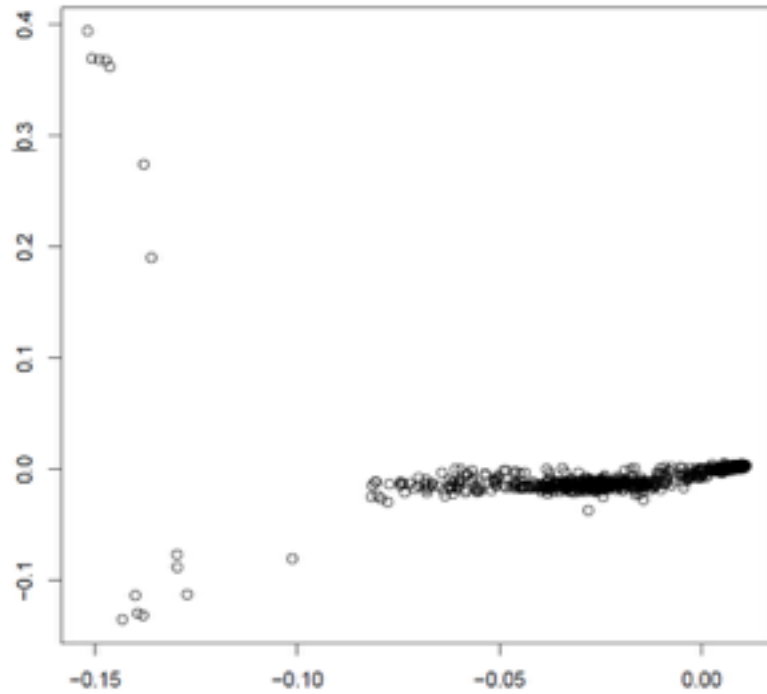
- **SSADDA:** 2379 European Americans
- **SAGE:** 2668 European Americans
- **Phenotype:** BMI, Substance dependence symptom score;
- **Genotype:** 988,306 SNPs (SSADDA)

Quality Control

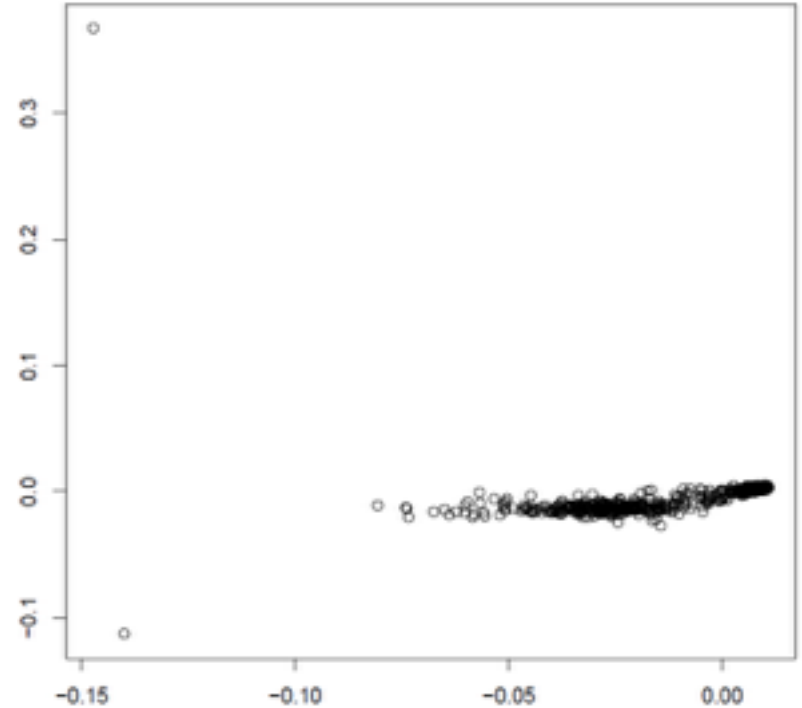
- a. Misidentified individuals
- b. Genotype failure rate 0.02
- c. Extreme heterozygosity (+/-3 sd)
- d. Duplicated or related individuals
 - Sample QC
 - (2379=>1828)



- a. MAF 0.01
- b. HWE $1e-06$
- c. Genotype missing rate 0.02
- d. Unbalanced genotype rates between case/control
 $p=1e-05$
 - SNP QC
 - (988,306=>805,782)



Before QC



After QC

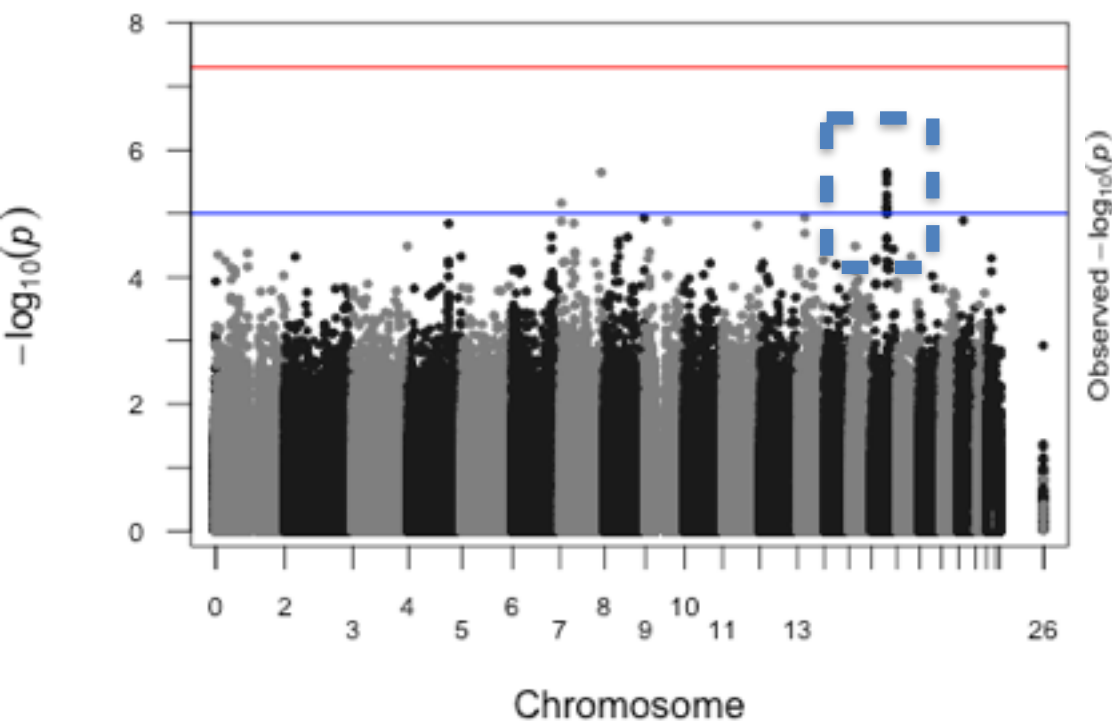
From PCA plots, most suspected outliers have been removed in the quality control (QC) process.

Single Marker Association Analysis

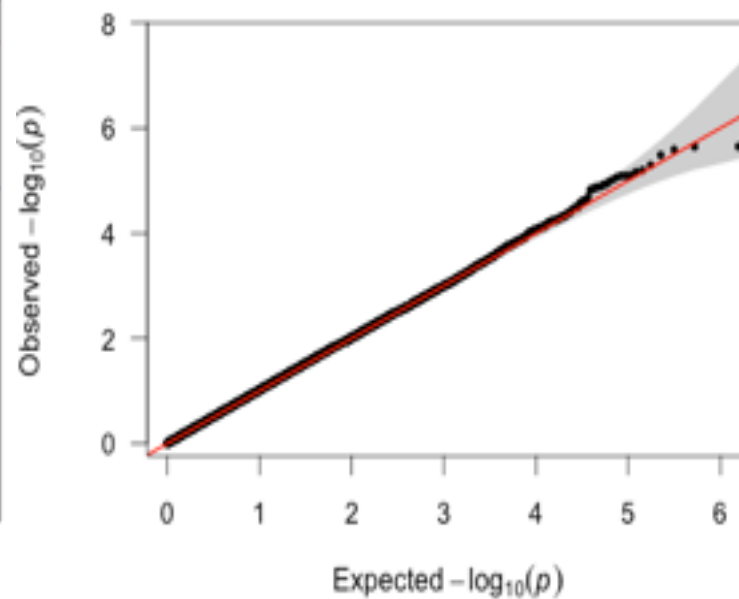
- **Genotype Model:** additive model
 - Assume there is a linear increase of risk with each additional risk allele.
- **Test Approach:** linear regression
- **Covariates:** adjusted in linear regression
 - Age and sex
 - first 4 scaling factors from MDS analysis (for population stratification)

Outcome: BMI

rs1121980



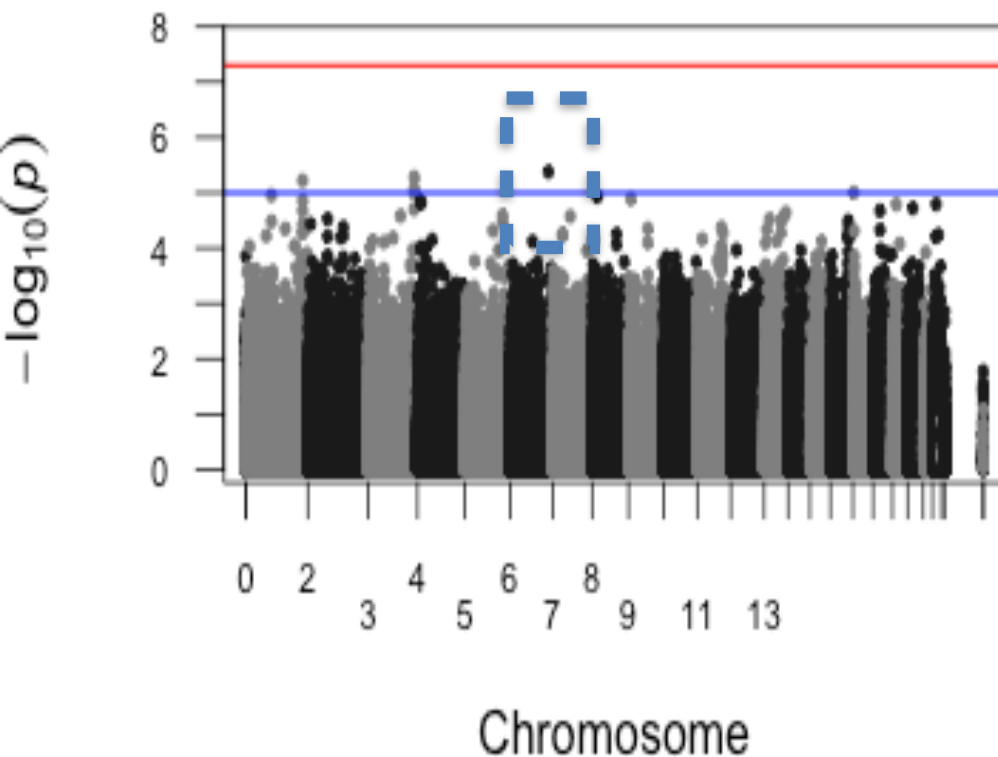
Inflation factor $\lambda=1.02$



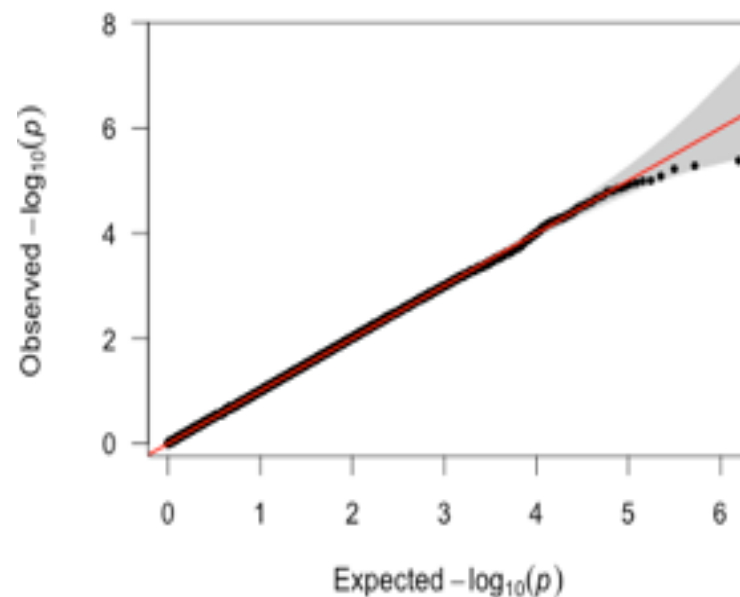
SNP	CHR	Nearest Gene	Beta	P-value
rs1121980	16	FTO	0.9207	2.26E-06

Outcome: Sub_Dep

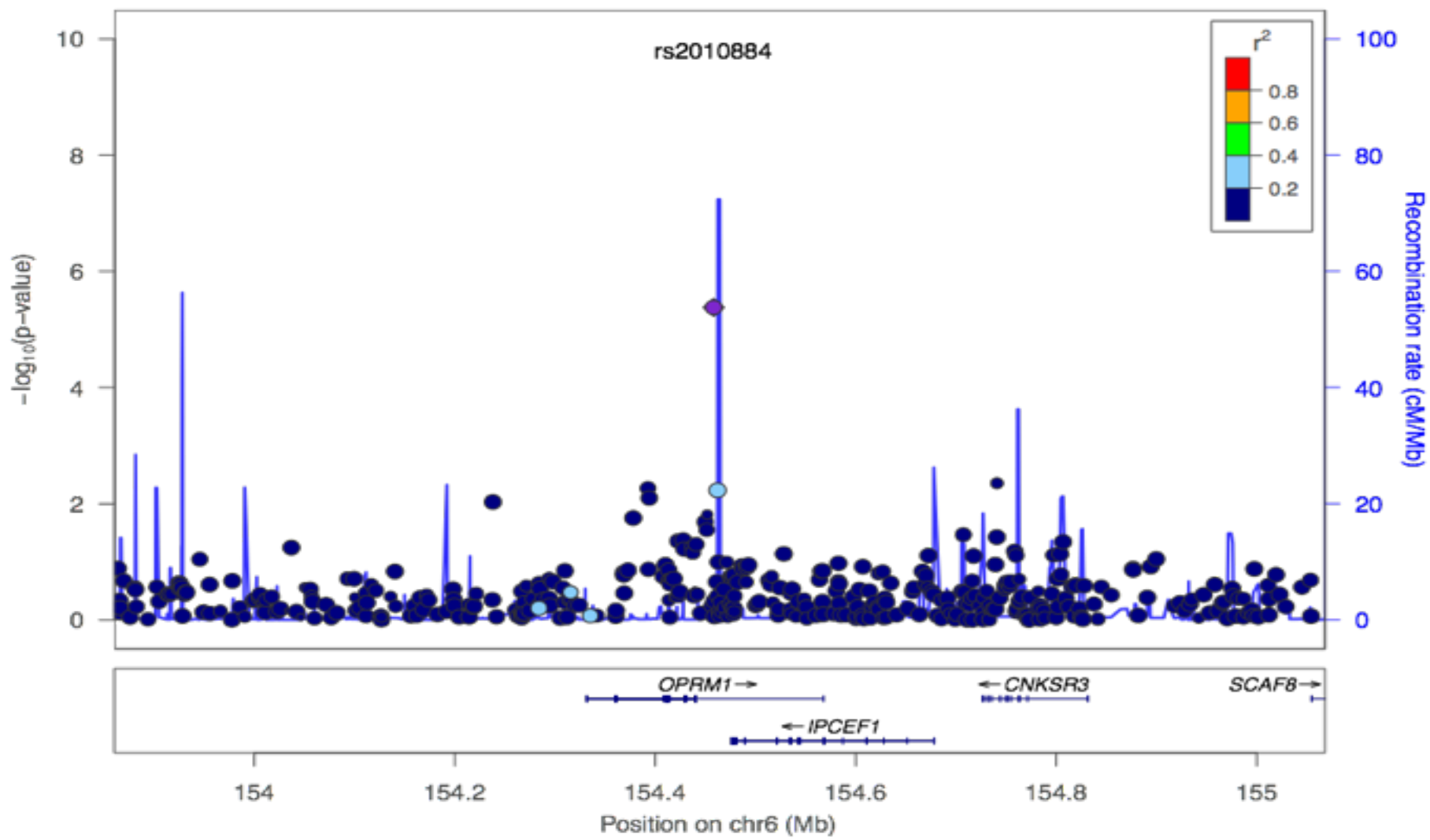
rs2010884



Inflation factor $\lambda=1.007$



Plotted SNPs



SNP	CHR	Nearest Gene	Beta	P-value
rs2010884	6	OPRM1	-1.39	4.18E-06

Mixed Effects Model Based Analysis

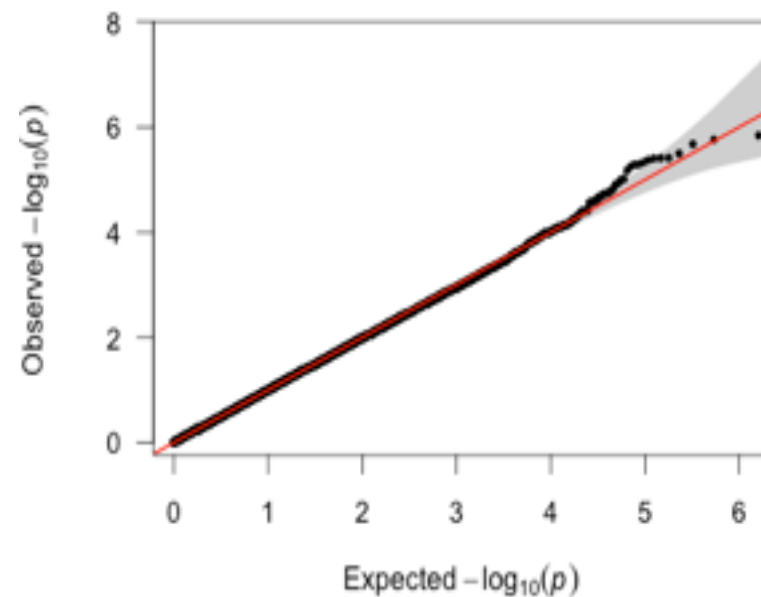
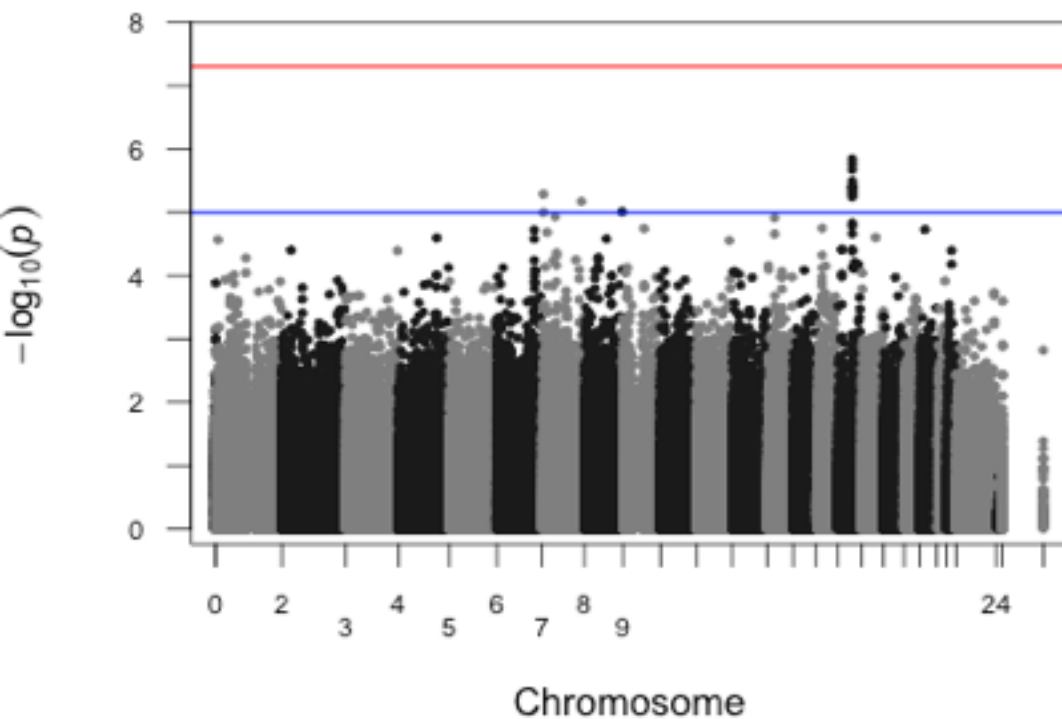
$$Y = X\beta + u + \varepsilon$$

- Y =phenotypes
- X =SNP genotypes (+covariates)
- $Var(u) = \sigma_g^2 A$, where A is the genetic relationship matrix (GRM)

$$A_{jk} = \frac{1}{M} \sum_i \frac{(g_{ij} - 2p_i)(g_{ik} - 2p_i)}{2p_i(1 - 2p_i)}$$

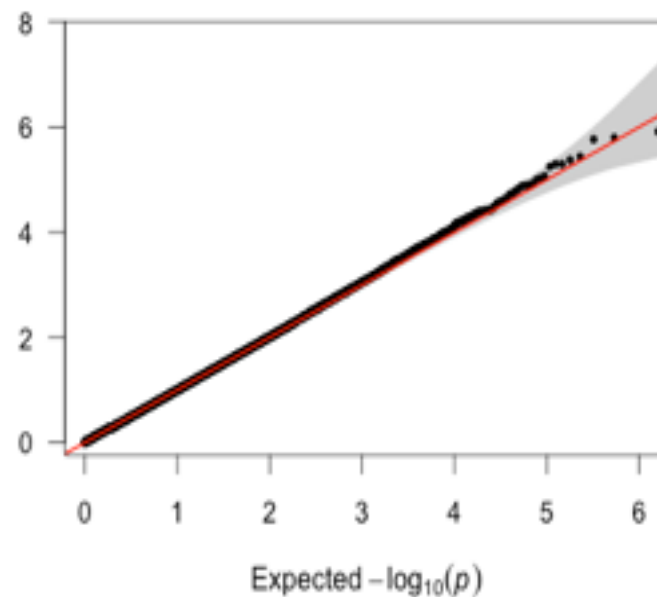
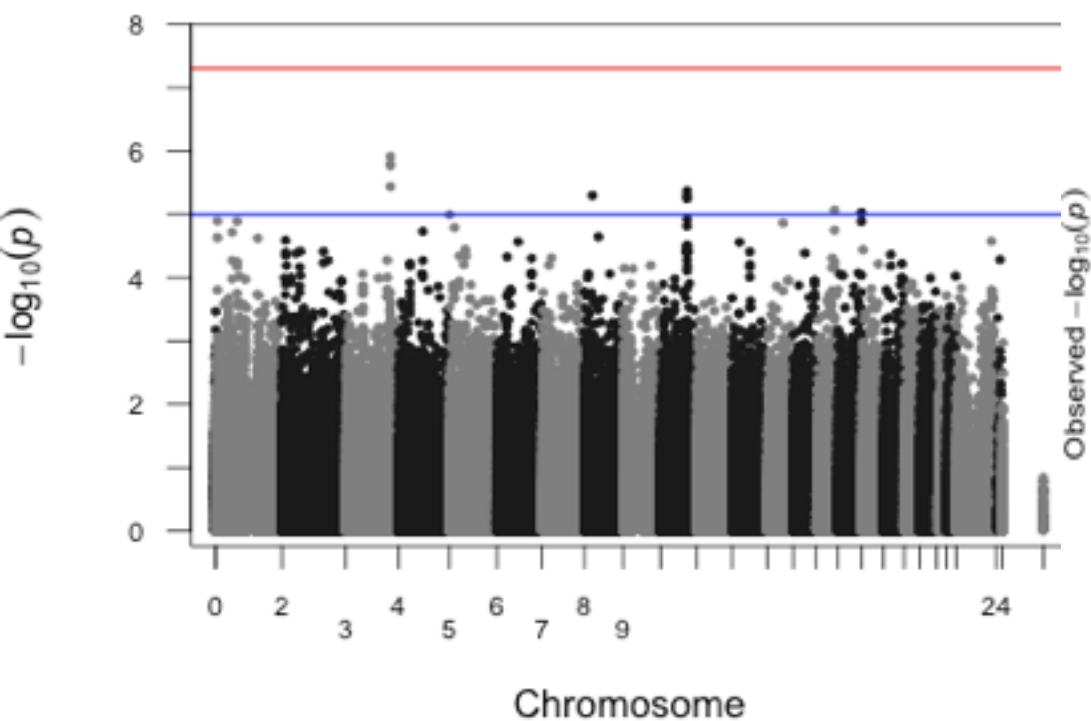
Outcome: BMI

rs1121980



Outcome: Sub_Dep

rs2010884



Heritability Estimates

$$Y = X\beta + u + \varepsilon$$

- $Var(u) = \sigma_g^2 A$, σ_g^2 is the variance explained by all the SNPs
- Estimated by the restricted maximum likelihood (REML) approach

Phenotype	N	Hg	SE	LRT	P-value
BMI	1828	0.2595	0.16	2.917	0.0438
Sub_Dep	1828	0.2156	0.16	1.890	0.0846

SNP Coheritabilities

- The variance-covariance matrix across the two traits is:

$$V = \begin{pmatrix} Z_1 A Z_1' + I \sigma_{g1}^2 & Z_2 A Z_1' \sigma_{g12} \\ Z_1 A Z_2' \sigma_{g12} & Z_2 A Z_2' + I \sigma_{g2}^2 \end{pmatrix}$$

- The genetic correlation coefficient is:

$$r_g^{SNP} = \frac{\sigma_{g12}}{\sigma_{g1} + \sigma_{g2}}$$

- Estimated by the Bivariate REML approach

	N	r_G	S.E.	P-value
BMI:Sub_Dep	1828	0.2408	0.41	0.71

Integrative Analysis of Two GWAS Datasets with Functional Annotations

- We have P-values from two independent GWAS datasets
- Indicator variable $Z_j = [Z_{j00}, Z_{j10}, Z_{j01}, Z_{j11}]$ for the j-th SNP: e.g, Z_{j11} means the j-th SNP is associated with both BMI and Sub-Dep
- Functional annotation data: $A \in \mathbb{R}^{M \times N}$, where $A_j \in \{0,1\}$ indicates whether the j-th SNP is functionally annotation.

- Model the relationship between Z_j and A_j as:

$$q_{00} = \Pr(A_j = 1 \mid Z_{j00} = 1)$$

$$q_{10} = \Pr(A_j = 1 \mid Z_{j10} = 1)$$

$$q_{01} = \Pr(A_j = 1 \mid Z_{j01} = 1)$$

$$q_{11} = \Pr(A_j = 1 \mid Z_{j11} = 1)$$

- The joint distribution of $\Pr(P, A)$ can be estimated by EM algorithm

$$\Pr(P, A) = \prod_{j=1}^M \left[\sum_{l \in \{00,10,01,11\}} \Pr(Z_{jl} = 1) \Pr(P_j, A_j \mid Z_{jl} = 1) \right]$$

- The summary statistics of two phenotypes:
BMI: 805,782 p-values;
Substance Dependence: 845,871 p-values;
- Overlapping SNPs of two phenotypes is 466,115
- Using the central neural system gene as annotation data, 63,274 (13.6%) of the SNPs were annotated

	00	10	01	11
$\hat{\pi}$	0.911(0.086)	0.046(0.053)	0.04(0.084)	0.02(0.049)
\hat{q}	0.126(0.013)	0.213(0.094)	0.268(0.086)	0.288(1.843)

Conclusion

- The strongest association signal for obesity: FTO gene;
- The strongest association signal for substance dependence: OPRM1 gene
- h_g^2 estimated for obesity was 0.26, 0.22 for substance dependence
- No evidence suggests pleiotropy between obesity and substance dependence in this data set.