## Prediction of cannabis use disorder severity from genetic and behavioral data

Ariel Ketcherside Milind Rao Shikha Prashad

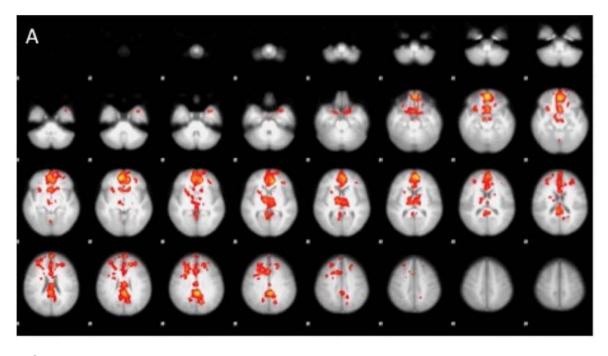
## Substance use disorders (SUDs)

- Continued use despite knowledge of negative consequences
  - Negative health consequences
  - Impacts on social relationships
- Neurological mechanisms of SUDs
  - Involve reward pathway
  - Can be examined through reports of behavior
  - Have a strong genetic component
    - Cannabis: 51–59% heritable (Agrawal et al. 2012)



## Cannabis use disorders (CUDs)

- 10% of cannabis users develop a cannabis use disorder
- Cannabis activates reward circuitry in the same way as other drugs of abuse



Cannabis users > controls when viewing cannabis-related stimuli

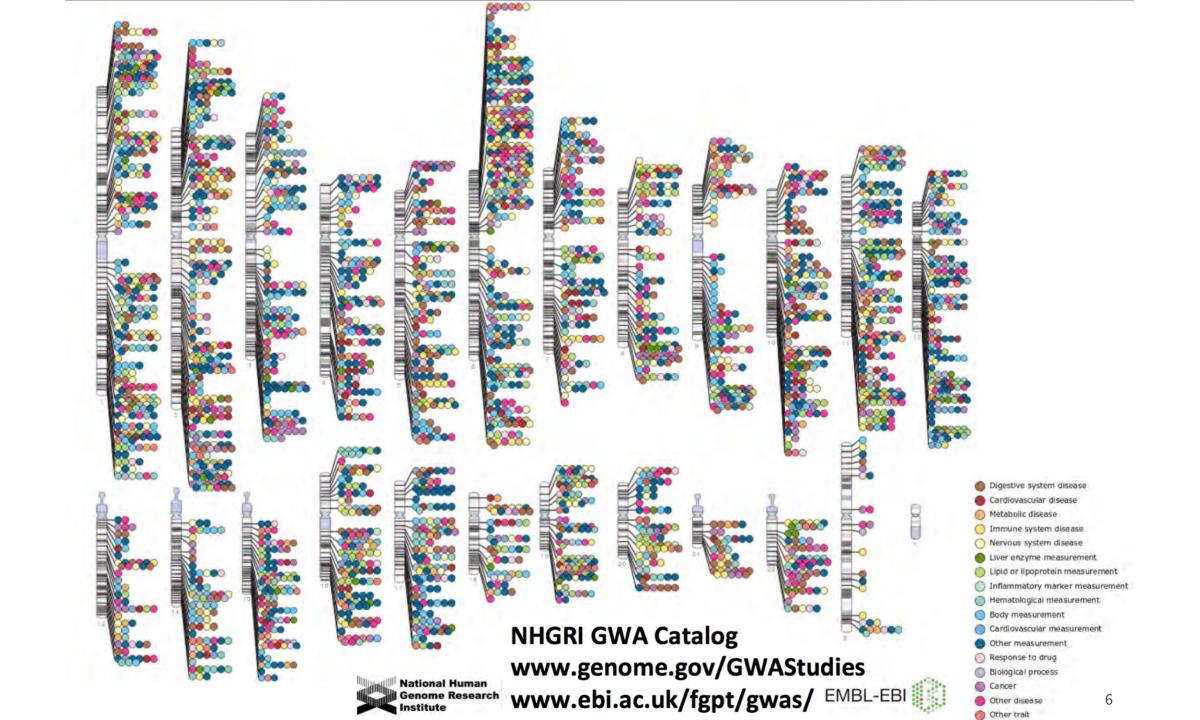
Filbey et al. 2016, Human Brain Mapping

#### Genetics 101

- Single nucleotide polymorphisms (SNPs) create genetic variability among individuals
- Genetic variability encodes molecular variability that influences neural processes
  - Motivation
  - Reward
  - Impulsivity
- We need a way to sample large quantities of SNP variability

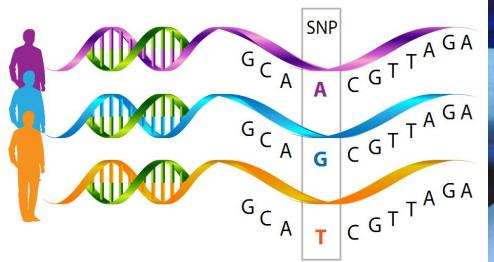






# Genome Wide Association Studies (GWAS)

- Single nucleotide polymorphisms
- Measured on high-throughput genotyping plates





#### Our data

- Our chip assesses 800,000 different loci on the genome
- Marijuana Problem Scale: Measure of cannabis use disorder severity (CUD)
  - E.g. "how often has marijuana caused you to have fights with family or friends?"
- Marijuana Craving questionnaire (MCQ)
  - Questions administered after 3 days of abstinence from cannabis
  - E.g. how much would you like to smoke marijuana right now?
- Participants: 235 cannabis users

### Aims: Clinical Application

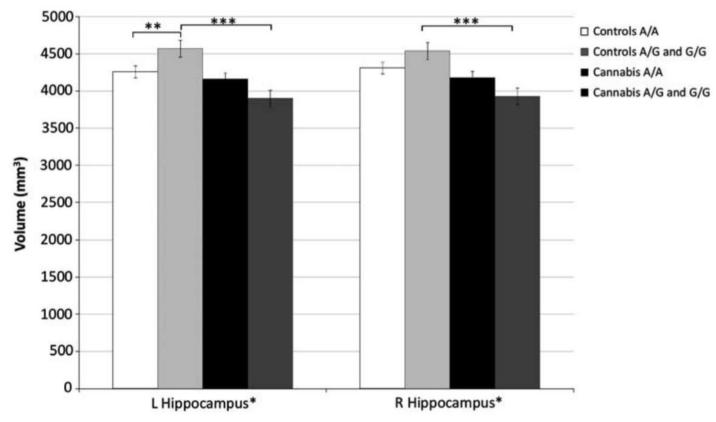
- Aim 1: To predict severity of cannabis use disorder based on their SNPs
- Which SNPs contribute most to CUD?
- Are there interaction effects?

- Aim 2: Determine whether severity can be more accurately predicted by including measures of craving
- Are some alleles related to greater reported subjective craving after 3 days of abstinence?

## Genetics Analysis Approaches

### Candidate allele studies

- One SNP
- Ex: rs2023239 in cannabis users



Schacht, et al. 2012

# Multilocus Genetic Profile Approach (MLGP)

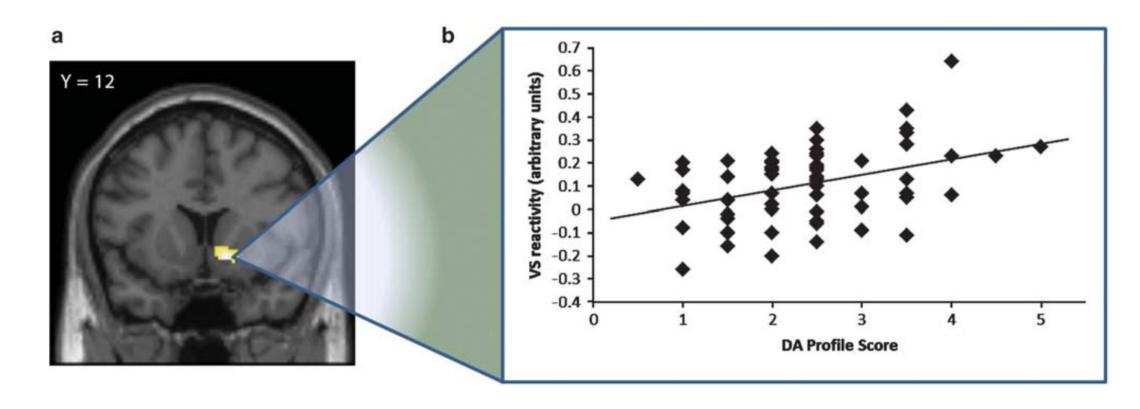
- *a priori* collection of specific SNPs
- used to calculate a risk score
- Limited to a linear model which may not encompass size of contribution of SNP

**Table I** Composition and Distribution of Multilocus Genetic Profile Scores

Polymorphism	Genotypes	N	DA profile score
DRD2 -141C Ins/Del	Del carrier	14	High
	Ins/Ins	55	Low
DATI VNTR	9-repeat carrier	35	High
	10/10	34	Low
DRD4 VNTR	7-repeat carrier	42	High
	All others	27	Low
COMT Val <sup>158</sup> Met	Met/Met	12	High
	Val/Met	41	Intermediate
	Val/Val	16	Low
DRD2 Taq1A	C/C	43	High
	C/T	23	Intermediate
	T/T	3	Low

Nikolovia et al. 2011

## Multilocus Genetic Profile Approach



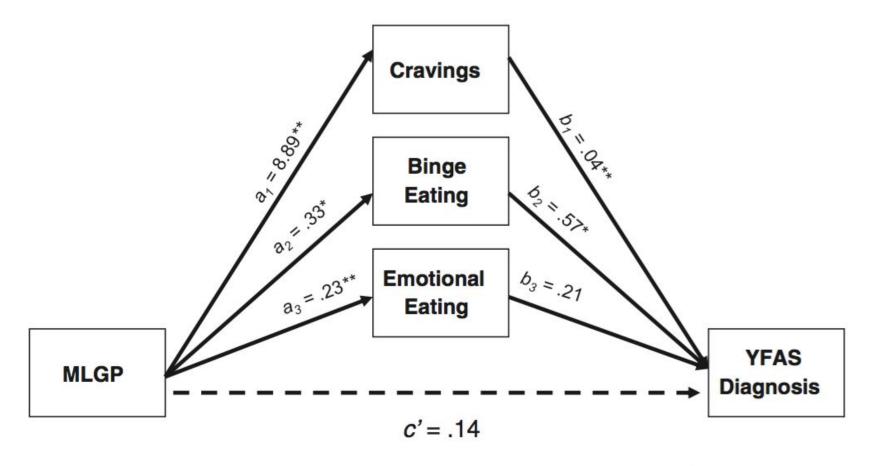
Nikolovia et al. 2011

### MLGP: Food addiction

Polymorphism	Genotypes	N	DA profile scores	
DRD2 Taq1A	A1+	54	Low	
	A1 —	66	High	
DRD2 -141C Ins/Del	Del+	39	High	
	Ins/Ins	81	Low	
DAT1 VNTR	9-repeat	56	High	
	10/10	64	Low	
COMT Val <sup>158</sup> Met	Met/Met	14	High	
	Val/Met	58	Intermediate	
	Val/Val	48	Low	
C957T	C-	26	High	
	C+	94	Low	
rs12364283	T+	23	High	
	T-	97	Low	

**Davis 2013** 

#### MLGP: Food addiction



Total Indirect effect (bias corrected 95% CI: .08–1.12; NagelKerke R<sup>2</sup>=0.58; N=117)

#### Few studies can focus on all GWAS loci

Gene Name	Polymorphism*	Model (Major allele > Minor Allele)	Cases vs. Controls (independent samples)	Fixed Effects OR (95% CI)	Random Effects OR (95% CI)	Heterogeneity <i>p</i> -Value	I- Square	Grade**
SLC4A7	rs3278	G > A	1410 vs. 906 (3)	2.34 (1.599-3.420)	2.28 (1.555-3.333)	0.51	0	В
DRD4	48-bp repeat	Other > 7/8 repeats	2324 vs. 1932 (6)	1.44 (1.155-1.804)	1.48 (1.000-2.197)	0.06	52	С
DRD2/ ANKK1***	Taq1A	A2 > A1	6312 vs. 7424 (20)	1.30 (1.192-1.410)	1.38 (1.096-1.733)	< 0.0001	84	С
BDNF	rs6265	G > A	2530 vs. 4126 (9)	1.31 (1.165-1.451)	1.38 (1.056-1.790)	< 0.0001	80	C
CCK	-45 C/T	C > T	860 vs. 2002 (6)	1.34 (1.089-1.650)	1.34 (1.083-1.646)	0.62	0	В
FAAH	rs324420	P > T	498 vs. 1570 (3)	1.38 (1.014-1.875)	1.32 (0.807-2.171)	0.24	28	В
OPRM1	rs1799971	A > G	2846 vs. 4072 (9)	1.24 (1.090-1.410)	1.31 (0.958-1.790)	< 0.0001	80	C
COMT	rs4680	Val > Met	862 vs. 1594 (3)	0.76 (0.634-0.923)	0.82 (0.644-1.051)	0.71	0	В
CNR1	(AAT)n	14 repeats > Other	2304 vs. 2144 (8)	0.76 (0.658-0.878)	0.75 (0.619-0.906)	0.17	32	В
HNMT	rs35953316	Thr > Ile	1540 vs. 1306 (3)	0.76 (0.598-0.975)	0.72 (0.444-1.179)	0.04	70	C
OPRK1	rs702764	A > G	292 vs. 246 (3)	0.62 (0.431-0.901)	0.62 (0.412-0.944)	0.99	0	В
OPRM1	C691G	C > G	796 vs. 786 (3)	0.52 (0.416-0.647)	0.61 (0.330-1.095)	0.0025	83	C

<sup>\*</sup>Variants were ranked based on the summary ORs. \*\*Degree of 'epidemiological credibility' based on published protocols (A, strong; B, modest; C, weak; see Methods for more details). \*\*\*Researchers previously associated the polymorphism Taq 1A to the *DRD2* gene. However, the polymorphism sits in an exon of the *ANKK1* gene.

Li et al. 2011

#### **GWAS Problems**

- Large number of alleles can lead to p hacking issues
- Small minor allele frequency → small effect size
- Genotyping currently costs ~\$300

## Methods

#### Work Flow

A priori addiction-specific words via Google Scholar search

Scrape SNPedia for SNPs related to these words

 Online database comprising over 87000 SNPs and research findings about them

Match relevant SNPs with the SNPs in our database

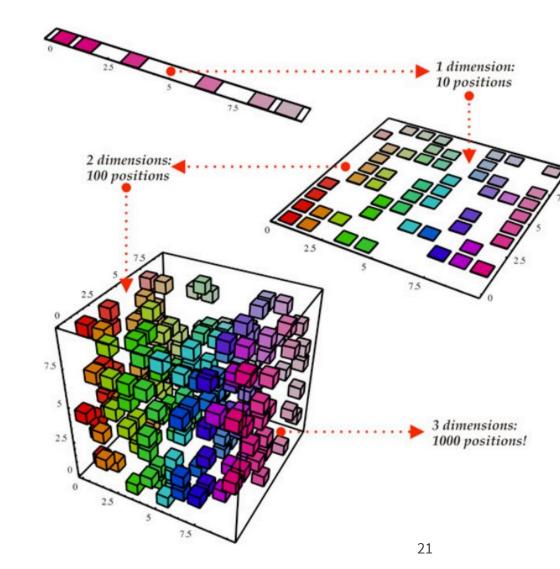
• 235 cannabis users

Determine risk alleles for CUD (additive risk, and risk relative to other SNPs)

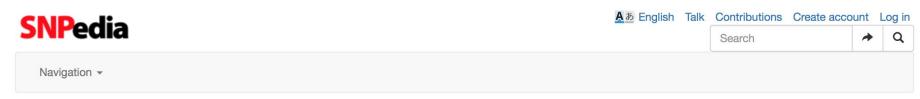


## Strategic Dimension Reduction

- Previous genetics studies either
  - Focus on one locus
  - Require a large sample size to account for variance in multi-locus models
  - In this study, we use the body of extant literature to strategically reduce the volume of data to combine the best of these two approaches



#### Tools: SNPedia



Page Discussion Edit with form Edit History

Have questions? Visit https://www.reddit.com/r/SNPedia

#### rs2023239

[PMID 187056883] rs2023239 influences cannabis dependence [PMID 200105523] Individual and additive effects of the CNR1 and FAAH genes on brain response to marijuana cues

[PMID 22669173] Associations between Cannabinoid Receptor-1 (CNR1) Variation and Hippocampus and Amygdala Volumes in Heavy Cannabis Users

[PMID 16741937 1] Association study of the CNR1 gene exon 3 alternative promoter region polymorphisms and substance dependence.

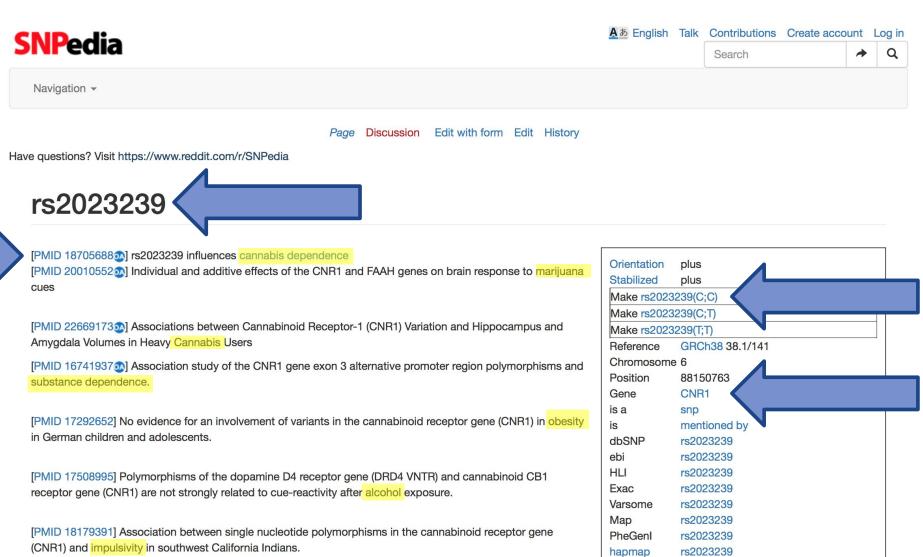
[PMID 17292652] No evidence for an involvement of variants in the cannabinoid receptor gene (CNR1) in obesity in German children and adolescents.

[PMID 17508995] Polymorphisms of the dopamine D4 receptor gene (DRD4 VNTR) and cannabinoid CB1 receptor gene (CNR1) are not strongly related to cue-reactivity after alcohol exposure.

[PMID 18179391] Association between single nucleotide polymorphisms in the cannabinoid receptor gene (CNR1) and impulsivity in southwest California Indians.

Orientation plus Stabilized plus Make rs2023239(C;C) Make rs2023239(C;T) Make rs2023239(T;T) Reference GRCh38 38.1/141 Chromosome 6 Position 88150763 CNR<sub>1</sub> Gene is a snp is mentioned by **dbSNP** rs2023239 ebi rs2023239 HLI rs2023239 Exac rs2023239 rs2023239 Varsome Мар rs2023239 PheGenI rs2023239 rs2023239 hapmap 1000 rs2023239

#### Tools: SNPedia



1000

rs2023239

23

## Determining Risk Alleles

- Interesting alleles selected from behavioral keywords
- Alleles selected with knowledge of specific pathways (eg. Cannabinoid pathways)
- Use the Least Absolute Shrinkage and Selection Operator (LASSO) technique to determine most relevant alleles for determining risk

#### Variable Selection Procedure - Lasso

- Let b represent the behaviour score of user i
- s<sub>j</sub> is the number of minor alleles in SNP j
  α<sub>i</sub> is the weight attached to SNP a measure of relevance
- We solve the Lasso problem to select a small number of non-zero relevant SNPs

$$\min_{\alpha} \sum_{i \in \text{users}} (b_i - \sum_{j \in \text{SNPs}} \alpha_j s_j)^2 + \lambda \sum_{j \in \text{SNPs}} |\alpha_j|$$

## Learning Interactions

- Interactions: SNP s<sub>j1</sub>, s<sub>j2</sub> may not have an effect individually.
  When minor alleles of both present, there may be more of an effect
- We consider a variant of Lasso where we consider the effects of these pairs and find sparse coefficients  $\alpha_{i,i'}$

$$\min_{\alpha} \sum_{i \in \text{users}} (b_i - \sum_{j,j' \in \text{SNPs}} \alpha_{j,j'} s_j s_{j'})^2 + \lambda \sum_{j,j' \in \text{SNPs}} |\alpha_{j,j'}|$$

## Expected outcomes

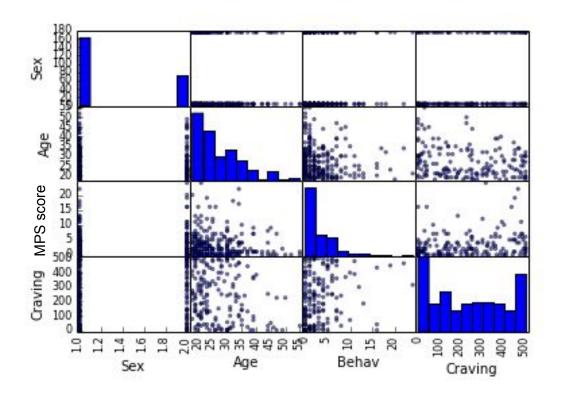
 A solution to getting a signal from high dimensional data in small sample sizes

 These data will delineate genetic variability that can predict CUD

## **Preliminary Results**

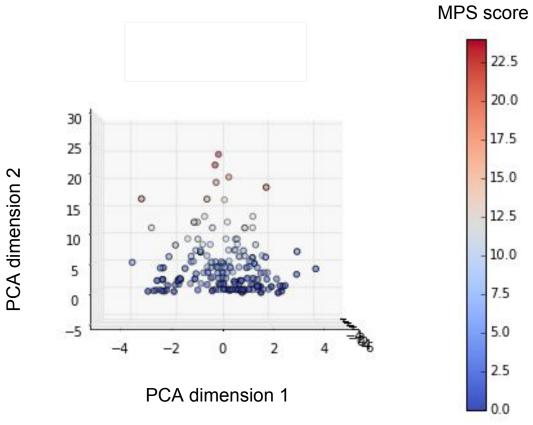
Distributions of variables of interest

- Greater number of males than females in sample
- Craving not correlated with other variables
- Small correlation between craving and MPS score

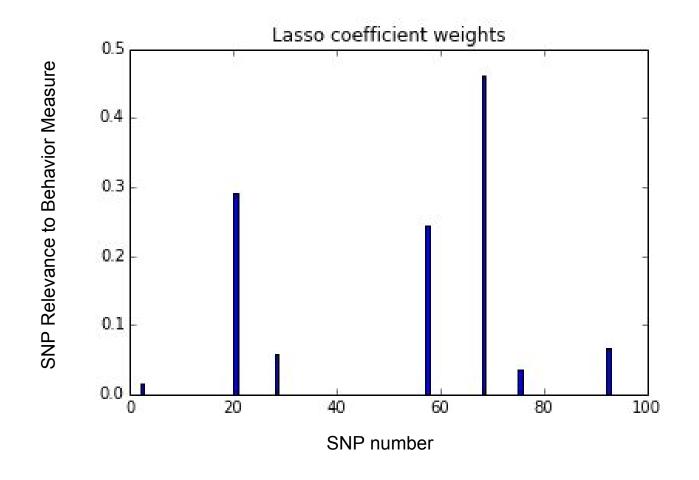


# Preliminary results: Principal component analysis

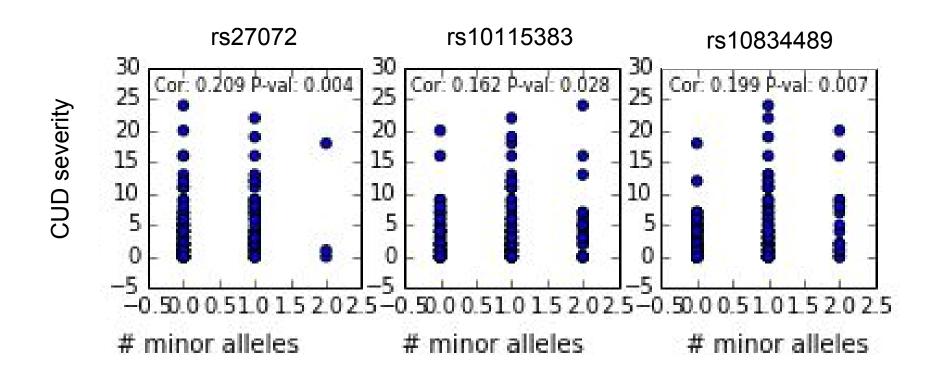
- PC1 General variability in sample (nothing interesting)
- PC2 Maps directly onto CUD severity



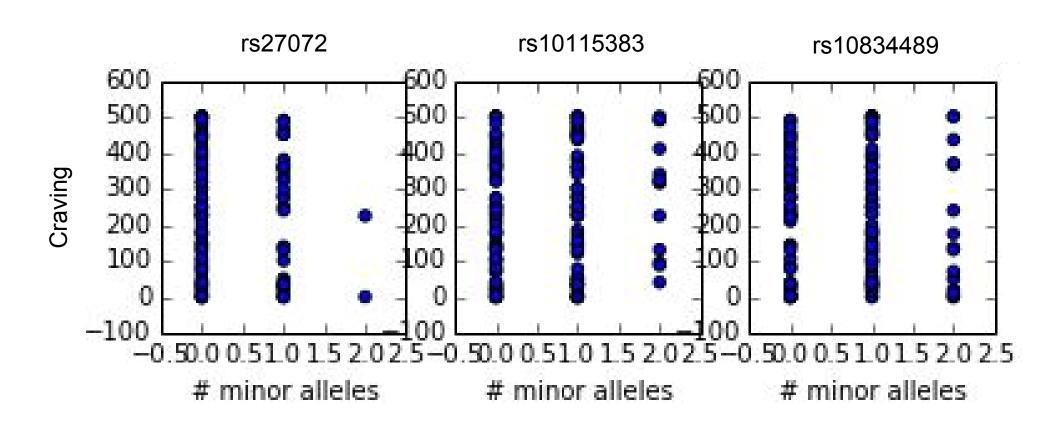
## Preliminary Results: variable selection



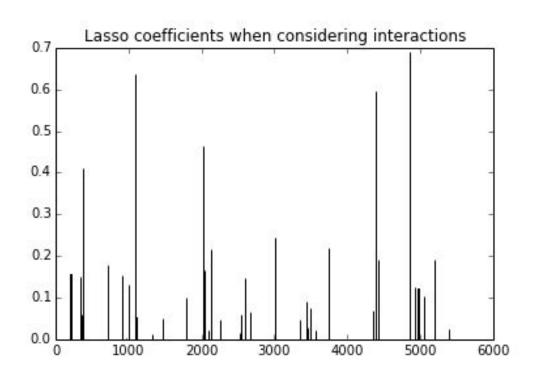
## Preliminary Results: CUD severity

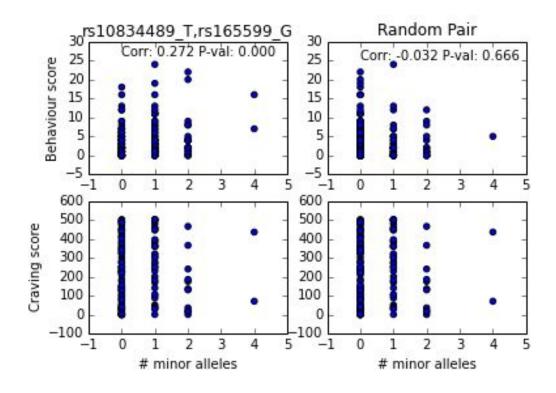


## Preliminary Results: subjective craving



## Preliminary Results - Learning Interactions





## Discussion

#### rs27072

- In the dopamine transporter, which takes dopamine from the synapse after it's been released
- Risk allele (T) is the major allele
- Associated with
  - alcohol withdrawal
  - ADHD in children
  - Genotype by age of onset interaction in tobacco and alcohol use in teenagers

#### rs165599

- In the Catechol *O*-Methyltransferase (COMT) gene
- COMT degrades dopamine
- G is both the risk allele and the minor allele
- Associated with
  - anxiety (Stein et al., 2005; Wray et al., 2008; Hettema et al., 2008)
  - impulsivity

#### Conclusions

- SNPs related to psychological constructs that are strongly associated with addiction
  - Negative emotions
  - Impulsivity and attention deficits
- Unknown what degrees different SNPS contribute to severity
- Interactions between SNPS also undetermined
- Challenging to study due to requirement of large sample sizes
- These analyses will
  - Identify individuals at risk for CUD
  - Provide potential solution for smaller studies

#### Future directions

- Determine the most appropriate method of adjusting for multiple comparisons
- Explore how changing parameters in the model changes weights of SNPs and their clinical relevance

■ Include SNPs that are susceptible to environmental changes

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