



# Using imaging genetics approach to development of potential biomarkers for psychological resilience

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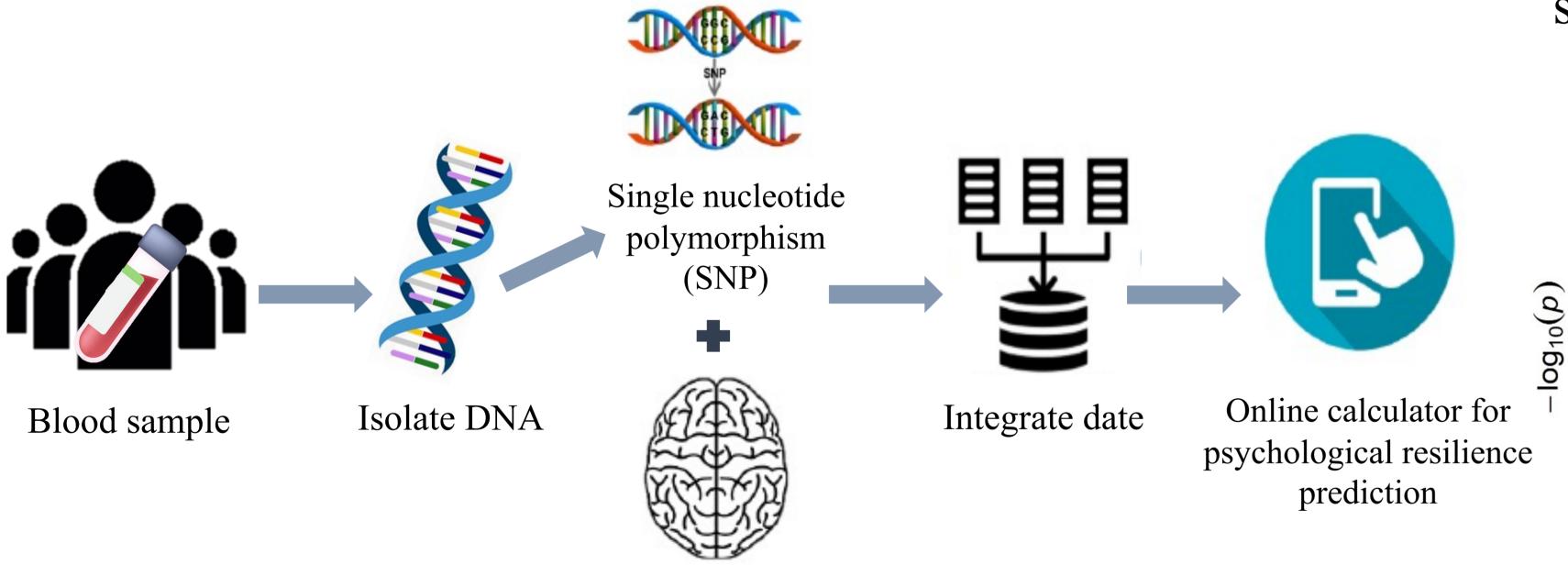
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# Background

Resilience is defined as the process of effectively negotiating, successful emotional, social adaptation or adjustment despite exposure to significant adversity, stress, or trauma. Our study combined neuroimaging (structural magnetic resonance imaging, sMRI) and genetic variation (single nucleotide polymorphism, SNP), used more effective statistical approaches to identify high-risk cases and provide more comprehensive recommendations for potential predictive factors of personal mental health.



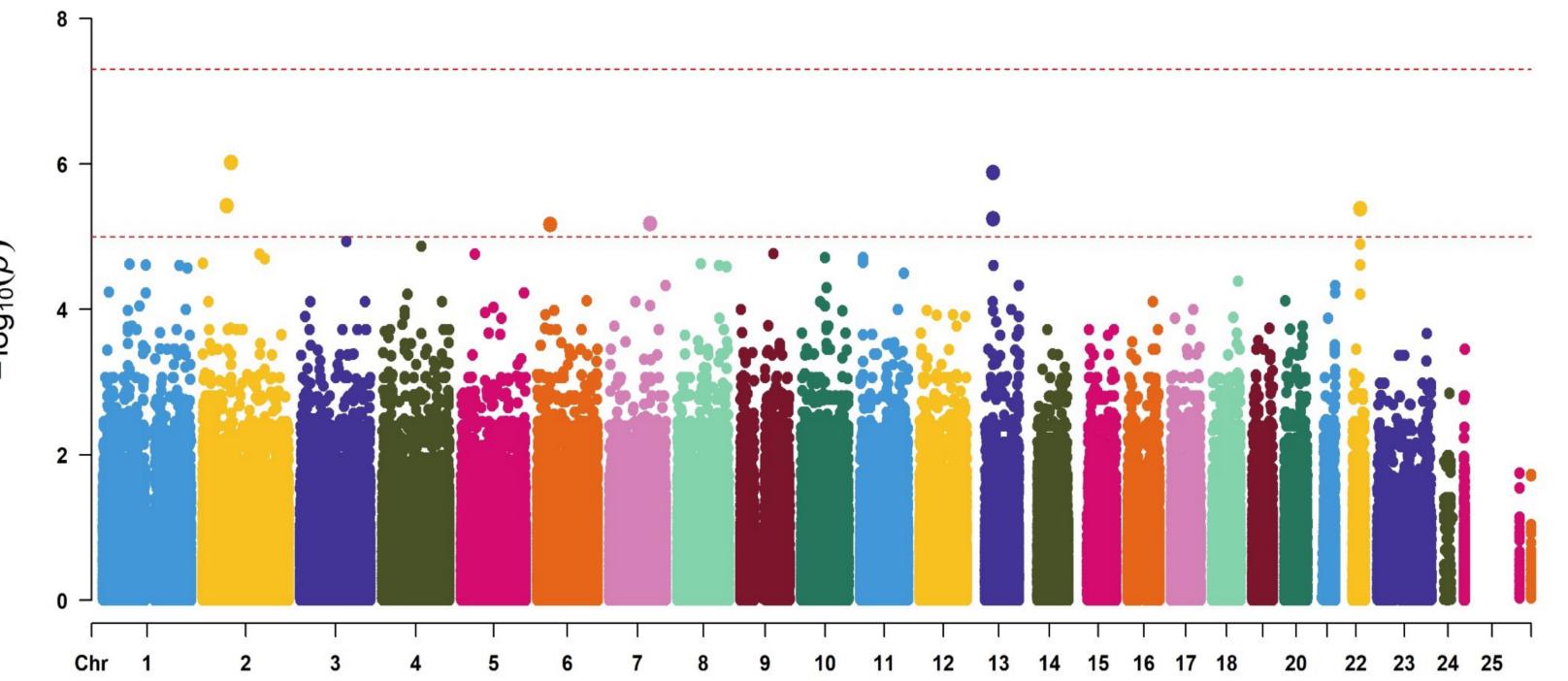
### Methods

Brian image

We recruited 94 young adults and divided them into two groups (high resilience and low resilience) by Connor-Davidson Resilience Scale (CD-RISC). All participants were conducted genome-wide association studies (GWAS) from blood DNA and scanned by magnetic resonance imaging. Then we integrated single nucleotide polymorphism (SNP) and structural MRI (sMRI) data by the canonical correlation analysis (CCA) and used ROC curve analysis for discriminating psychological resilience.

#### Results

Individuals were separated into high ( $\geq$ 60) and low (<60) resilience groups by the scores of the CD-RISC. Our study selected 9 SNPs that correspond p-value < 10<sup>-5</sup> in the genome-wide association analysis using PLINK. We excluded the three SNPs due to strong multicollinearity, and thus the 6 SNPs were analyzed in CCA. The proportion of explained variance is 54.26 (p-value = 0.016) based on the first pair of canonical variables of SNP and sMRI data using CCA. Then the result of ROC curve analysis with 10-fold cross validation showed the AUC and accuracy values of 0.92 and 0.90, respectively.



Summary results	Canonical correlation analysis			
	1 (first pair)	2	3	4
Eigenvalue	1.069	0.450	0.232	0.118
Proportion of explained variance	<b>54.26</b>	22.82	11.76	5.98
Canonical correlation	0.72	0.56	0.43	0.32
F value	1.44	0.93	0.66	0.48
p-value	0.016	0.624	0.951	0.990

Total Points: 58 Risk of Low Resilience: 43.2%						
Gender		Male	Female			
SNP						
rs17017697	TT	ТС	СС			
rs9566600	GG	AG	AA			
rs75035711	GG	AG	AA			
rs3218253	GG	AG	AA			
rs2520583	СС	TC	TT			
rs78927014	GG	AG	AA			
MRI						
Caudalanteriorcingulate (rh)		3250				
Fusiform (rh)		9042				
Lateraloccipital (rh)		7860				
Parstriangularis (rh)		5805				
Transversetemporal (rh)		1200				
Insula (rh)		7530				
Lateralorbitofrontal (lh)		8530				
Medialorbitofrontal (lh)		4270				
Paracentral (Ih)		4312				
Precuneus (Ih)		8640				
Rostralanteriorcingulate (lh)		2590				
Rostralmiddlefrontal (lh)		17520				
Supramarginal (lh)		11980				

# Conclusion

Case 1

Case 3

In our study, the results not only showed the selected SNPs related to the resilience traits, but also integrated the SNPs and neuroimaging data to distinguish high from low resilience subjects. In future, we will perform the external validation study in an independent sample based on the training set model of resilience.