

Epistatic Interactions for Brain eGWAS in Alzheimer's Disease

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Introduction

Complex genetic diseases, such as, Alzheimer's Disease (AD) are likely influenced by multiple genetic and environmental factors.

We hypothesize that some of these factors influence risk for disease through effects on gene expression.

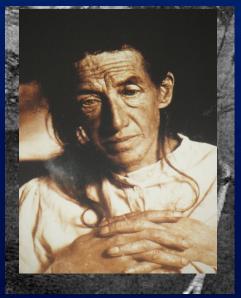
The goal of the current study is to identify pairs of genetic variants that influence brain gene expression levels: epistasis.

This is the first study to analyze brain gene expression data using an epistasis approach.

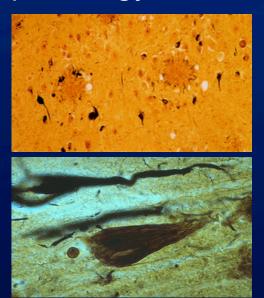


Background – Why this matters.

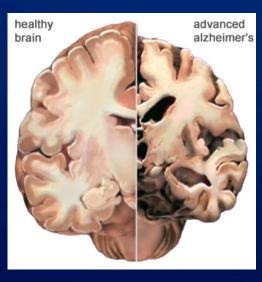
Alzheimer's Disease is the most common type of dementia with a specific neuropathology.



Memory, language and other cognitive problems NOT normal for age.



Abnormal accumulation of extracellular amyloid β (senile plaques) and intracellular tau (neurofibrillary tangles).



Atrophy of brain tissue

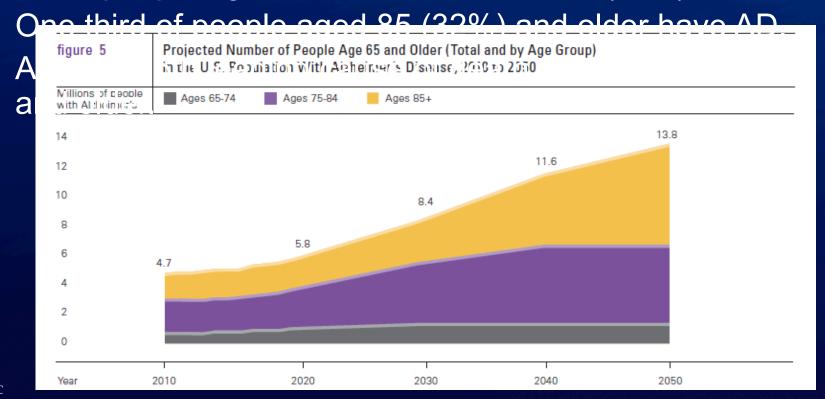


Incidence

Alzheimer's Disease is a Deadly Epidemic

5.2 million Americans have Alzheimer's disease (AD).

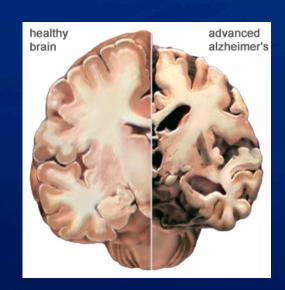
1 in 9 people aged 65 and older have AD (11%).





Treatments

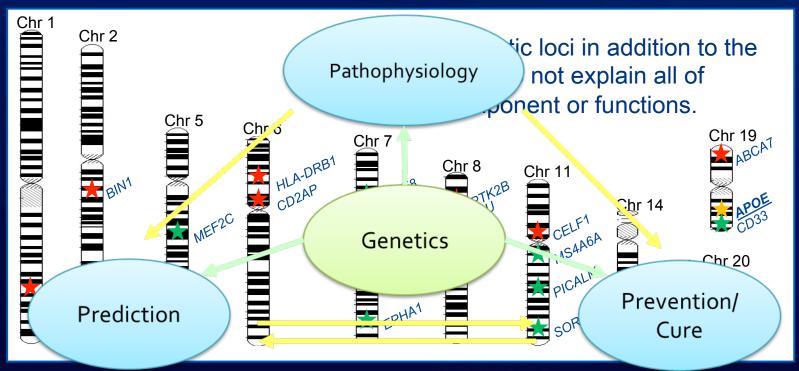
- Pharmacological:
 - Five FDA approved drugs are available that can provide temporary relief of the symptoms in some patients.
 - None of the current therapies for AD slows or stops the disease process.
- Non Pharmacological
 - Aimed at improving quality of life by managing symptoms.
 - Include, physical therapy, reminiscence therapy, cognitive stimulation.
 - Do not cure or prevent the disease.



Genetics

Mutations in 3 genes: *APP, PSEN1* and *PSEN2* known to cause rare early onset familial AD.

Up to 80% of risk for Late-Onset AD (LOAD) is predicted to be accounted for by genetics.





Approach

Can leverage endophenotypes, such as gene expression levels, to

- identify additional genetic factors.
- determine mechanism of action.

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PLOS GENETICS

Brain Expression Genome-Wide Association Study (eGWAS) Identifies Human Disease-Associated Variants

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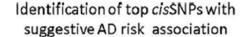
Neurology[®]

Novel late-onset Alzheimer disease loci variants associate with brain gene expression

Mariet Allen, Fanggeng Zou, High Seng Chai, et al.

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Imputed eGWAS

Significant cerebellar and temporal cortex cisSNP/transcript associations

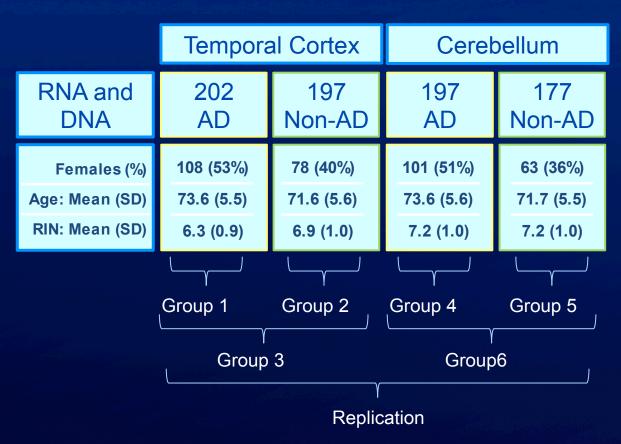
Cerebellar: 77,126 (63,652 cisSNPs, 2,338 genes) Temporal cortex: 68,172 (57,922 cisSNPs, 2,201 genes)

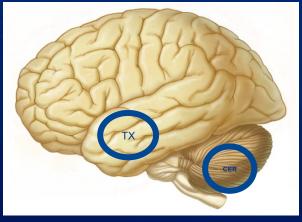
Significant eGWAS cisSNPs with suggestive AD risk associations in the ADGC GWAS.

Cerebellar + AD risk: 380 cisSNPs Temporal cortex + AD risk: 432 cisSNPs Cerebellar+Temporal cortex+AD risk: 356 cisSNPs



Approach: Samples





Frozen Brain Tissue from the Mayo Clinic Brain Bank (Dr. Dennis Dickson)

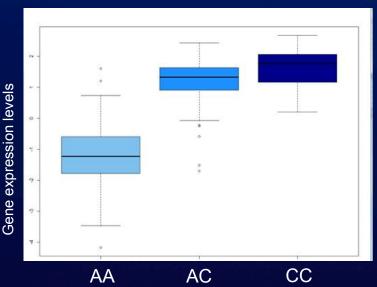


Approach – Genotypes and Phenotypes.

 Genotypes for >300,000 common SNPs distributed throughout the genome.
 For each subject 1 of 3 possible genotypes at each SNP is obtained:
 Carrasquillo et al, Nature Genetics 2009.

SNP A

Phenotypes are gene expression levels (mRNA) measured using Illumina Whole-Genome DASL: 24,526 probes (18,401 genes). Zou et al, PLoS Genetics 2012.

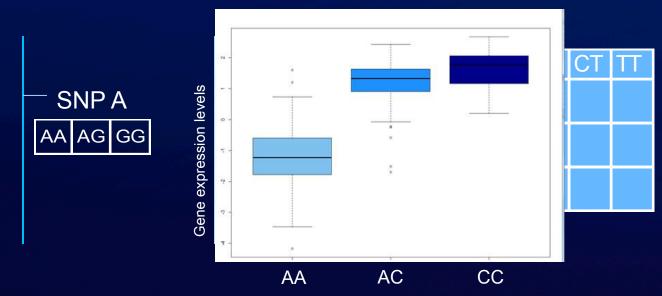




Why epistasis?

Single SNP/single phenotype approach is simplistic and cannot fully explain the known heritability of various diseases and phenotypes studied.

Epistasis allows for the study of interaction effects of pairs of SNPs on a given phenotype and can uncover additional genetic factors that influence gene expression and disease.





Key Challenges

Computation resources: Increase quadratically with the number of SNP interactions being considered. How do we to compute analysis of ~300,000 x ~300,000 epistatic interactions for 24,000 phenotypes?

Accounting for covariates: statistical applications that facilitate analysis of epistatic interactions do not allow for incorporation of covariates in regression analysis.

Storage: epistasis analysis of the scale described here generates large amounts of data which must be stored and organized.



Why BlueWaters?

Computation Resources:

Home Clusters

Software: PLINK

Phenotypes: 1 at a time

Estimated time:

75 hours/phenotype

BlueWaters

Software: Fast-Epistasis

Phenotypes: 32 at a time

Actual Time:

< 2 days for all phenotypes

Storage:

- Simultaneous FastEpistasis computation on increasing number of phenotypes quickly saturates the aggregate disk I/O on standard academic clusters.
- Intermediate files generated quickly add up to hundreds of terabytes per analysis.
- Easily handled by Blue Waters' petabyte storage facility.

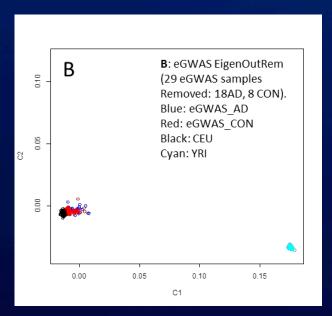


Data preparation workflow

3. Diploid, Common, High Quality: 273,848 SNPs

2. GWAS Post QC: 304,624 SNPs

1. GWAS genotypes: 318,257 SNPs



4. LD Prune: 223,632 SNPs

5. Remove population outliers.

6. Generate Residuals for Phenotypes (Covariates)

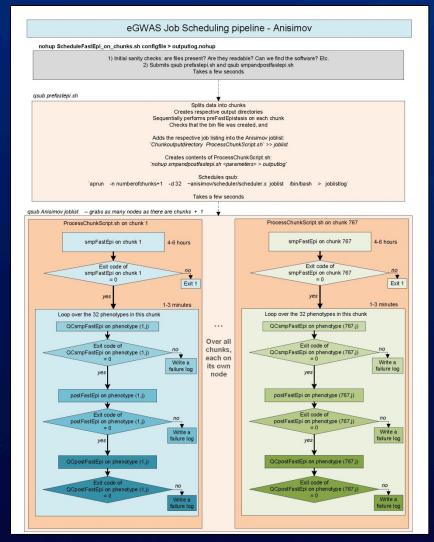


Accomplishments

NCSA senior scientist Liudmila Mainzer

- determined that Fast-Epistasis runs most optimally with 32 phenotypes at a time.
- designed code to launch Fast-Epistasis on BlueWaters.

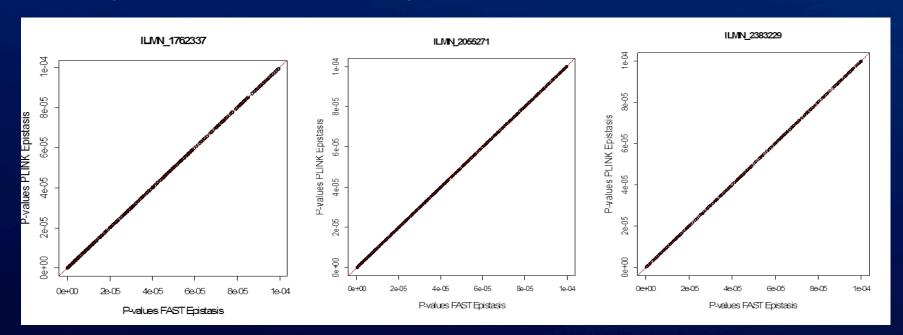
Fast Epistasis author
Thierry Schuepbach, is
collaborating with us to
make further improvements
to the application.



Accomplishments

PLINK and Fast-Epistasis give the same results

Have successfully completed the first of 6 analysis – Temporal Cortex AD samples





Future Directions

- Complete analysis of 2 additional groups of data Non-AD and AD+Non-AD.
- Completion of analysis of same 3 groups using expression data from Cerebellum.
- Filtering of results by counts for genotypes.
- Analysis of many hundreds of additional samples using gene expression data collected using RNAseq.







Acknowledgements

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