# Clinical Utility of Pharmacogenomic Findings: <br> Beyond Single Variants 

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## Successes and Challenges of Genome Studies

- GWAS/Sequencing
- 10K robustly associated genetic variants
- New insights into biology of many traits
- Biological understanding is still lacking


## Pharmacogenomic Findings

| Evidence <br> Level | Counts | $\%$ |
| :---: | :---: | :---: |
| 1 a | 40 | 3 |
| 1 b | 17 | 1 |
| 2 a | 96 | 6 |
| 2 b | 74 | 5 |
| 3 | 1175 | 76 |
| 4 | 145 | 9 |
| Total | 1547 | 100 |

https://www.pharmgkb.org/


## Genetic Architecture of Complex Traits



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## Single Variants Not Relevant for Highly Polygenic Traits



## Genes mirror geography within Europe

John Novembre ${ }^{1,2}$, Toby Johnson ${ }^{4,5,6}$, Katarzyna Bryc ${ }^{7}$, Zoltán Kutalik ${ }^{4,6}$, Adam R. Boyko ${ }^{7}$, Adam Auton ${ }^{7}$, Amit Indap ${ }^{7}$, Karen S. King ${ }^{8}$, Sven Bergmann ${ }^{4,6}$, Matthew R. Nelson ${ }^{8}$, Matthew Stephens ${ }^{2,3}$ \& Carlos D. Bustamante ${ }^{7}$


## Prediction and Dissection to Achieve Clinical Utility

## Prediction

## Dissection

- Etiology of complex traits
- Mechanism by which genetic variation drives phenotypic variation
- Druggable targets
- Disease
- risk stratification
- intervention strategies
- Adverse events
- Efficacy of treatment


## Projecting the performance of risk prediction based on

 polygenic analyses of genome-wide association studiesNilanjan Chatterjee ${ }^{1}$, Bill Wheeler ${ }^{2}$, Joshua Sampson ${ }^{1}$, Patricia Hartge ${ }^{1}$, Stephen J Chanock ${ }^{1} \&$ Ju-Hyun Park $^{1,3}$

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T2D


## 

## Whole Genome Prediction Approaches

## LETTERS

## Common polygenic variation contributes to risk of schizophrenia and bipolar disorder

The International Schizophrenia Consortium*

## REPORT

GCTA: A Tool for Genome-wide Complex Trait Analysis

Jian Yang, ${ }^{1, *}$ S. Hong Lee, ${ }^{1}$ Michael E. Goddard, ${ }^{2,3}$ and Peter M. Visscher ${ }^{1}$

## Whole Genome Prediction Approaches

## Research Article

## Genetic

Epidemiology

## Poly-Omic Prediction of Complex Traits: OmicKriging

Heather E. Wheeler, ${ }^{1}$ Keston Aquino-Michaels, ${ }^{2}$ Eric R. Gamazon, ${ }^{2}$ Vassily V. Trubetskoy, ${ }^{2}$ M. Eileen Dolan, ${ }^{1}$ R. Stephanie Huang, ${ }^{1}$ Nancy J. Cox, ${ }^{2}$ and Hae Kyung Im ${ }^{3 *}$

MultiBLUP: improved SNP-based prediction for complex traits
Doug Speed and David J Balding
Genome Res. published online June 24, 2014
Access the most recent version at doi:10.1101/gr.169375.113

## Polygenic Modeling with Bayesian Sparse Linear Mixed Models

Xiang Zhou ${ }^{1 *}$, Peter Carbonetto ${ }^{1}$, Matthew Stephens ${ }^{1,2 *}$

## Whole Genome Prediction Approaches

J. R. Statist. Soc. B (2005)

67, Part 2, pp. 301-320

## Regularization and variable selection via the elastic net

Hui Zou and Trevor Hastie

## SparSNP: Fast and memory-efficient analysis of all SNPs for phenotype prediction

Gad Abraham ${ }^{1 *}$, Adam Kowalczyk ${ }^{1}$, Justin Zobel ${ }^{1}$ and Michael Inouye ${ }^{2,3}$

## GWAS hits vs. Whole Genome Prediction (OmicKriging)



## Collective Approaches to Dissect Complex Traits

- Enrichment of functional classes
- Partitioning heritability into functional classes
- Aggregation into functional units


## Trait-Associated SNPs Are More Likely to Be eQTLs: Annotation to Enhance Discovery from GWAS



## Cross-tissue and tissue-specific eQTLs: locating the missing

 heritability of a complex trait across populationsJason M. Torres, Eric R. Gamazon, Esteban J. Parra, Jennifer E. Below, Adan Valladares-Salgado, Niels Wacher, Miguel Cruz, Craig L. Hanis, Nancy J. Cox*


## Regulatory variants explain much more heritability than coding

 variants across 11 common diseasesAlexander Gusev, S Hong Lee, Benjamin M Neale, et al.


DHS: DNAse hypersensitivity sites, control accessibility of the region thus levels of transcription

## Gene Based Tests

- Gene based association tests
- VEGAS (Liu et al 2010 AJHG)
- SKAT (Wu et al 2012 AJHG)
- C-Alpha (Neale and Rivas et al 2011 Plos Genetics)
- Used extensively in whole exome studies
- Designed to address low power of rare variants


## Gene Based Tests

- Limited success of gene based tests
- More functional data needs to be integrated
- Enrichment studies indicate important role of gene regulation
- To address this issues, we propose PrediXcan
- predict expression levels of a gene
- correlate predicted levels with complex traits
- scan whole genome


## Genetic Control of Disease Through Gene Regulation



## PrediXcan Flow

## Start with genetic data

No transcriptome data is needed

Predict whole genome effect on expression level of a gene

Correlate predicted expression with phenotype

Only genotype and phenotype data needed

No reverse causality

Replicate Genes with Independent

Training sets Test sets

Validate Genes in Model Systems

Mechanism is built in

## Additive Model for Genetic Effect Prediction

## Predicted Expression Trait

$$
t_{i}=\sum_{k=1}^{M} w_{k} G_{k i}
$$

$t_{i}$ is predicted effect on gene expression level for individual $i$ $G_{k i}$ number of reference alleles for SNP $k$ and individual $i$ $w_{k}$ weight for SNP $k$

## Simple Polygenic Model

- $w_{k}=$ single variant regression coefficient (Matrix eQTL output)
- $w_{k}$ set to zero if $p$ value $>0.05$ for cis SNPs ( 1 Mb TSS)
- $w_{k}$ set to zero if p value $>10^{-6}$ for trans SNPs


## Expression Data

- GTEx - Genotype of Tissue Expression
- Large scale Common Fund project
- 900 organ donors
- 45 tissues
- RNAseq, whole exome seq, whole genome seq
- gEUVADIS
- RNAseq 462 individuals from the 1000 Genomes Project
- Cerebellum expression (Array GSE35974)


## Good Prediction Performance

## Prediction R^2



## Training with GTEx Testing in 1K Genomes

## Replication R^2



Replicate RNAseq Pickrell et al 2010 vs. 1K Genomes 2013

Sahar Mozaffari

## Examples of Well Predicted Genes



## Genes Associated with Rheumatoid Arthritis



## PrediXcan Results for Crohn's Disease and Hypertension



## PrediXcan Outperforms VEGAS



## Enrichment of Known Crohn's Genes Among Findings

## 100 qqplot with

 random samples of 205 genes

## No Enrichment Among Hypertension Findings

100 qqplots with random samples of 133 genes


## Whole blood may not be

 relevant tissuerandinm nonec -lncin HT nual

## Bipolar Disorder WTCCC results



## Significant Concordance Between Independent Bipolar Studies

## Higher correlation for cerebellum based predictions than whole blood based



## PrediXcan: a Gene Discovery Approach

- PrediXcan is a powerful gene based association test
- It directly tests the molecular mechanism through which genetic variants affect phenotype
- Reduced multiple testing burden compared to single variant approach
- Unlike other gene based tests, it provides direction of effects
- Advantages relative to gene expression studies
- Applicable to any GWAS datasets gene expression levels are predicted from genotype data
- No reverse causality disease status does not affect germline DNA
- Multiple Tissues can be evaluated tissue expressions are only needed to build prediction models


## Challenges of Pharmacogenomic Studies

- Smaller sample size
- Even more important to integrate prior data
- Integrate other functional data
- Heritability estimates are harder
- Limited family data
- Usually samples greater than 1K are needed for GCTA


## Bevacizumab Induced Hypertension

- Bevacizumab is a humanized monoclonal antibody that inhibits VEGF induced angiogenesis
- Hypertension is a common adverse event to bevacizumab treatment
- The incidence of hypertension with bevacizumab is $20-30 \%$, while grade 3 or greater hypertension occurs in only 10-15\% of patients.


## Bevacizumab Trials

- CALGB 90401
- a randomized double-blinded placebo controlled phase III trial comparing docetaxel and prednisone with and without bevacizumab in men with hormone refractory prostate cancer
- $\mathrm{n}=664$ (with genotype data after QC)
- PI: Howard McLeod
- CALGB 80303
- a randomized phase III trial of gemcitabine plus bevacizumab versus gemcitabine plus placebo in patients with advanced pancreatic cancer
- $\mathrm{n}=152$ (with genotype data after QC)
- PI: Federico Innocenti


## Bevacizumab Induced Hypertension

- Is primary hypertension risk score predictive of bevacizumab induced hypertension
- Hypertension results from Cross Consortia Pleiotropy group ( $\mathrm{n} \sim 20 \mathrm{~K}$ )
- Can we predict drug induced hypertension?
- 90401 training set
- 80303 test set
- Dissection of Hypertension

Keston Aquino Michaels \& Heather Wheeler

## Primary Hypertension score Predicts Bev-induced HT



## Bev-Hypertension Predicted Within Study



## Bev-Hypertension Predicted in Independent Study



Keston Aquino Michaels

## PrediXcan Results Primary and Bev-Hypertension



## Bevacizumab Hypertension



Heather Wheeler

## PrediXcan Results Primary and Bev-Hypertension



## Bevacizumab Hypertension



Heather Wheeler

## Hypertension and ICAM1

- Genetically predicted expression levels of ICAM1 was associated with
- Primary hypertension WTCCC
- Bevacizumab induced hypertension CALGB 90401
- Genetically predicted serum levels of ICAM1 was associated with
- Bev induced hypertension CALGB 90401
- Increased levels of ICAM1 were associated with blood pressure in induced hypertension mice model


## Summary

- Most single variant findings have limited clinical utility
- Whole genome approaches to prediction improves utility
- Aggregation, partitioning and enrichment studies improves dissection
- Bevacizumab induced hypertension example
- primary hypertension results help in predicting drug induced hypertension
- succesfully predicted bevacizumab induced hypertension in indenpendet study
- PrediXcan: novel gene based test that test mechanism yielded promising findings
- Need more samples and better methods


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## Contributors

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- CALGB 90401 Howard McLeod
- CALGB 40101 Deanna Kroetz
- GTEx Consortium


## Data sources

- Cross Consortia Pleiotropy XCP summary results
- WTCCC


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## Lipid Markers AUC

Manickam et al 2011 J Clinical Lipidology


