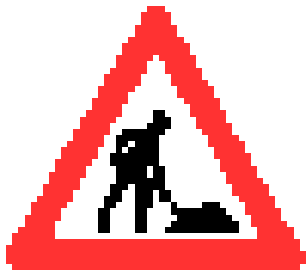




A simple method for detecting heterogeneity in meta-analyses of GWAS studies



David Evans
University of Bristol



welcometrust

learning
discovery
enterprise

MRC

Centre for Causal
Analyses in Translational
Epidemiology

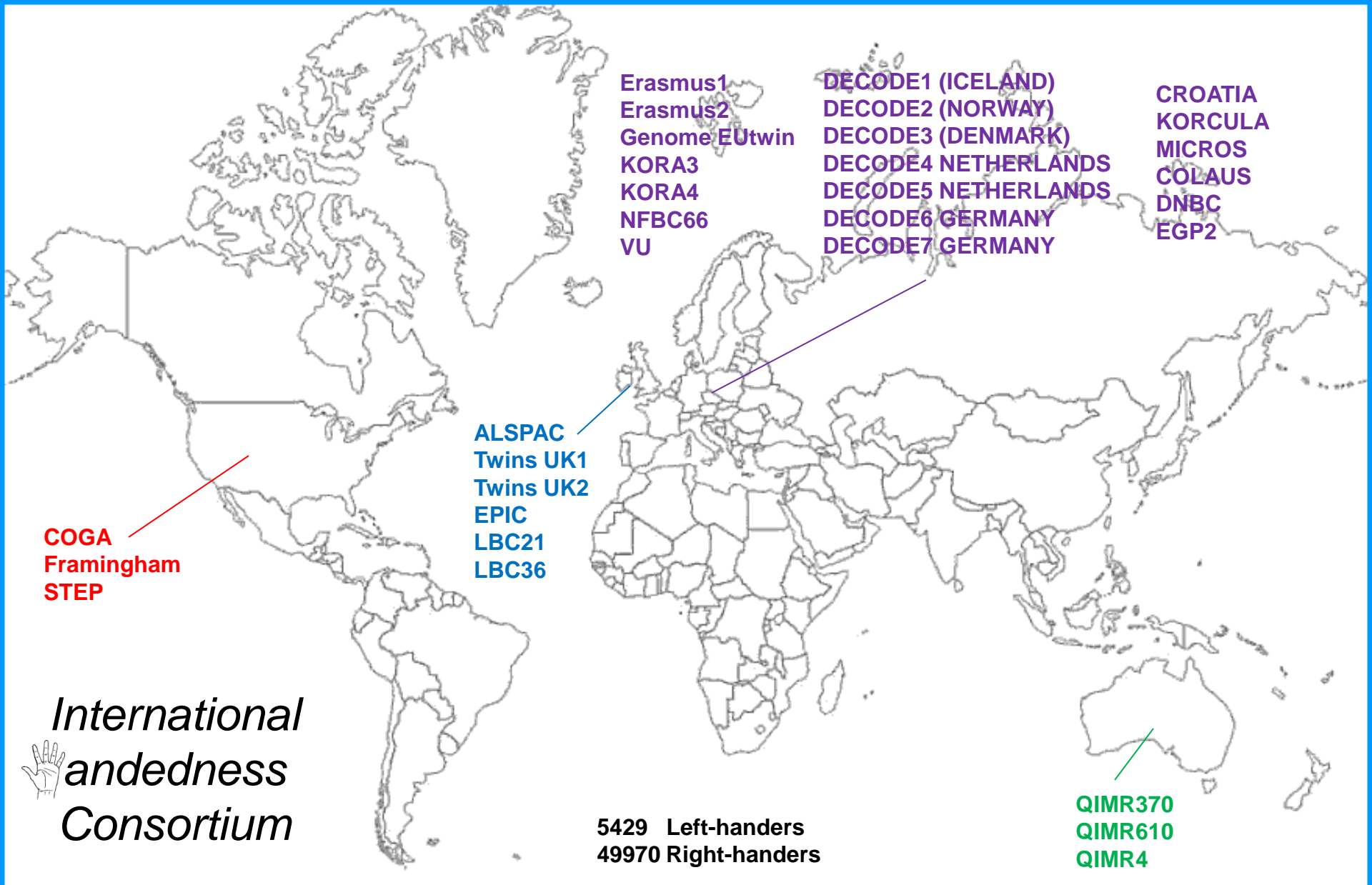


Handedness

- First evidence at 9-10 weeks gestation
- Best predictor of cerebral lateralization
- “Which hand do you write with...”?
- Prevalence ~10%
 - Increasing since turn of century
 - Mirror-imaging?
- Heritability ~25% (Medland et al. 2009)
 - Single gene?
 - No evidence of C
 - Environmental effects include birth weight

welcometrust





*International
andedness
Consortium*



Heterogeneity in Meta-analyses of GWAS

- GWAS studies can differ not only because of phenotypic and genetic heterogeneity, but also error
- Traditionally SNPs examined on a locus by locus basis
 - Tests have low power
 - Discards information present genome-wide
- Particularly relevant to handedness phenotype
 - Possibility of reverse coding
 - Particularly refractive to finding genes
 - No locus of strong effect to act as a positive control
 - Relevant to e.g. eczema etc

welcometrust



Polygenic Theory of Complex Traits

- Complex traits due to the action of hundreds of loci of small effect across the genome
- The results from GWA studies of the same (or similar) phenotypes should be more similar than GWA studies of etiologically different traits / GWA studies where there is added heterogeneity or systematic error
- Is it possible to harness this information to help detect heterogeneity between genome-wide association scans?

welcometrust



Harnessing the information contained within genome-wide association studies to improve individual prediction of complex disease risk

David M. Evans^{1,*}, Peter M. Visscher² and Naomi R. Wray²

¹Department of Social Medicine, MRC Centre for Causal Analyses in Translational Epidemiology, University of Bristol, Bristol, UK and ²Genetic Epidemiology and Queensland Statistical Genetics, Queensland Institute of Medical Research, Australia

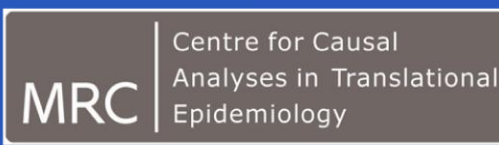
Table 1. Median AUC values for the seven diseases using the count and log odds methods

Threshold	BD	CHD	HT	CD	RA	T1D	T2D
Count method							
0.8	0.653 (0.527)	0.599 (0.527)	0.602 (0.538)	0.617 (0.554)	0.591 (0.522)	0.620 (0.513)	0.589 (0.520)
0.5	0.664 (0.527)	0.598 (0.533)	0.600 (0.534)	0.622 (0.553)	0.594 (0.528)	0.624 (0.515)	0.593 (0.513)
0.1	0.646 (0.537)	0.570 (0.532)	0.587 (0.534)	0.596 (0.524)	0.592 (0.515)	0.637 (0.515)	0.578 (0.516)
0.05	0.625 (0.537)	0.552 (0.525)	0.589 (0.532)	0.591 (0.521)	0.599 (0.524)	0.673 (0.537)	0.576 (0.529)
0.01	0.570 (0.555)	0.588 (0.508)	0.566 (0.518)	0.561 (0.514)	0.625 (0.522)	0.697 (0.531)	0.556 (0.516)
0.001	0.539 (0.534)	0.590 (0.534)	0.570 (0.521)	0.581 (0.532)	0.645 (0.546)	0.712 (0.544)	0.567 (0.549)
0.0001	0.533 (0.518)	0.551 (0.542)	0.568 (0.526)	0.624 (0.542)	0.647 (0.540)	0.716 (0.540)	0.565 (0.543)
0.00001	0.521 (0.525)	0.553 (0.509)	0.526 (0.536)	0.607 (0.558)	0.642 (0.528)	0.717 (0.540)	0.545 (0.515)
Log odds method							
0.8	0.668 (0.529)	0.595 (0.534)	0.610 (0.530)	0.614 (0.541)	0.646 (0.530)	0.721 (0.531)	0.601 (0.518)
0.5	0.668 (0.531)	0.592 (0.531)	0.610 (0.525)	0.618 (0.536)	0.642 (0.534)	0.724 (0.518)	0.601 (0.513)
0.1	0.636 (0.547)	0.580 (0.534)	0.599 (0.523)	0.598 (0.519)	0.652 (0.522)	0.743 (0.515)	0.574 (0.522)
0.05	0.620 (0.537)	0.560 (0.527)	0.596 (0.524)	0.592 (0.521)	0.656 (0.526)	0.747 (0.526)	0.568 (0.523)
0.01	0.567 (0.548)	0.600 (0.509)	0.585 (0.517)	0.574 (0.522)	0.666 (0.530)	0.749 (0.525)	0.569 (0.529)
0.001	0.533 (0.527)	0.590 (0.528)	0.580 (0.519)	0.597 (0.535)	0.661 (0.547)	0.749 (0.545)	0.575 (0.558)
0.0001	0.528 (0.520)	0.545 (0.544)	0.571 (0.534)	0.627 (0.544)	0.658 (0.557)	0.748 (0.534)	0.569 (0.549)
0.00001	0.529 (0.521)	0.556 (0.522)	0.520 (0.531)	0.612 (0.555)	0.655 (0.533)	0.749 (0.533)	0.544 (0.526)

welcometrust



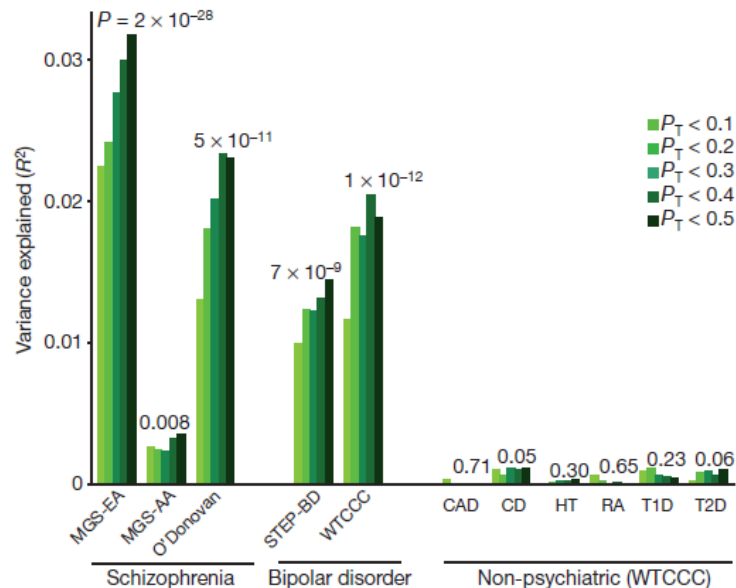
learning
discovery
enterprise



LETTERS

Common polygenic variation contributes to risk of schizophrenia and bipolar disorder

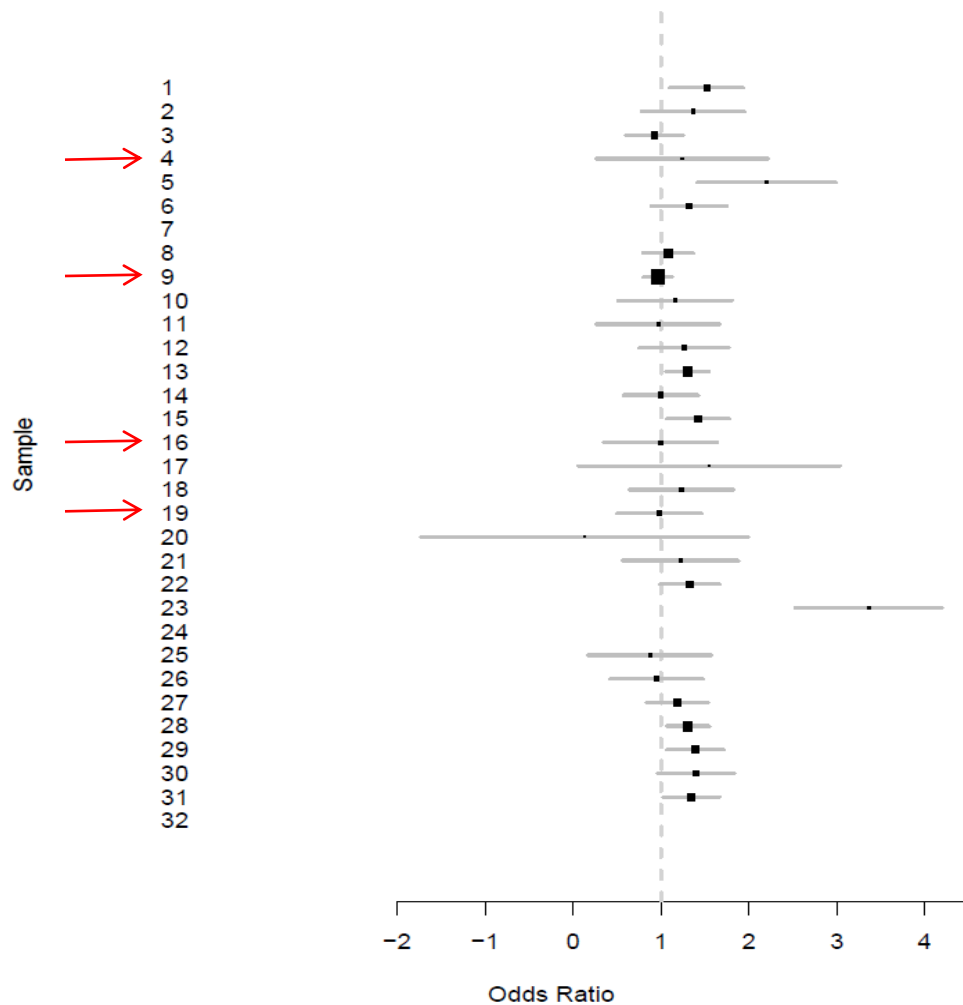
The International Schizophrenia Consortium*



ISC (2009) Nature

welcometrust

Which cohorts show heterogeneity?



$p = 2.7 \times 10^{-6}$

METHOD



Start with $N_{\text{Cohort}} \times N_{\text{SNP}}$ matrix of p values



Transform each element to $-\log_{10} p_{\text{val}} \times \text{Direction of effect}$

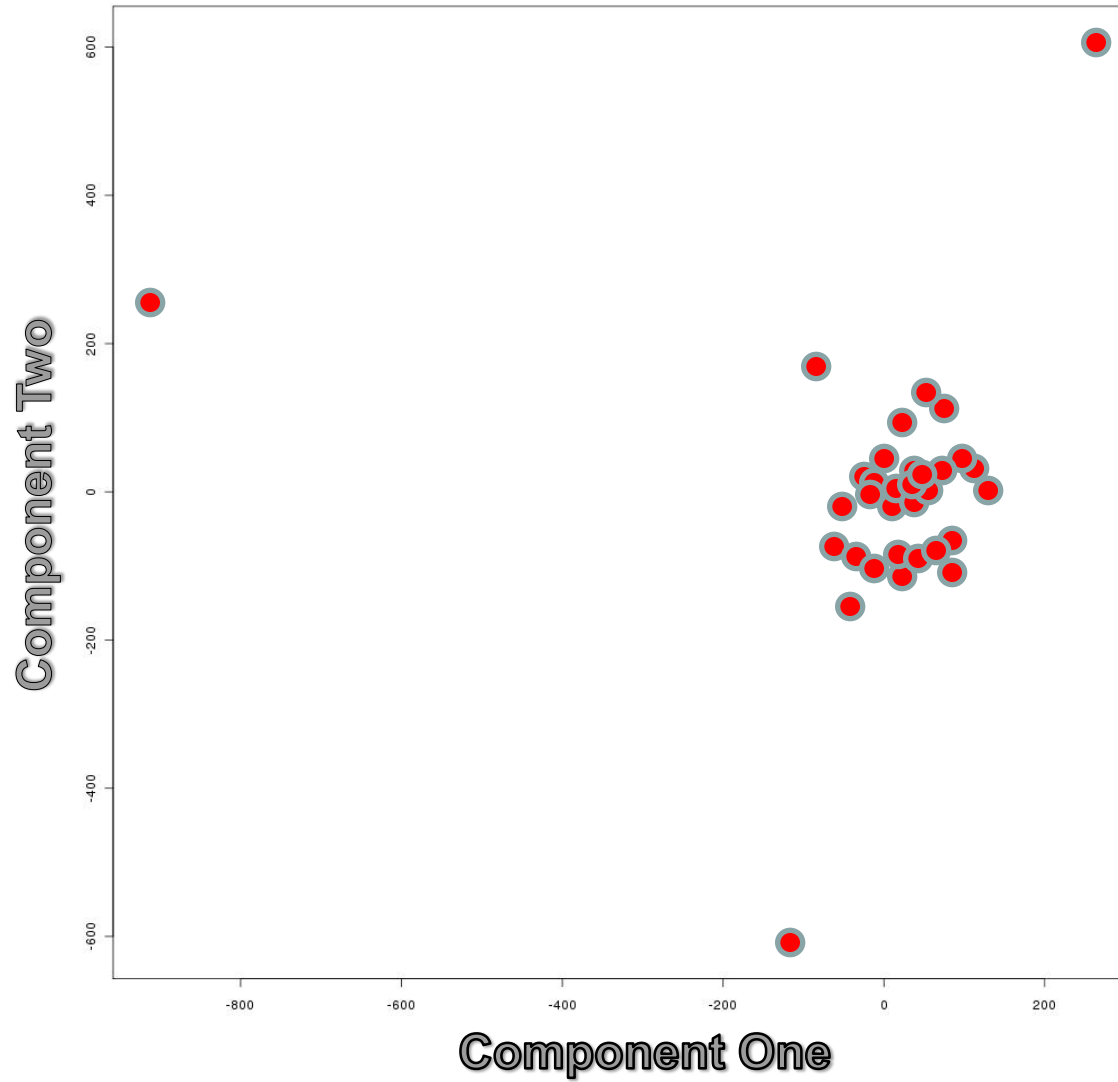


Generate $N_{\text{Cohort}} \times N_{\text{Cohort}}$ matrix
of Euclidean distances



Multi-dimensional scaling

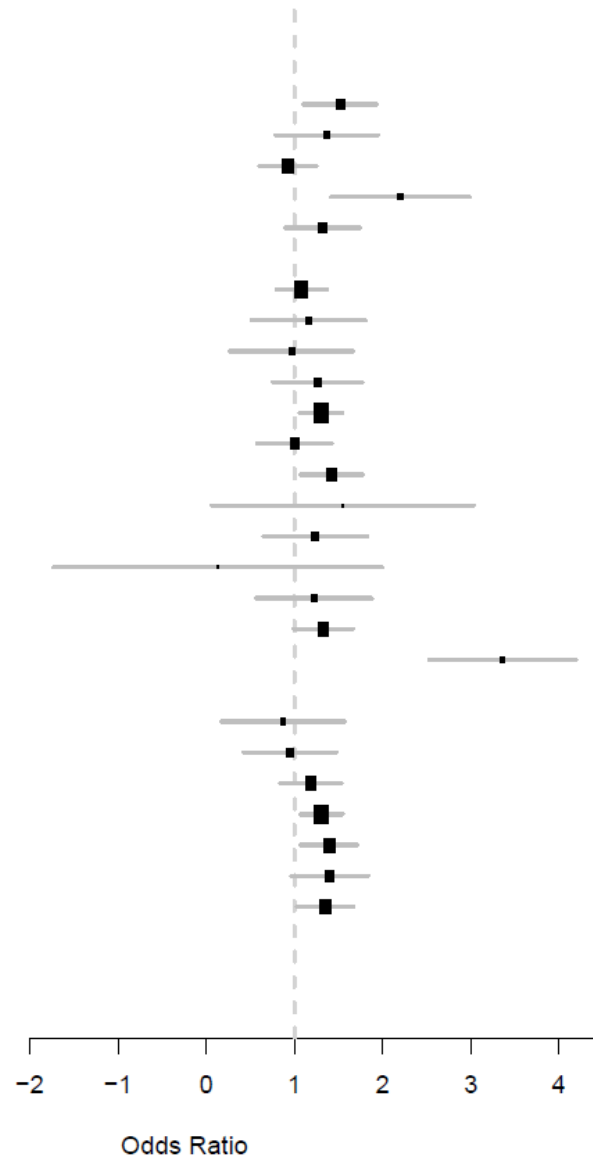
Metric MDS





Sample

1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
22
23
24
25
26
27
28



$p = 6.1 \times 10^{-8}$



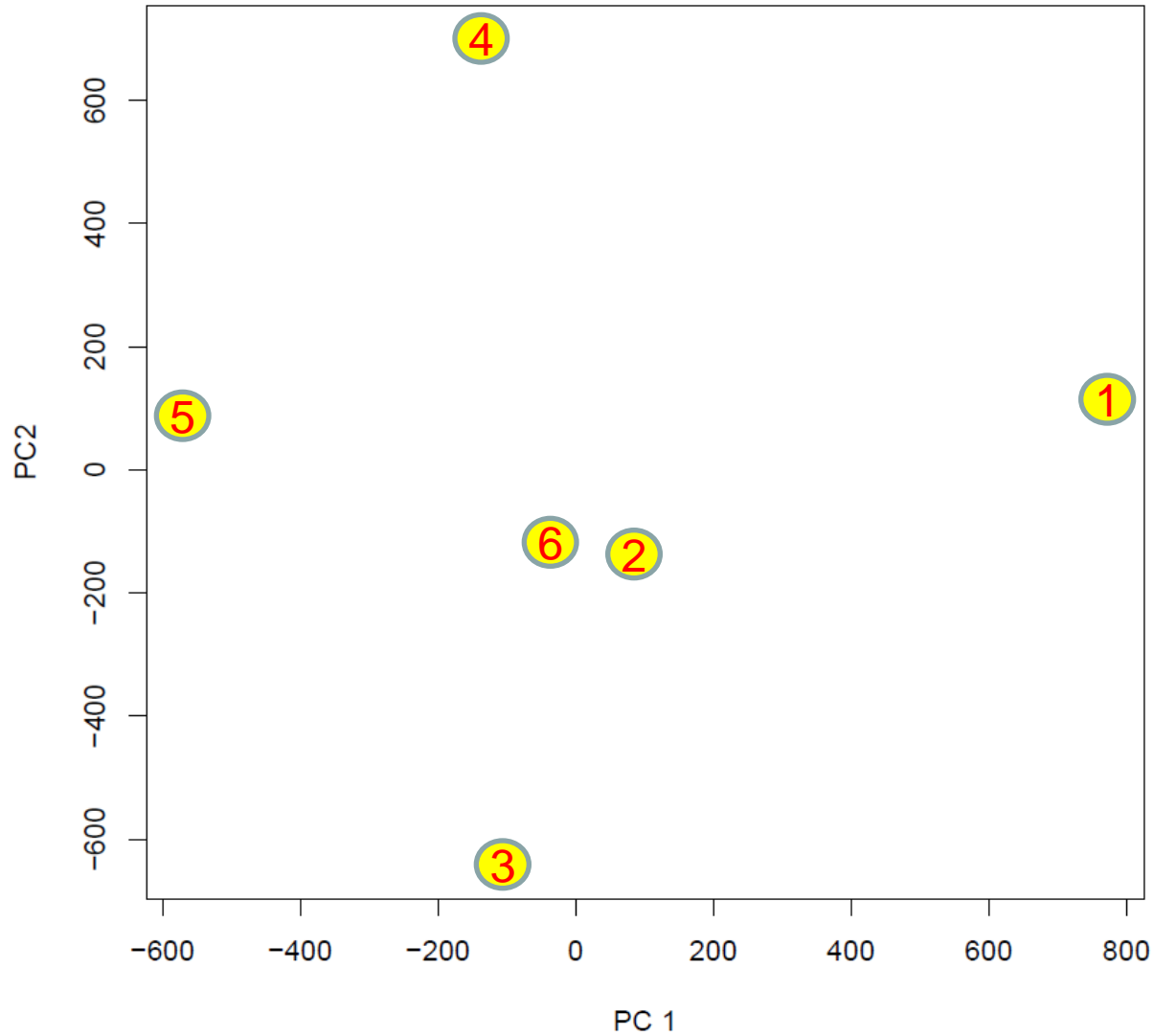
Simulations

- 3000 individuals generated by HAPGEN
- 100 QTLs totalling 75% heritability randomly distributed across genome
- 6 x 500 individuals
- Dataset “6” has reverse coding

welcometrust



All Loci



METHOD



Start with $N_{\text{Cohort}} \times N_{\text{SNP}}$ matrix of p values



Similarity measure is n loci with $p < \text{threshold}$ in both cohorts in same direction divided by n loci with $p < \text{threshold}$ in both cohorts

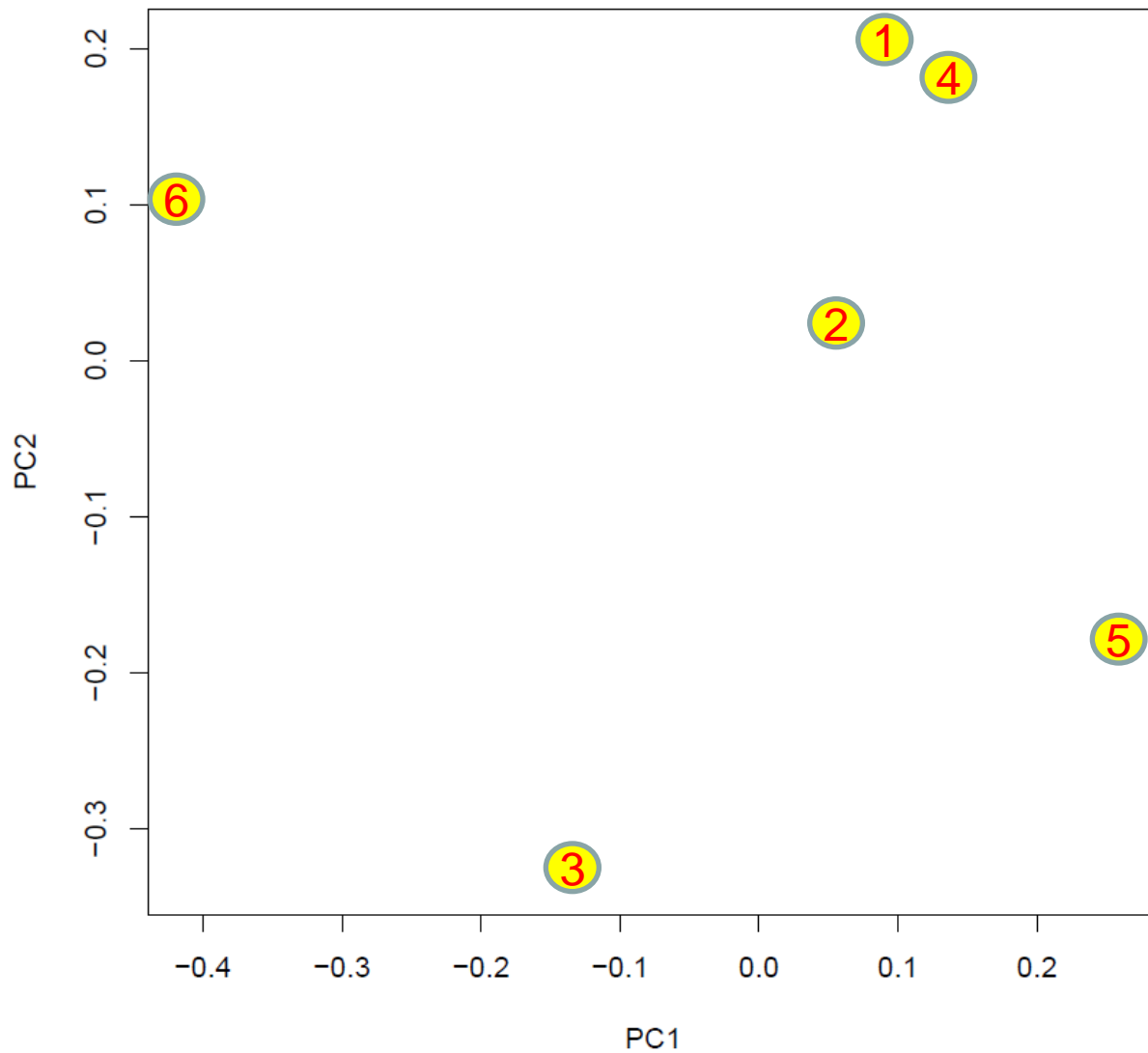


Generate $N_{\text{Cohort}} \times N_{\text{Cohort}}$ matrix of Euclidean distances

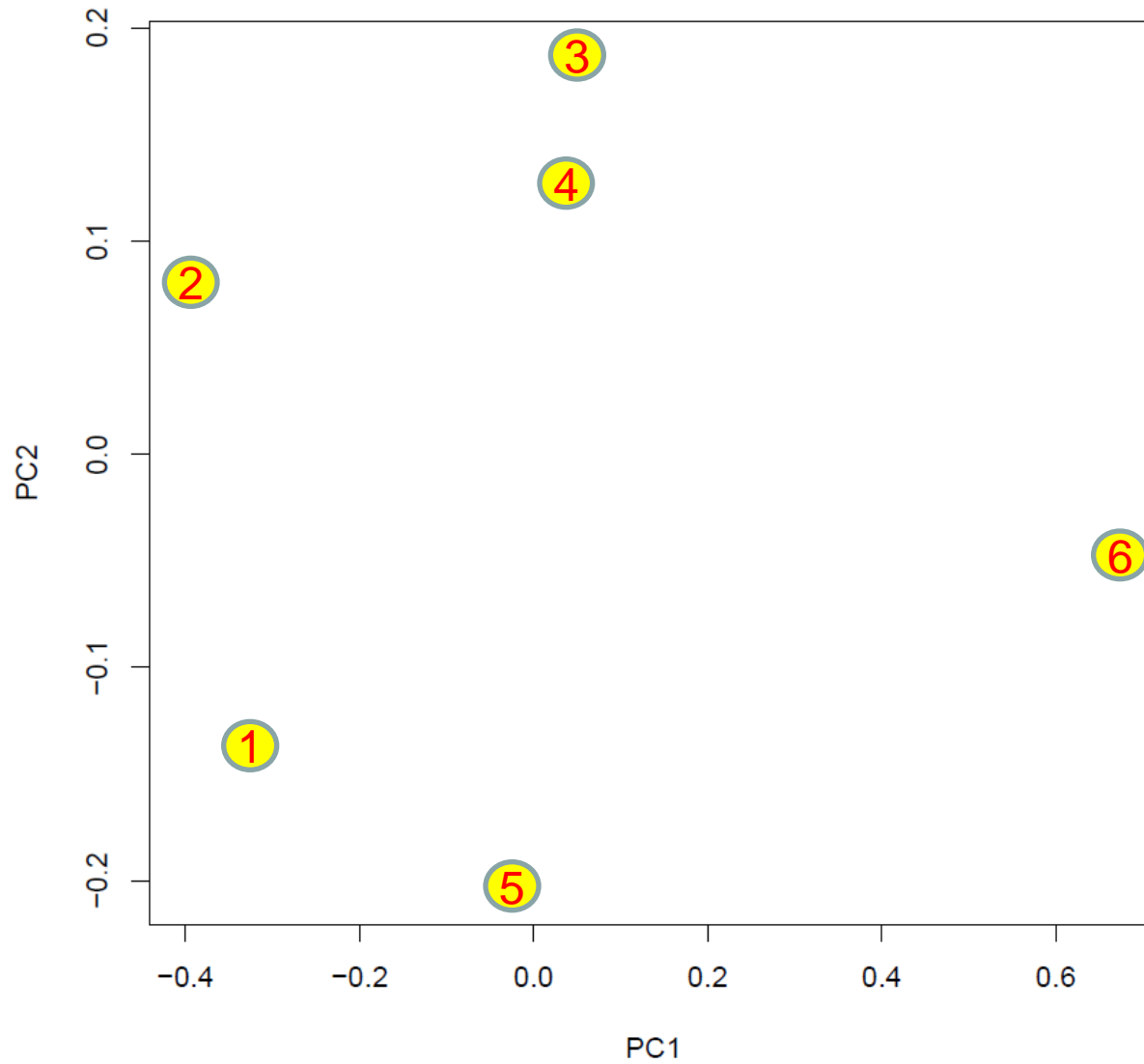


Multi-dimensional scaling

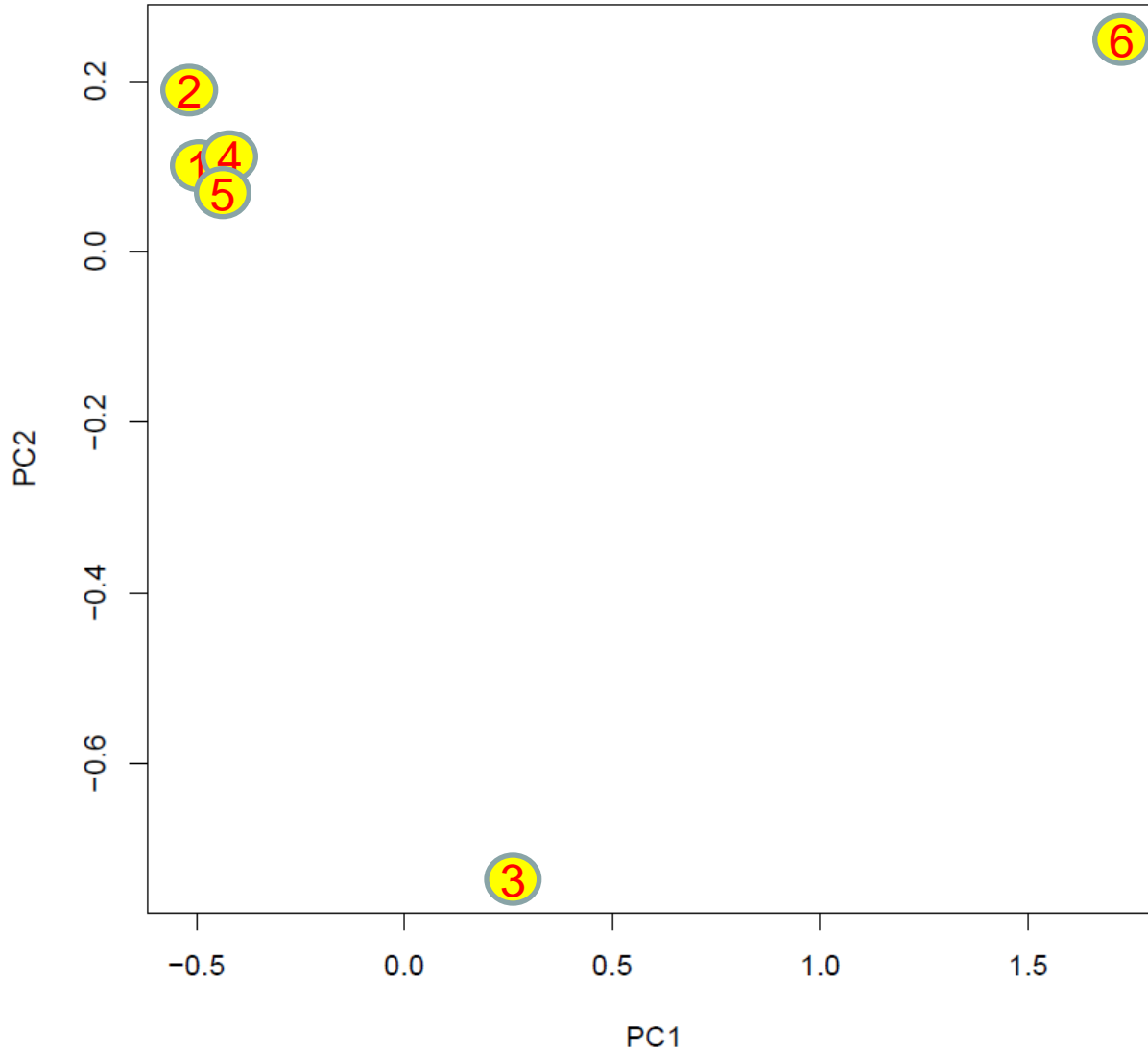
Shared Loci $P < 0.1$



Shared Loci $P < 0.01$



Shared Loci $P < 0.001$



Conclusions So Far?

- Clustering using number of shared “significant” loci may be better than using the whole genome (noisy)
- Stricter thresholds for selecting loci do better in terms of clustering properties
- Worryingly, cohorts without heterogeneity built into their model cluster strangely
- Strange clustering in Handedness Consortium may not be due to heterogeneity?

welcometrust



To Do

- Refine method
- Apply to simulated data with other sorts of heterogeneity (stratification etc), different genetic models, random and systematic genotyping error, sample sizes etc.
- Reapply to real data

welcometrust



Acknowledgments

- Sarah Medland Queensland Institute of Medical Research
- International Handedness Consortium

welcometrust

