# The design and specification of biobanks

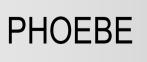
#### **Paul Burton**

Professor of Genetic Epidemiology University of Leicester



P<sup>3</sup>G Consortium

PROMOTING HARMONISATION OF







## Thanks!

- Anna Hansell and Paul Elliott
  - Imperial College
- Isabel Fortier
  - P<sup>3</sup>G



#### Structure of talk

- What determines the size and shape of a genetic epidemiology study?
- The statistical power of case-control studies
- Expected event rates in large cohort studies
- International biobank harmonization



# What determines "size" and "shape"?



#### What determines "shape"?

- The scientific question
  - Unrelated individuals v families
    - Association v linkage
    - CDCV v rare alleles with large effects
  - Case-control v cohort designs
  - Classical public health research
  - Special populations
- Pragmatic opportunities and challenges
  - Record linkage
  - Special approaches to recruitment
- Ethico-legal considerations



**P** 

#### What determines "size"?

- The scientific question
  - Statistical power
  - Type of end-point
  - Main effects v interactions
  - Time required to generate enough cases
- Cost and resources
- Pragmatic restrictions







#### Two classes of cohort biobanks

- Very large
  - Primary focus on binary end-points
    - Nested case-control studies
  - Hundreds of thousands of recruits
  - e.g. UK Biobank, LifeGene, Kadoorie Study
- Large
  - Primary focus on quantitative end-points
  - Tens of thousands of recruits
  - e.g. ALSPAC, Generation Scotland, CARTaGENE





#### The £59,000,000 question!!!

- Just how big does a cohort-based biobank have to be?
  - Interest in binary disease related events and (some) binary exposures
  - Middle aged recruits (40-69 years)
  - Population-based recruitment



# The statistical power of case-control studies

- Contemporary pre-eminence of genetic association studies rather than genetic linkage studies
- Covers *both* stand-alone case-control studies, *and* nested case-control studies in large cohorts. Main issue is the number of cases.
- Sample size determining in *both* settings



#### Simulation-based power calculations

- Work with the least powerful (common) setting
  - Disease outcome and exposures all binary
- Logistic regression; interactions = departure from a multiplicative model
- Four controls per case
- Complexity (arbitrary but realistic)



#### **Formal power calculations**

- Realistic bio-analytic complexity
  - Logistic regression
  - Assessment errors, frailty, p<10<sup>-4</sup>,10<sup>-7</sup>,10<sup>-10</sup>
  - $\approx$  4 controls per case

**ESPRESSO**: (Estimating Sample-size and Power in **R** by Exploring Simulated Study Outcomes). <u>http://www.p3gobservatory.org/powercalculator.htm</u>

See also: Paul R Burton; Anna L Hansell; Isabel Fortier; Teri A Manolio; Muin J Khoury; Julian Little; Paul Elliott. International Journal of Epidemiology 2008; doi: 10.1093/ije/dyn147



RG



How small is "small"?

Most in range: 1.1 – 1.5

PROMOTING HARMONISATION OF

**EPIDEMIOLOGICAI** 

**d**G

Disease	Gene	Polymorphism	Approximate frequency of the disease associated allele	Approximate odds ratio for disease associated allele	Ref
Thrombophilia	F5	Leiden Arg506Gln	0.03	4	12
Crohn's disease	CARD15	3 SNPs	0.06(composite)	4.6	67
Alzheimer's disease	APOE	ε2/3/4	0.15	3.3	13,68
Osteoporotic fractures	COL1A1	Sp1 restriction site	0.19	1.3	69,70
Type 2 diabetes	KCNJ11	Glu23Lys	0.36	1.23	71
Type 1 diabetes	CTLA4	Thr17Ala	0.36	1.27	72,73
Graves' Disease	CTLA4	Thr17Ala	0.36	1.6	74
Type 1 diabetes	INS	5' VNTR	0.67	1.2	75
Bladder Cancer	GSTM1	Null (gene deletion)	0.70	1.28	76
Type 2 diabetes	PPARG	Pro12Ala	0.85	1.23	11

Hattersley AT, McCarthy MI. Lancet 2005;366:1315-1323 Examples of some polymorphisms or haplotypes that have shown consistent association with complex disease

#### **Recent findings\*\*\***

Type 1 diabetes<sup>1,2</sup> Type 2 diabetes<sup>2,6</sup> Coronary heart disease<sup>2,7-9</sup> Breast cancer<sup>10,11</sup> Colorectal cancer<sup>12-14</sup> Prostate cancer<sup>15,16</sup> Age-related macular degeneration<sup>17-19</sup> Crohns disease<sup>2,20</sup>

\*\*\*See full reference list in reserve slides



### An example: Diabetes mellitus defined by Hba1C>97.5 percentile

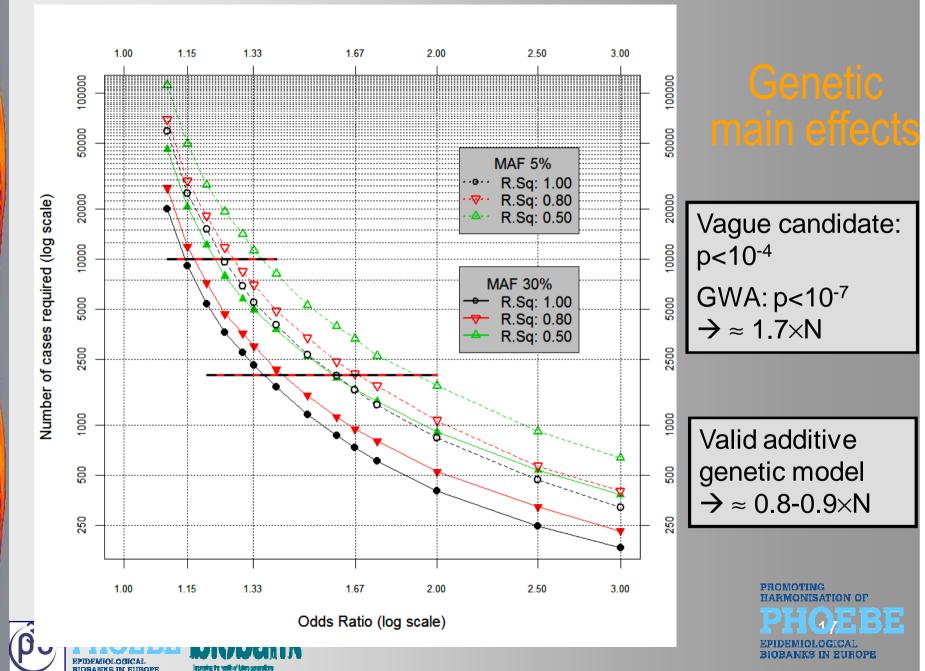


#### Simulation-based power calculations

Complexity (arbitrary but realistic).

Frailty variance	Genotyping error	Environmental error	Sensitivity disease phenotype	Specificity disease phenotype	Critical P-value	Power
10 fold	R <sup>2</sup> = 0.5, 0.8	Reliability = 0.3-1.0	89%	97.4%	10 <sup>-4</sup> 10 <sup>-7</sup>	80%





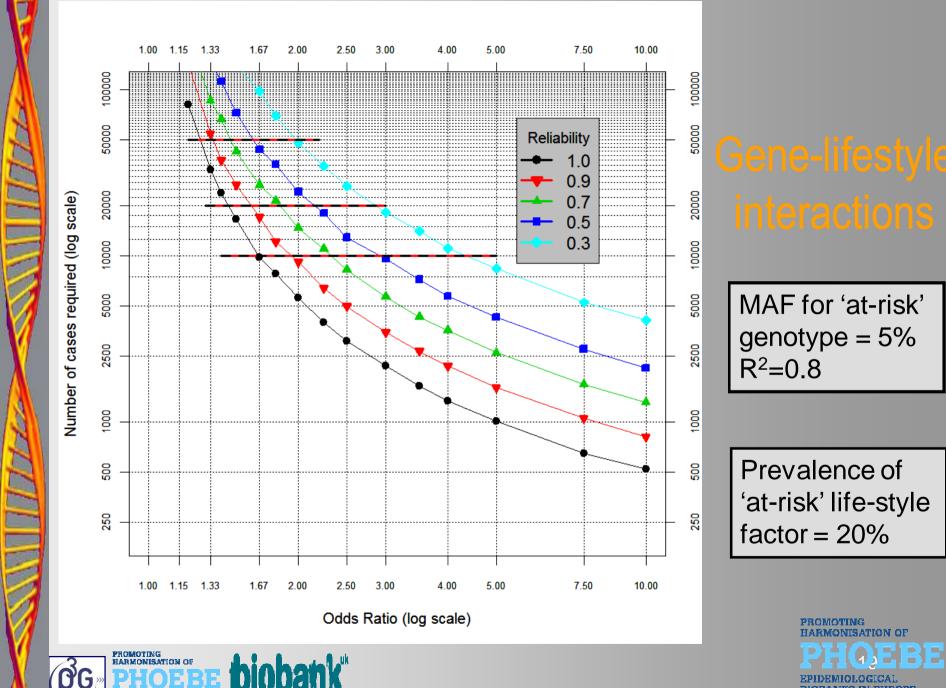
#### **Gene-lifestyle interactions**

Table 2 Formal estimates of test-retest reliability for a number of exemplar lifestyle/environmental determinants that are widely studied

Reliability of measurement	Lifestyle/environmental factor
≥0.95	Body mass index (BMI) calculated from measured height and weight in various studies <sup>76</sup>
~0.9	Measured hip or waist circumference <sup>76,77</sup>
~0.7	Blood pressure measurement in the Intersalt Study <sup>78</sup>
~0.5	Many nutritional components in a dietary recall study, mean of four 24 h assessments <sup>79</sup>
~0.3	Many nutritional components in a dietary recall study, a single 24h assessment <sup>79</sup>



PROMOTING HARMONISATION OF PHOSEBE EPIDEMIOLOGICAL BIOBANKS IN EUROPE



**EPIDEMIOLOGICAI** RIORANKS IN EUROP **BIOBANKS IN EUROPE** 

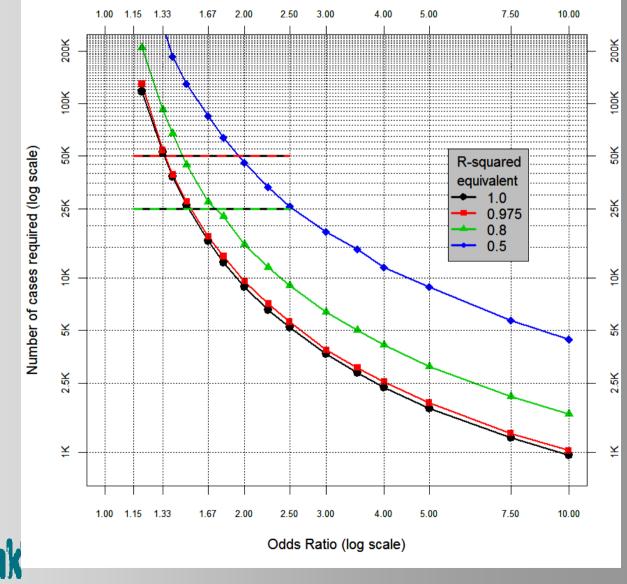
# Gene:gene interactions

MAF 5% : 25% 4 controls/case GWA: *p*<10<sup>-10</sup>

PROMOTING HARMONISATION OF

EPIDEMIOLOGICAL BIORANKS IN EUROP

ÔG



#### Additive genetic model

- Binary or additive genetic model?
- If truly additive, additive model could add substantial power
- If truly binary, binary model is slightly more powerful
- But, the gain in power is greater when MAF is high
  - When MAF is low, very few subjects are homozygote for MA, and so the locus is almost binary, and the fall in required sample size is small
  - But when MAF is high, power is not such an issue

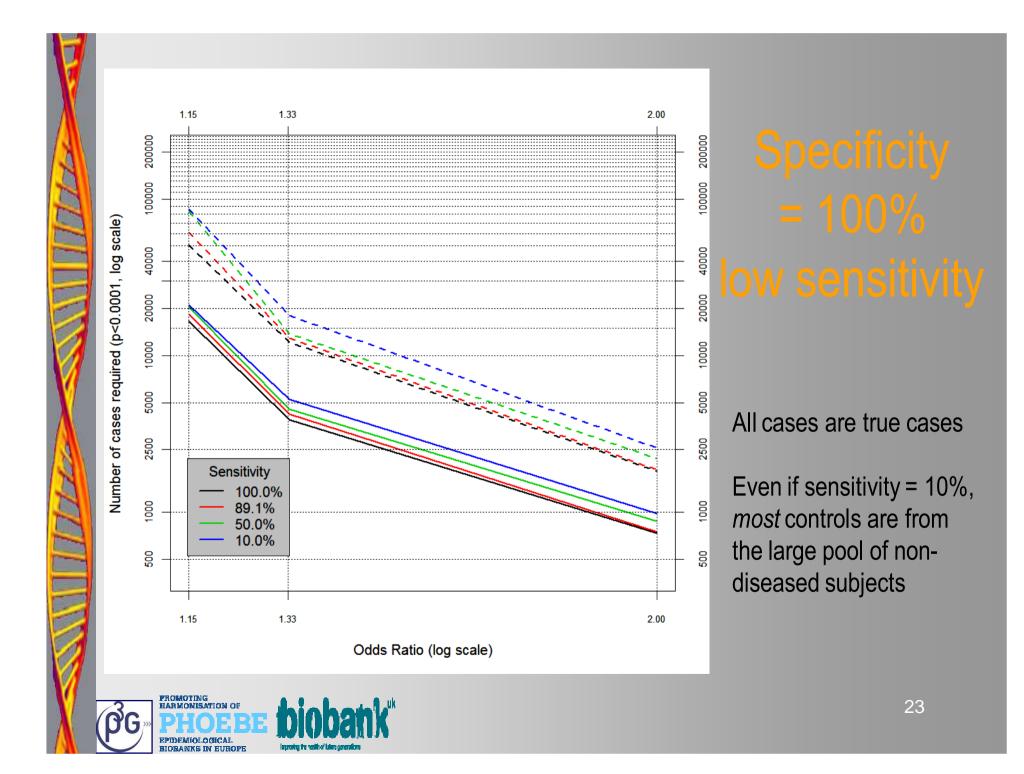


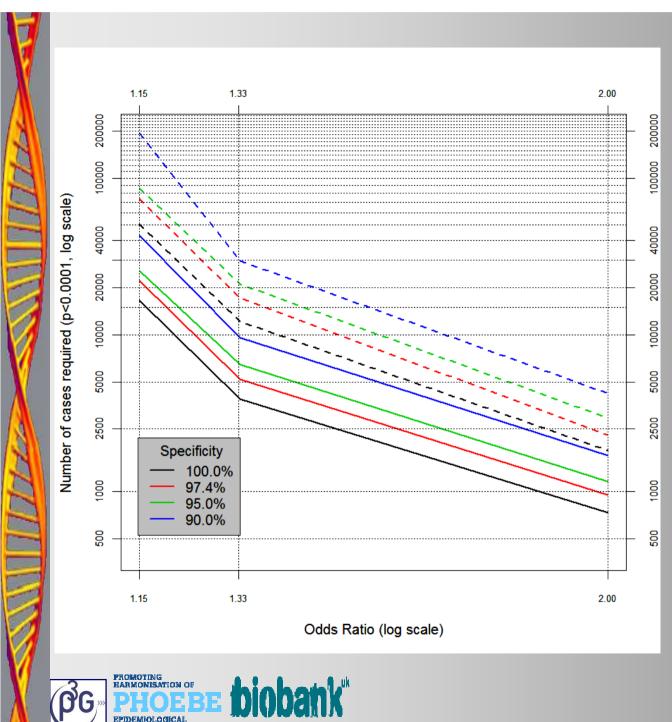
**A**C

## Be specific, not sensitive

but only if you have time!!







Sensitivity= 100% low specificity

All controls are true controls

Even if specificity is as high as 90%, *many* "cases" are from the large pool of truly nondiseased subjects that have been misclassified

#### How many cases?

- Genetic main effects
  - 2,000 minimum, 5,000 better
- Lifestyle main effects
  - 2,000-20,000
- Gene-lifestyle "interactions"
  - Absolute minimum 10,000, often need at least 25,000, a comprehensive platform needs at least 50,000
- Pooling and replication!!

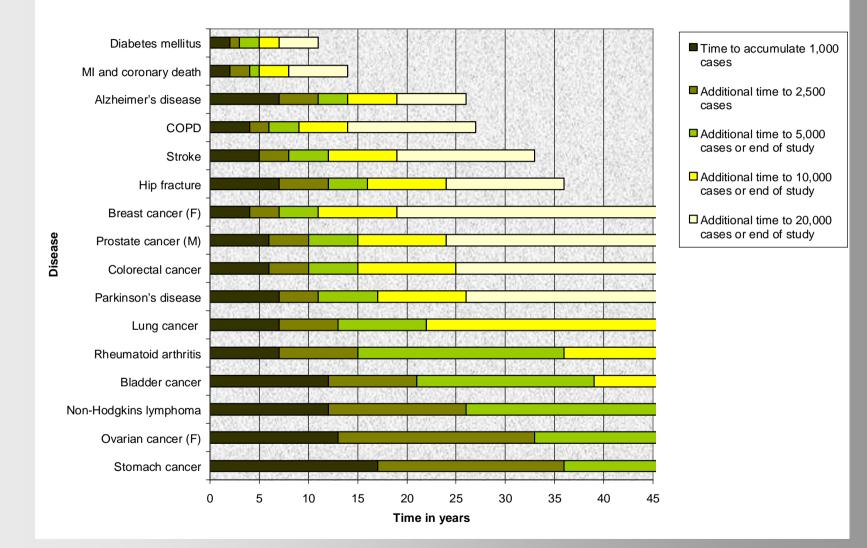


#### How long is "LONG"?

- Age range at recruitment 40-69 years
- Recruitment over 5 years
- All cause mortality
- Disease incidence ("healthy cohort effect")
- Migration overseas
- Withdrawal from the study



#### How long is "LONG"?





#### So how can we get enough power?

- Can it be achieved at all?
  - Recent successes
- Is genetic epidemiology beyond its limits?
  - Taubes, Science, 1995
  - Protection afforded by 'Mendelian Randomization'
- Large disease-based biobanks
- Very large cohort-based biobanks
- CDCV v rare alleles with large effects



#### So how can we get enough power?

- Conduct studies with optimal designs
  - Enhance the quality of individual studies.
  - When relevant, use continuous disease-related traits and health determinants.
- Increase the size of individual studies
- Promote the conduct of meta-analysis
  - Sharing results (traditional meta-analysis): Ideally based on published and unpublished results.
  - Sharing raw data and samples: Need to promote harmonization between biobanks to enable pooling of raw information.



#### International biobank harmonization programs

- P<sup>3</sup>G
  - Public Population Program in Genomics (P<sup>3</sup>G)
- PHOEBE
  - Promoting Harmonization Of Epidemiological Biobanks in Europe
- BBMRI
  - Biobanking and BioMolecular Resources Research Infrastructure
- ISBER
  - International Society for Biological and Environmental Repositories
- HuGENet

**A**C

Human Genome Epidemiology Network



#### Biobanks associated with P<sup>3</sup>G



# Number of participants targeted (recruited or to be recruited) (N=87)

Number of participants	Number of studies	Number of participants TARGETED
Less than 49 000	47	900,000
50 000 to 99 000	14	1,000,000
100 000 to 499 000	18	2,800,000
500 000 and more	8	4,900,000
PROMOTING HARMONISATION OF high and uk		Total: 9,600,000



Extracted from P<sup>3</sup>G Observatory

## Thanks!!



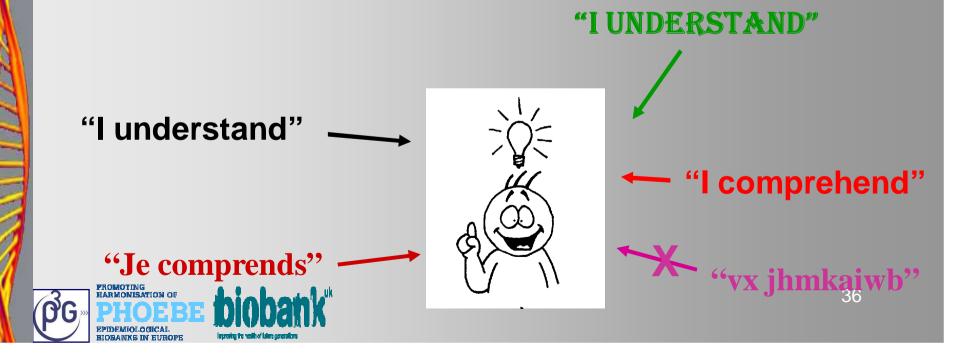


### **Reserve slides**



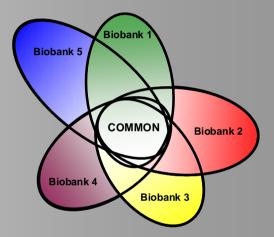
#### **Biobank harmonization**

 "A set of procedures that promote, both now and in the future, the effective interchange of valid information and samples between a number of studies or biobanks, accepting that there may be important differences between those studies"



# Information common to all the cohorts and information specific to some of them

- Questionnaire
- Physical and cognitive measures
- Environmental measures
- Biochemical measures
- Governmental databases

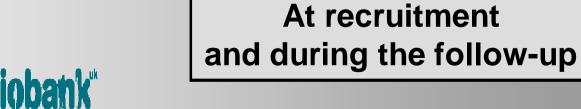


DATASHaPER

PROMOTING

**N**N

• Template to facilitate harmonization between pre-existing biobanks and support the design of emerging ones.



#### References for slide 16

- 7. Wellcome\_Trust\_Case\_Control\_Consortium. Genome-wide association study of 14,000 cases of seven common diseases and 3,000 shared controls. *Nature* 2007;447:661-678.
- 32. Todd JA, Walker NM, Cooper JD, Smyth DJ, Downes K, Plagnol V, et al. Robust associations of four new chromosome regions from genome-wide analyses of type 1 diabetes. Nat Genet 2007;39(7):857-864.

**d**G

**EPIDEMIOLOGICA**I

- 33. Zeggini E, Weedon MN, Lindgren CM, Frayling TM, Elliott KS, Lango H, et al. Replication of Genome-Wide Association Signals in U.K. Samples Reveals Risk Loci for Type 2 Diabetes. *Sciencexpress* 2007;10.1126:1-4.
- 34. Saxena R, Voight BF, Lyssenko V, Burtt NP, de Bakker PI, Chen H, et al. Genome-wide association analysis identifies loci for type 2 diabetes and triglyceride levels. Science 2007;316(5829):1331-6.
- 35. Scott LJ, Mohlke KL, Bonnycastle LL, Willer CJ, Li Y, Duren WL, et al. A genome-wide association study of type 2 diabetes in Finns detects multiple susceptibility variants. Science 2007;316(5829):1341-5.
- 36. Samani NJ, Erdmann J, Hall AS, Hengstenberg C, Mangino M, Mayer B, et al. Analysis of two genome-wide association studies identifies and validates novel gene loci for myocardial infarction. New England Journal of Medicine 2007:in press.
- 37. Helgadottir A, Thorleifsson G, Manolescu A, Gretarsdottir S, Blondal T, Jonasdottir A, et al. A common variant on chromosome 9p21 affects the risk of myocardial infarction. Science 2007;316(5830):1491-3.
- 38. McPherson R, Pertsemlidis A, Kavaslar N, Stewart A, Roberts R, Cox DR, et al. A common allele on chromosome 9 associated with coronary heart disease. Science 1488-91. 38



- 39. Easton DF, Pooley KA, Dunning AM, Pharoah PDP, Thompson D, Ballinger DG, et al. Genomewide association study identifies novel breast cancer susceptibility loci. *Nature* 2007;advanced online publication.
- 40. Stacey SN, Manolescu A, Sulem P, Rafnar T, Gudmundsson J, Gudjonsson SA, et al. Common variants on chromosomes 2q35 and 16q12 confer susceptibility to estrogen receptor-positive breast cancer. *Nat Genet* 2007;39(7):865-9.
- 41. Haiman CA, Le Marchand L, Yamamato J, Stram DO, Sheng X, Kolonel LN, et al. A common genetic risk factor for colorectal and prostate cancer. *Nat Genet* 2007.

- 42. Tomlinson I, Webb E, Carvajal-Carmona L, Broderick P, Kemp Z, Spain S, et al. A genome-wide association scan of tag SNPs identifies a susceptibility variant for colorectal cancer at 8q24.21. Nat Genet 2007.
- 43. Zanke BW, Greenwood CM, Rangrej J, Kustra R, Tenesa A, Farrington SM, et al. Genome-wide association scan identifies a colorectal cancer susceptibility locus on chromosome 8q24. Nat Genet 2007.
- 44. Gudmundsson J, Sulem P, Steinthorsdottir V, Bergthorsson JT, Thorleifsson G, Manolescu A, et al. Two variants on chromosome 17 confer prostate cancer risk, and the one in TCF2 protects against type 2 diabetes. *Nat Genet* 2007.
- 45. Gudmundsson J, Sulem P, Manolescu A, Amundadottir LT, Gudbjartsson D, Helgason A, et al. Genome-wide association study identifies a second prostate cancer susceptibility variant at 8q24. Nat Genet 2007;39(5):631-7.
- 46. Klein RJ, Zeiss C, Chew EY, Tsai JY, Sackler RS, Haynes C, et al. Complement factor H polymorphism in age-related macular degeneration. *Science* 2005;308(5720):385-9.
- 47. Haines JL, Hauser MA, Schmidt S, Scott WK, Olson LM, Gallins P, et al. Complement factor H variant increases the risk of age-related macular degeneration. *Science* 2005;308(5720):419-21.
- 48. Edwards AO, Ritter R, 3rd, Abel KJ, Manning A, Panhuysen C, Farrer LA. Complement factor H polymorphism and age-related macular degeneration. *Science* 2005;308(5720):421-4.
- 49. Rioux JD, Xavier RJ, Taylor KD, Silverberg MS, Goyette P, Huett A, et al. Genome-wide
  An association study identifies new susceptibility loci for Crohn disease and implicates autophagy in disease pathogenesis. Nat Genet 2007;39(5):596-604.