Integrated transcriptional profiling and linkage analysis for mapping disease genes and regulatory gene networks analysis

Enrico Petretto

Research Fellow in Genomic Medicine

Imperial College Faculty of Medicine enrico.petretto@imperial.ac.uk

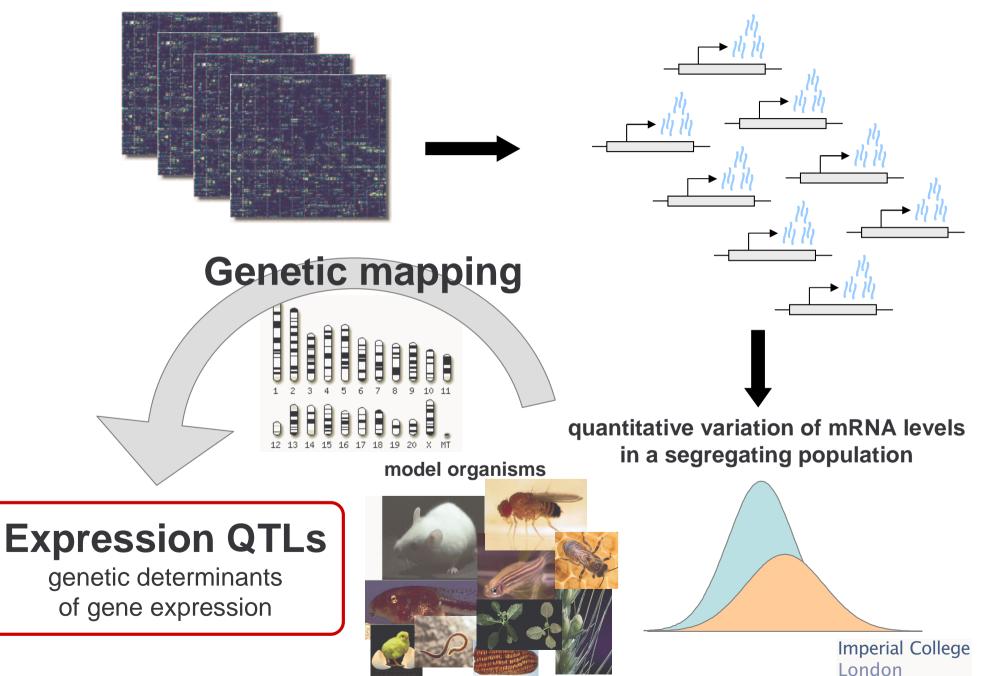




Outline

- Introduction: the biological framework
 - Expression QTL mapping using animal models
 - eQTL analysis in multiple tissues
- Integrating genome-wide eQTL data to identify gene association networks
 - Data mining of eQTLs
 - Graphical Gaussian models (GGMs)
 - Example of identification of disregulated pathway
 - Master transcriptional regulator

Genetical Genomics



The rat is among the leading model species for research in physiology, pharmacology, toxicology and for the study of genetically complex human diseases

Spontaneously Hypertensive Rat (SHR): A model of the metabolic syndrome

- Spontaneous hypertension
- Decreased insulin action
- Hyperinsulinaemia
- Central obesity
- Defective fatty acid metabolism
- Hypertriglyceridaemia



Specialized tools for genetic mapping: Rat Recombinant Inbred (RI) strains



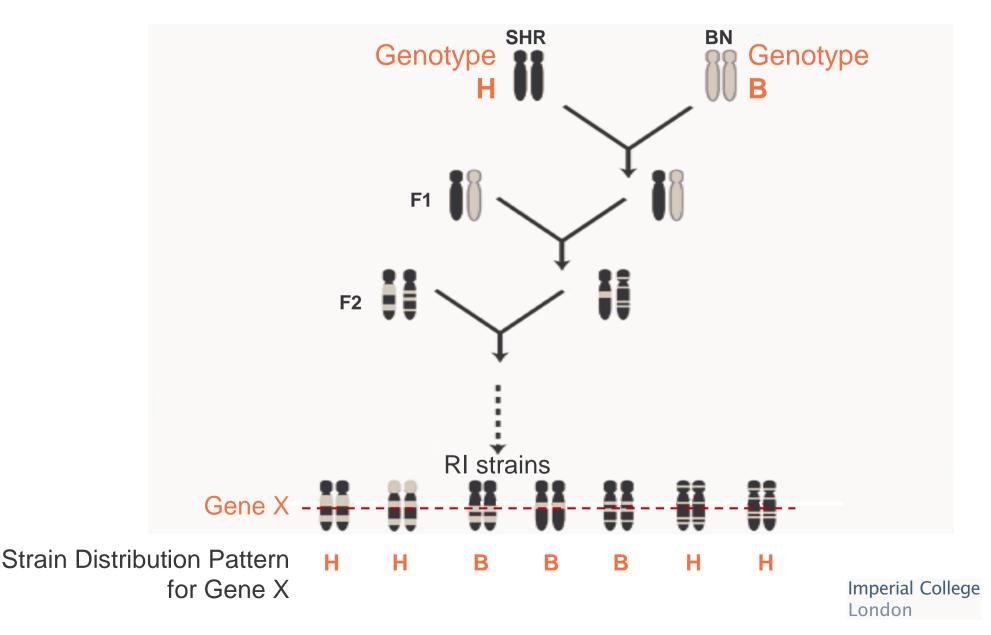
Mate two inbred strains

F1 offspring are identical

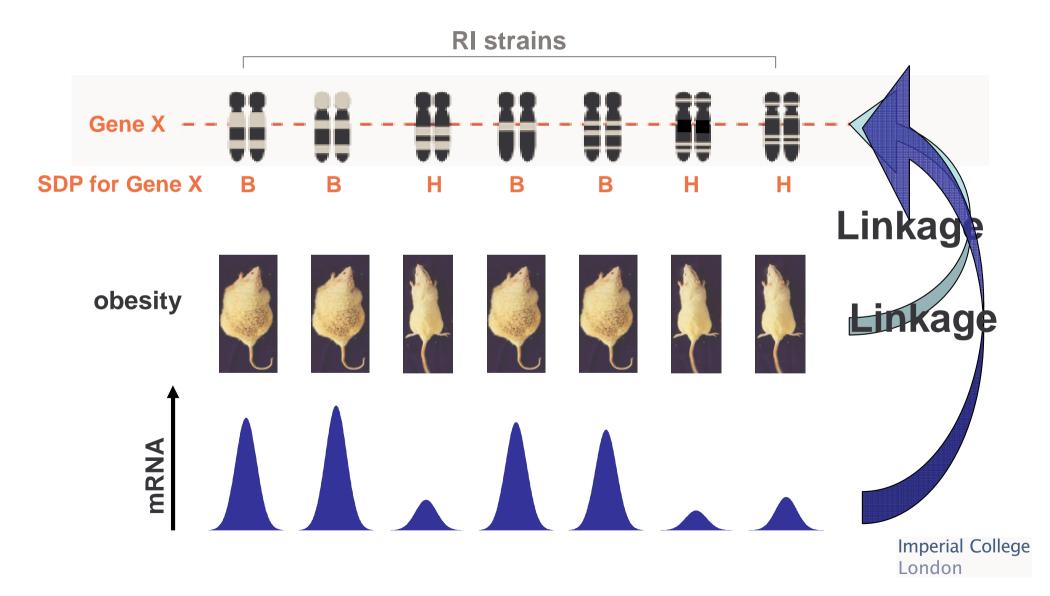
F2 offspring are different (due to recombination)

Brother sister mating over >20 generations to achieve homozygosity at all genetic loci

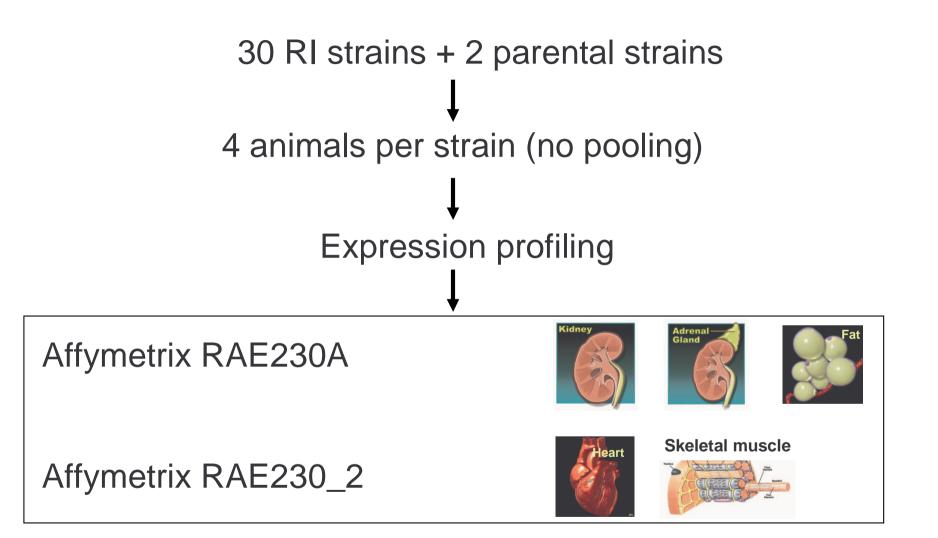
Cumulative, renewable resource for phenotypes and genetic mapping



Mapping of QTLs compare strain distribution pattern of markers and traits



Gene expression analysis in the Rat

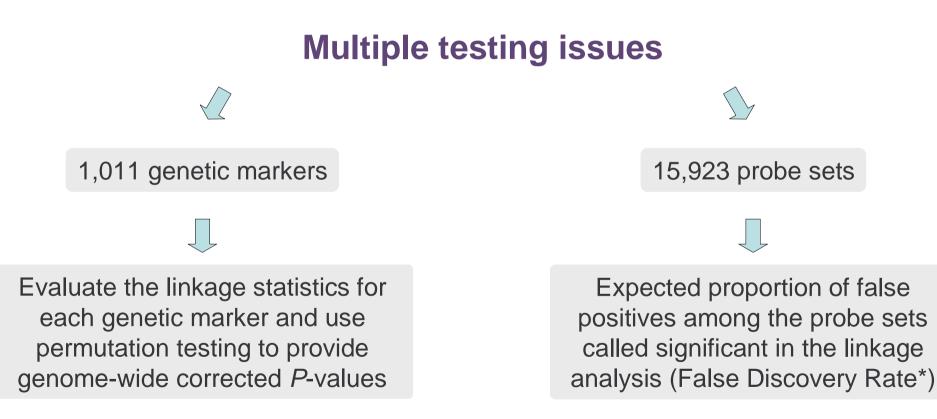


640 microarray data sets

- ~ 16,000 probe sets per array (fat, kidney, adrenal)
- ~ 30,000 probe sets per array (heart, skeletal muscle) mperial College

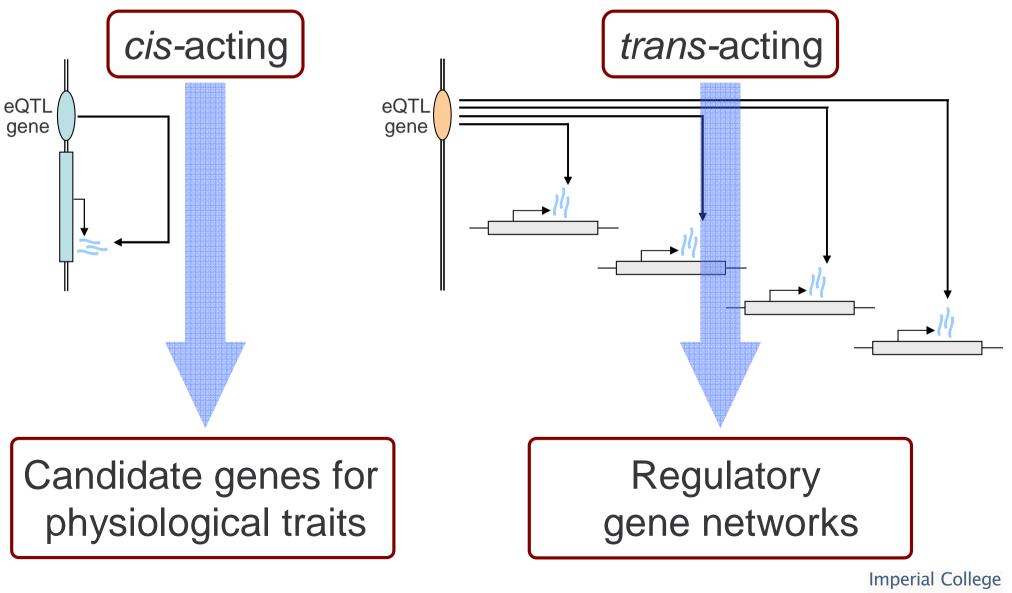
eQTL Linkage Analysis

 For each probe set on the microarray, expression profiles were regressed against all 1,011 genetic markers



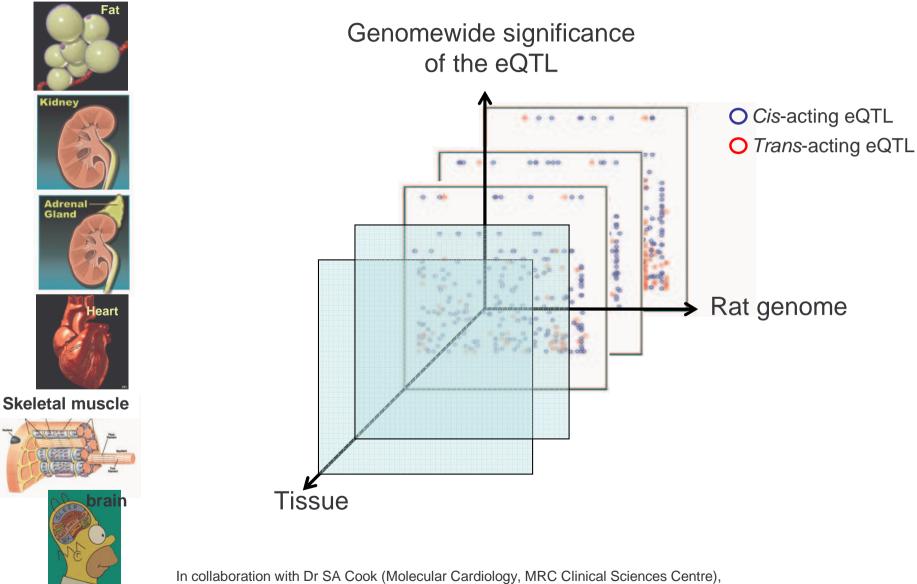
* Storey 2000

cis- and trans-acting eQTLs



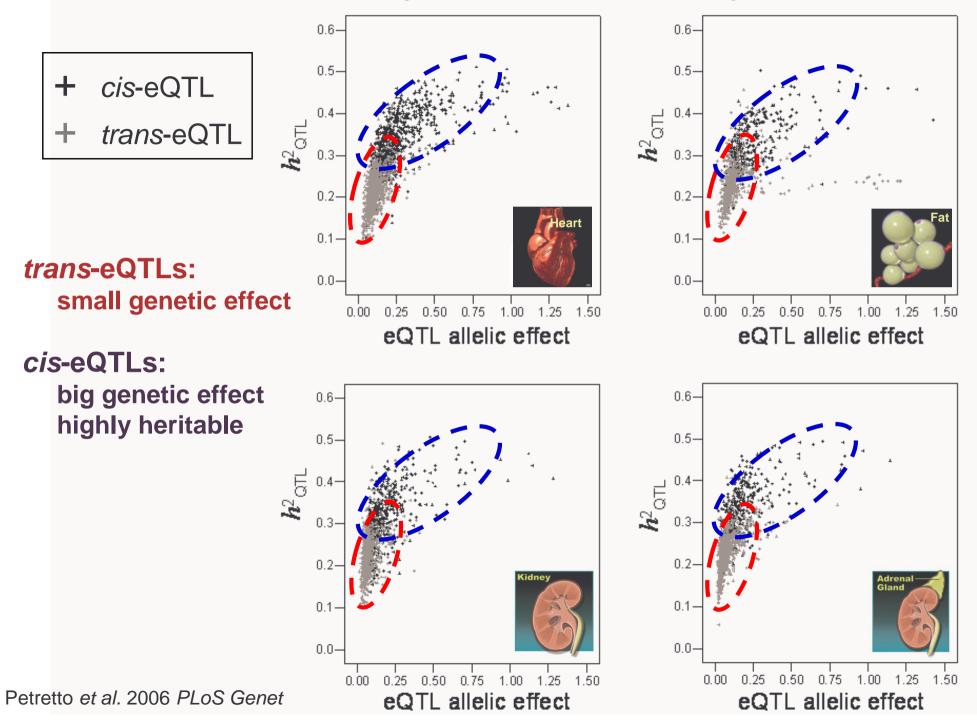
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eQTL datasets in the rat model system

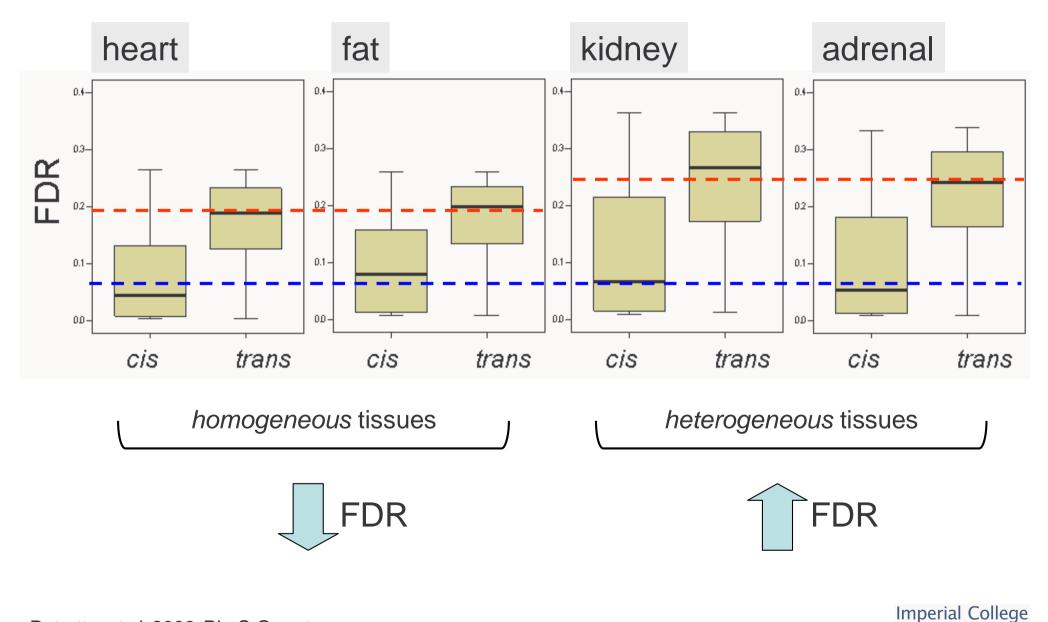


In collaboration with Dr SA Cook (Molecular Cardiology, MRC Clinical Sciences Centre), Dr M Pravenec (Czech Academy of Sciences, Prague) and Prof N Hubner (MDC, Berlin)

Genetic architecture of genetic variation in gene expression



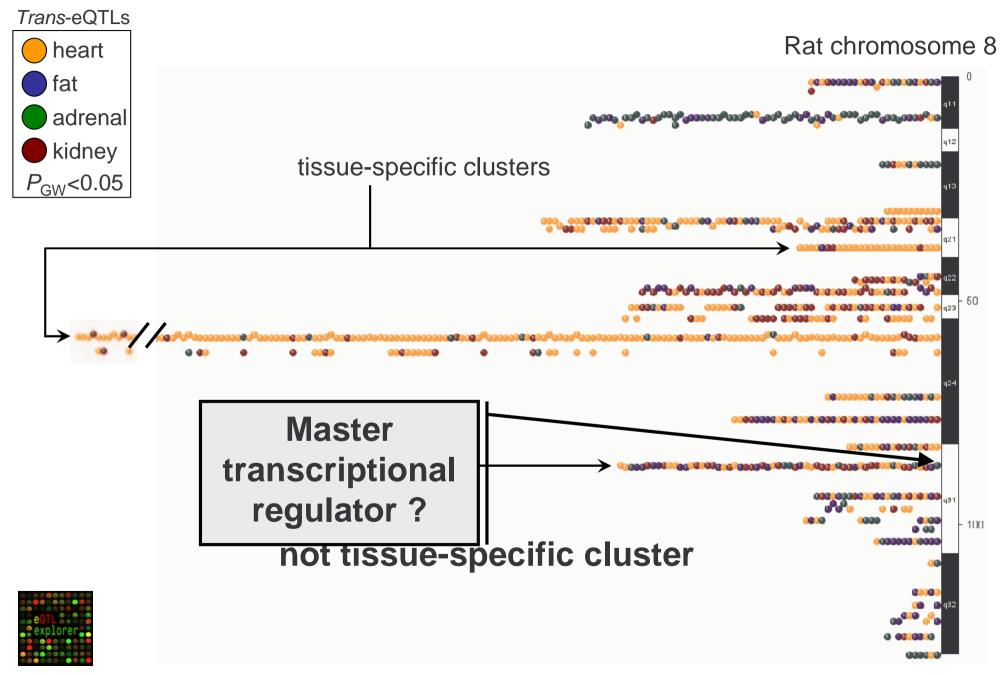
FDR for cis- and trans-eQTLs

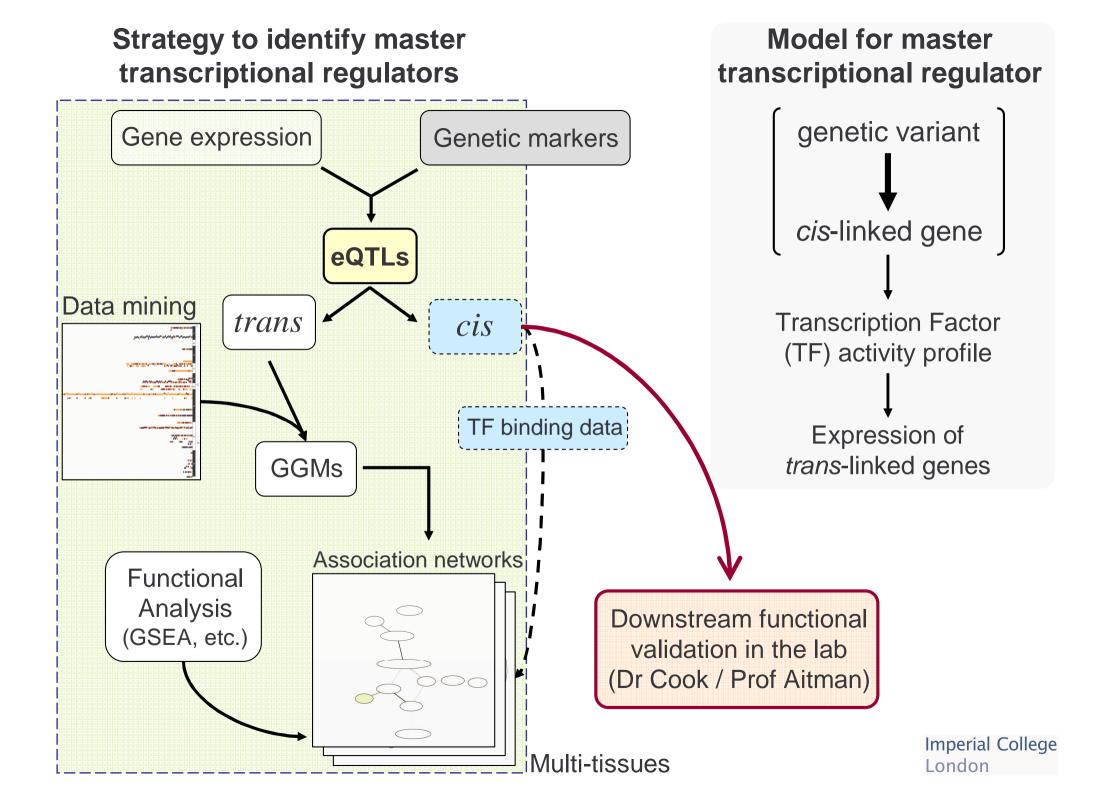


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Petretto et al. 2006 PLoS Genet

trans-eQTLs hot-spots





GGMs

- Partial correlation matrix $\Pi = (\pi_{ij})$
- Inverse of variance covariance matrix *P*

$$\Omega = (\omega_{ij}) = P^{-1}$$

$$\pi_{ij} = -\omega_{ij} / (\omega_{ii} \omega_{jj})^{\frac{1}{2}}$$

- small *n*, large *p*
- Regularized covariance matrix estimator by shrinkage (Ledoit-Wolf approach)
 - Guarantees positive definiteness

Schafer and Strimmer 2004, Rainer and Strimmer 2007

Partial correlation graphs

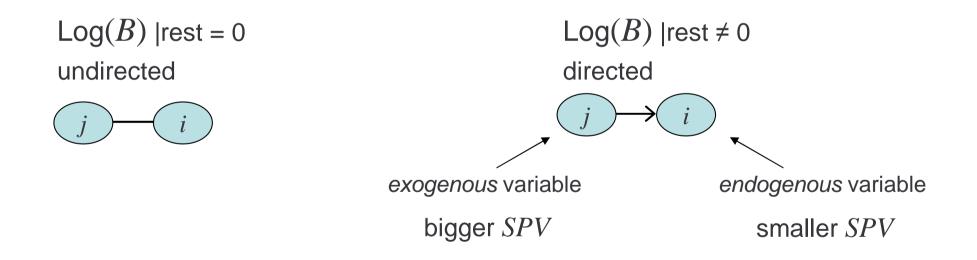
- Multiple testing on all partial correlations
 - Fitting a mixture distribution to the observed partial correlations (p)

Schafer and Strimmer 2004, Rainer and Strimmer 2007

GGMs

Infer partial ordering of the node

- Standardized partial variances (SPV_i)
- Proportion of the variance that remains unexplained after regressing against all other variables
- Log-ratios of standardized partial variances $B = (SPV_i / SPV_j)^{\frac{1}{2}}$



Inclusion of a directed edge into the network is conditional on a non-zero partial correlation coefficient

Schafer and Strimmer 2004, Rainer and Strimmer 2007

Hypothesis driven analysis

- 1. Gene expression levels under genetic control (i.e., 'structural' genetic perturbation)
- 2. Co-expression of *trans*-eQTLs point to common regulation by a single gene

Graphical Gaussian models

• Detect conditionally dependent *trans*-eQTL genes

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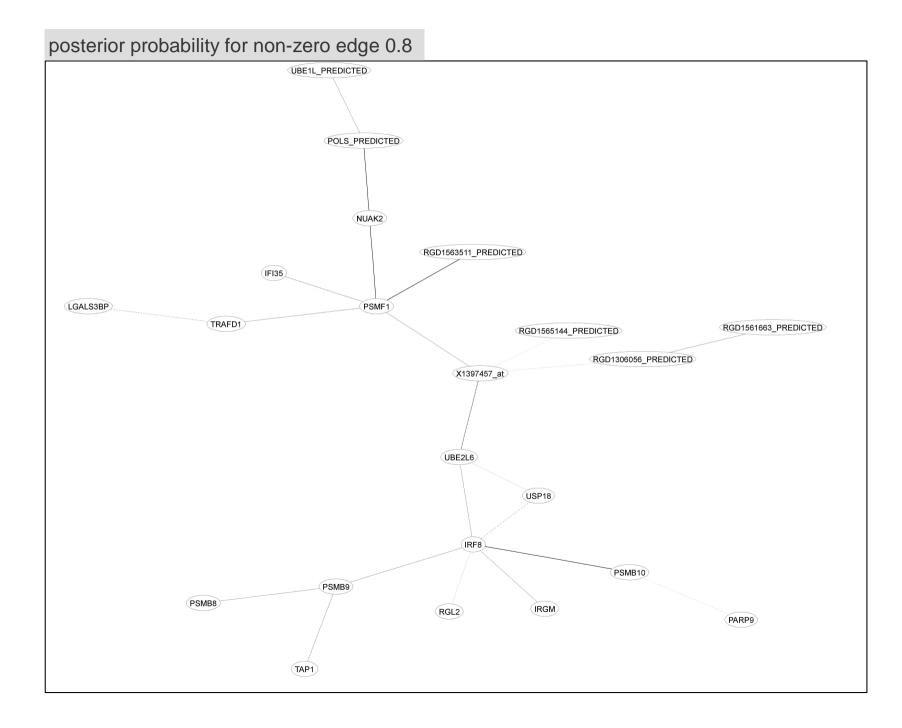
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 Infer partial ordering of the nodes (directed edges)

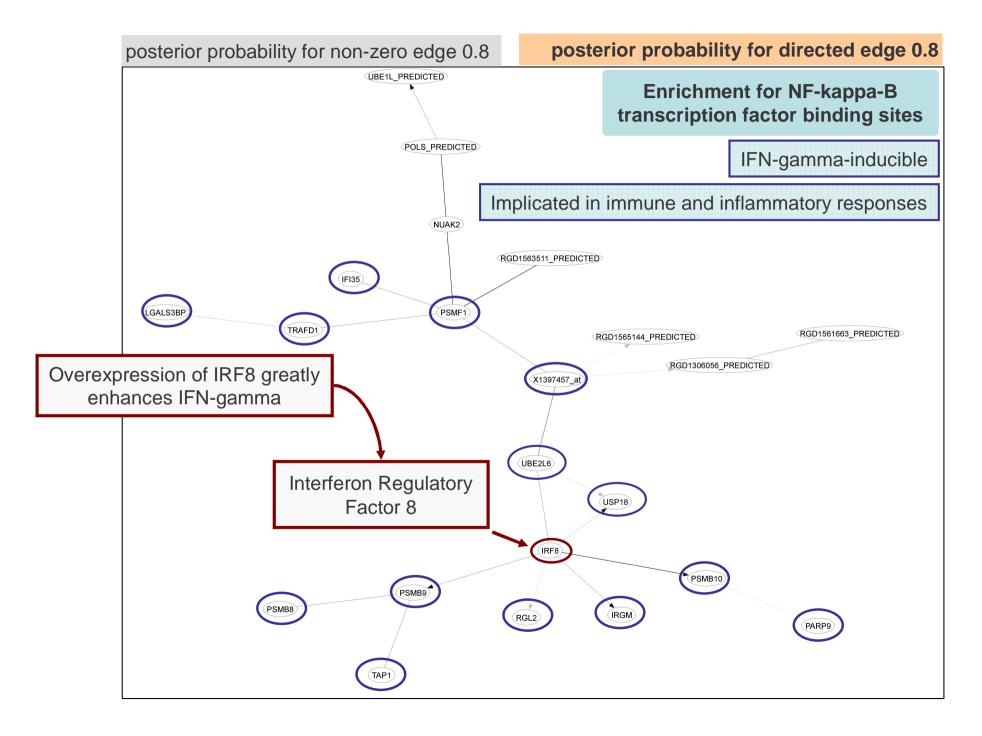
120 160 100 140 80 20 40 60 0 c17.6 c17.38 c15.108 c15.11 c16.0 c11.31 c15.75 Chromosome 15, 108 Mb, D15Rat29 trans-eQTLs hot spots c6.136 c4.93 c15.78 c1.87 c10.25 c11.32 Locus (chromosome.Mb) c4.148 c8.45 c8.87 c8.53 c4.91 c4.161 c10.21 c4.151 c16.46 c15.80 c17.40 c8.9 c16.50 c3.41 c20.44 c3.112 c8.49 c13.9 c17.87 c3.130 c5.151 kidney fat adrenal heart c7.142 c8.32 c15.58 c13.38 c1.248 c8.38 c1.90 c12.7 c3.129 c6.131 Imperial College

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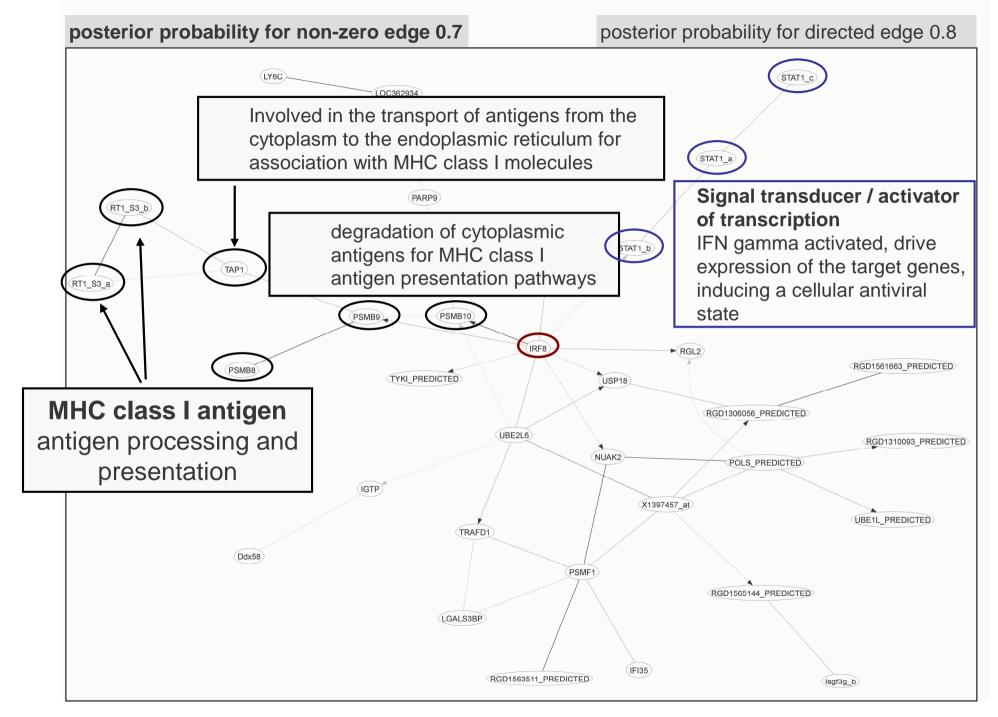
Heart tissue, *trans*-eQTLs hot-spot (chromosome 15)



Heart tissue, trans-eQTLs hot-spot (chromosome 15)

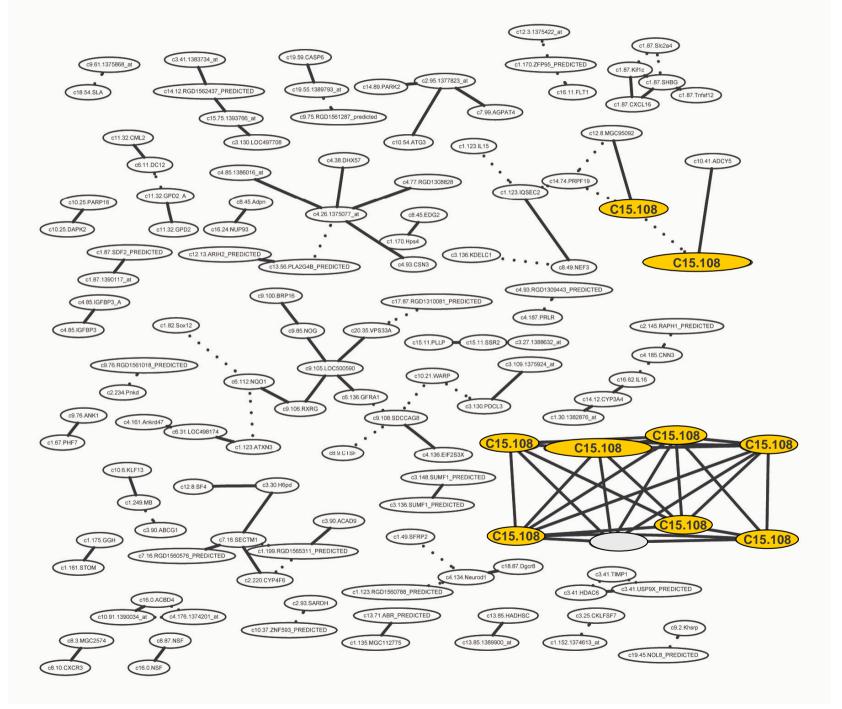


Relaxing the threshold...

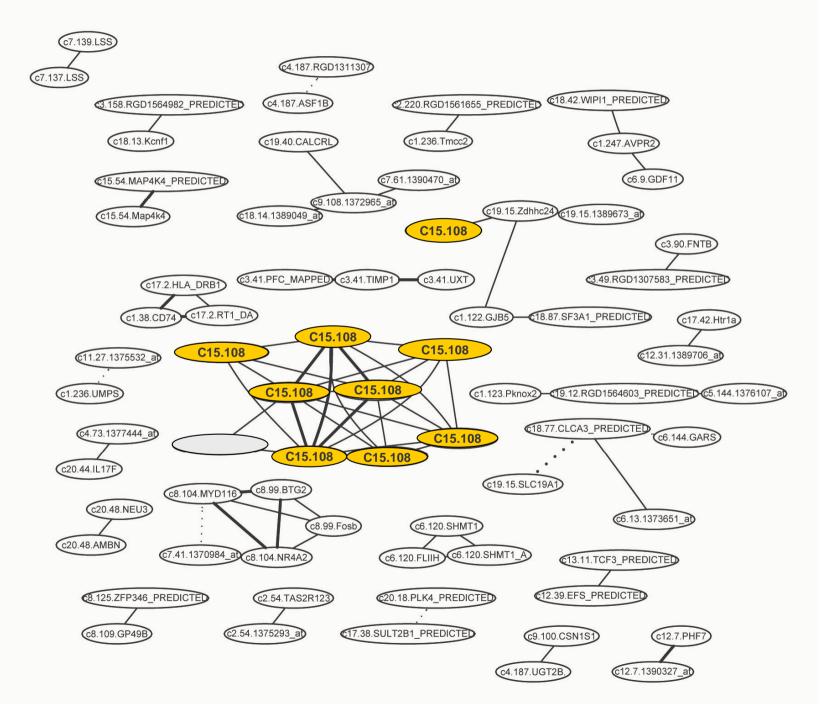


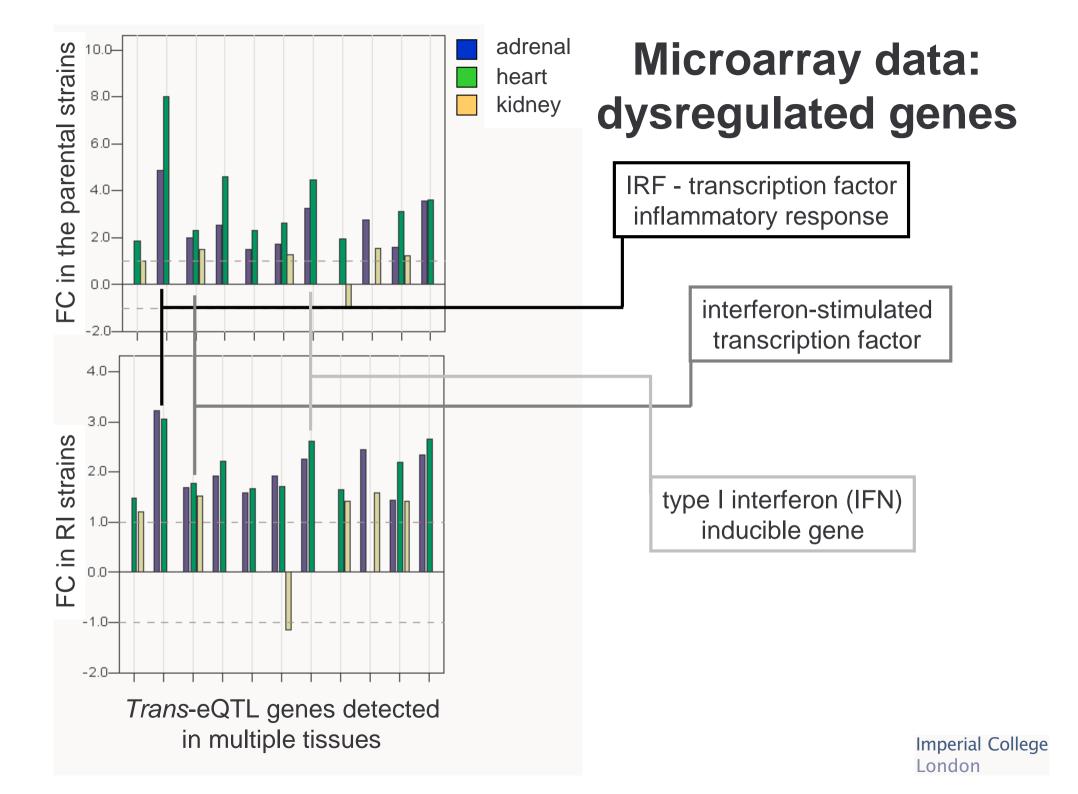
Is this association graph tissue specific?

kidney, all trans-eQTLs, posterior probability 0.95



Adrenal, all trans-eQTLs, posterior probability 0.95





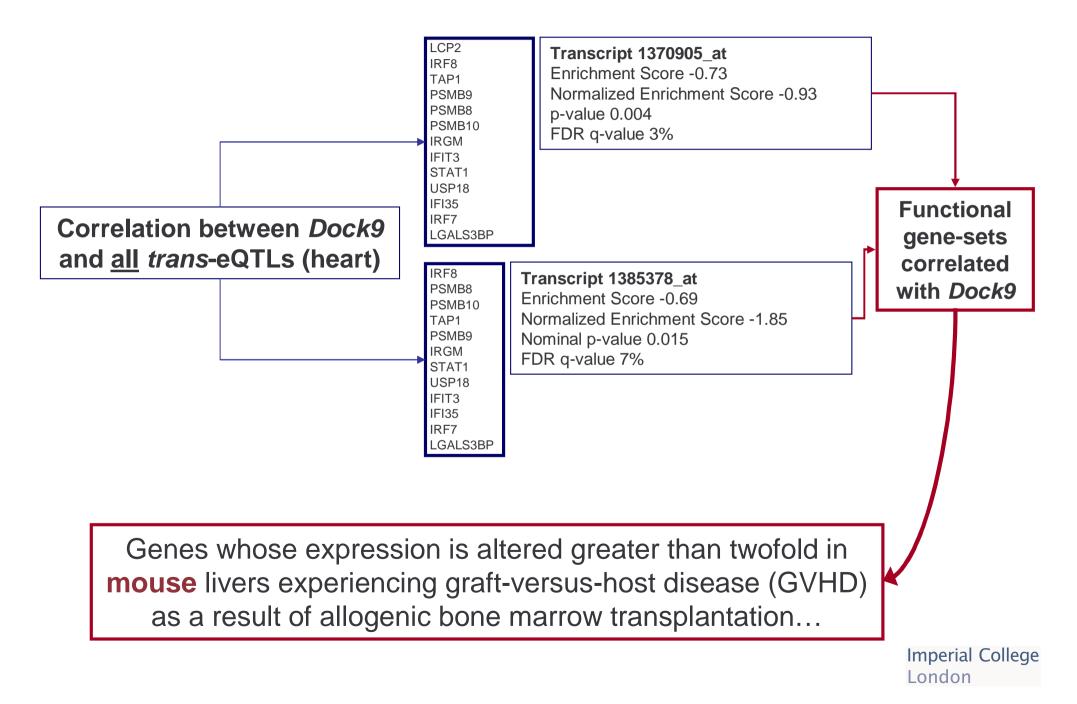
cis-acting eQTLs within the cluster region

Transcripts representing *Dock9* gene

transcriptional regulator genetic variant cis-linked gene **Transcription Factor** (TF) activity profile Expression of trans-linked genes

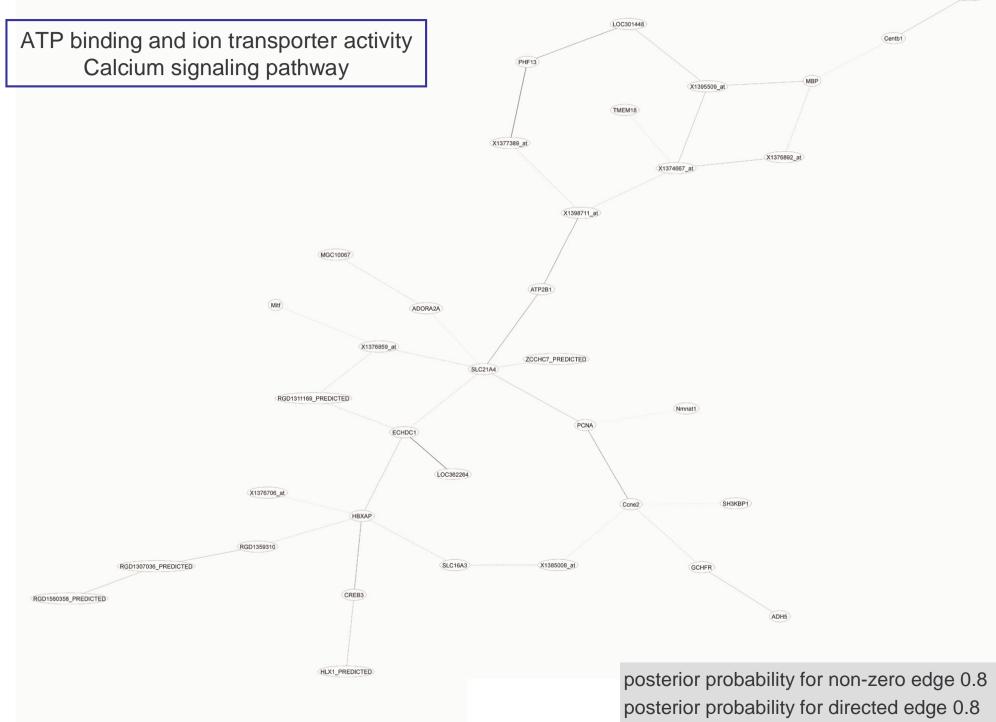
Model for master

<u>Gene Set Enrichment Analysis</u>

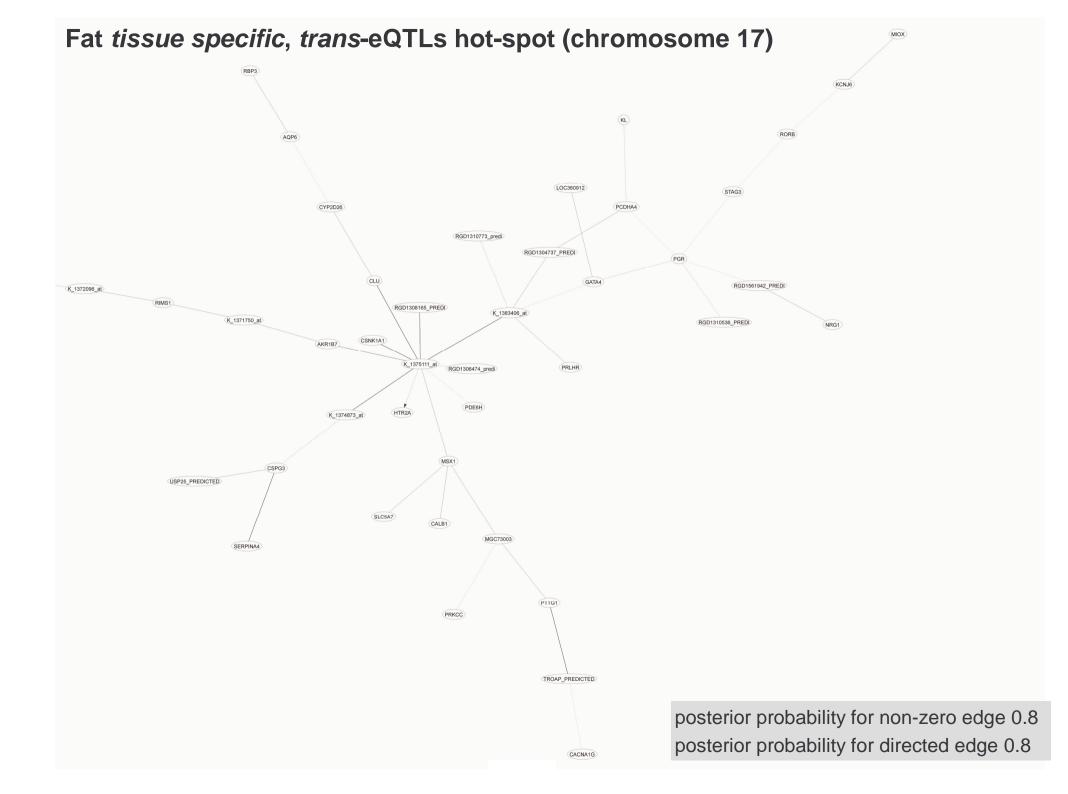


Other examples

Heart tissue, trans-eQTLs hot-spot (chromosome 15, 78Mb)



X1383133 at



Summary

- Genome-wide eQTL data provide new insights into gene regulatory networks
- GGMs applied to *trans*-eQTL hotspots identified dysregulated pathway related to inflammation

 Hypothesis-driven inference can be a powerful approach to dissect regulatory networks

Acknowledgments

<u>Sylvia Richardson</u> Tim Aitman

<u>Stuart Cook</u> Jonathan Mangion Rizwan Sarwar



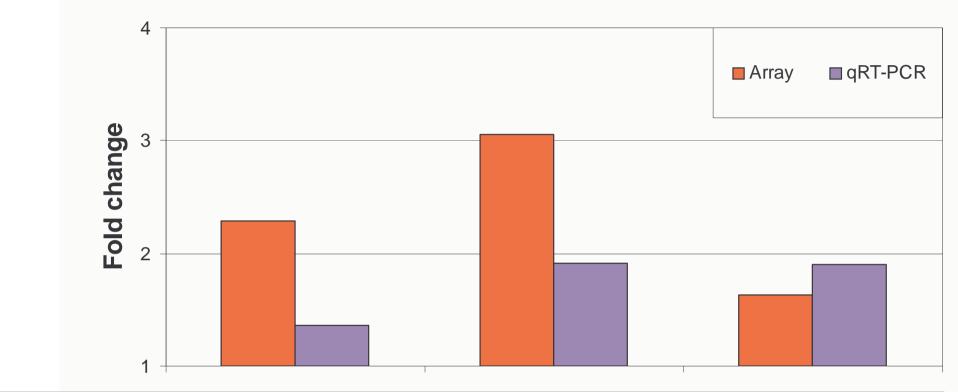
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<u>collaborators:</u> Norbert Hubner (MDC, Berlin) Michael Pravenec (Institute of Physiology, Prague)

Extra slides

Chr 15 qRT-PCR validation in RI strains



Gene	Rarresin1_pred	Irf7_pred	Stat1
Array	2.28	3.06	1.63
Р	4.0E-05	8.6E-05	1.4E-04
	•	•	
qRT-PCR	1.36	1.91	1.90
Р	0.039	0.004	0.036

Rpt4 and *Irf7* mRNA levels increase in response to interferon

- H9c2 cells (rat cardiac embryonic myoblast)
- Stimulated with recombinant rat interferon for 3 hours
- RNA extracted, assayed by qRT-PCR (SYBR Green I)
- 3 independent expts, 3 biological replicates

