

# Family-based analysis of genome-wide gene $\times$ gene interactions

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## Motivation

## Methods

Family-based Association Test

External Data

## Results

Example

Discussion

# Outline

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# Epistasis

- ▶ Epistasis: interaction between two or more genes
- ▶ known to be fundamental for the function of regulatory pathways in mammals
- ▶ implies its importance for the development of complex diseases such as cancer, Alzheimer's disease, diabetes

# Traditional Approaches

- ▶ for yeast and worms large scale double knock-outs and knock-downs exist
- ▶ linkage and association studies in mammals concentrate on either single locus associations or interactions between few preselected loci
- ▶ major reasons: non-availability of large and suitable data for analysis of interaction effects, low power of the studies

# Genome-Wide Screen in Mammals

- ▶ recent advances in biotechnology allow genome-wide genotyping of thousands of individuals  
→ can be used to **study epistatic effects over whole genome**
- ▶ genotyped individuals possibly related  
→ take population structure into account; even **make use of known relationships**

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## Method

- ▶ idea: two markers whose genotypes are correlated are likely to interact
- ▶ measure association via  $\chi^2$ -test for contingency table

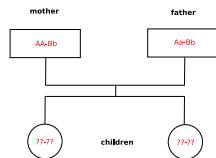
|    | BB         | Bb         | bb         |
|----|------------|------------|------------|
| AA | $n_{AABB}$ | $n_{AABb}$ | $n_{AAbb}$ |
| Aa | $n_{AaBB}$ | $n_{AaBb}$ | $n_{Aabb}$ |
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## Method

- ▶ idea: two markers whose genotypes are correlated are likely to interact
- ▶ measure association via  $\chi^2$ -test for contingency table
- ▶ make use of family information to avoid spurious findings: compare observed allele combination with what could have been inherited from parents
- ▶ additional correction for allelic drift

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(example: 10,000 markers:  $\sim 10^8$  interactions)
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- ▶ need to complement with additional, **external information**



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# Databases

- ▶ use public knowledge about gene  $\times$  gene interactions to confirm results;  
e.g. STRING: database of known and predicted physical and functional interactions
- ▶ include information from regulatory pathways

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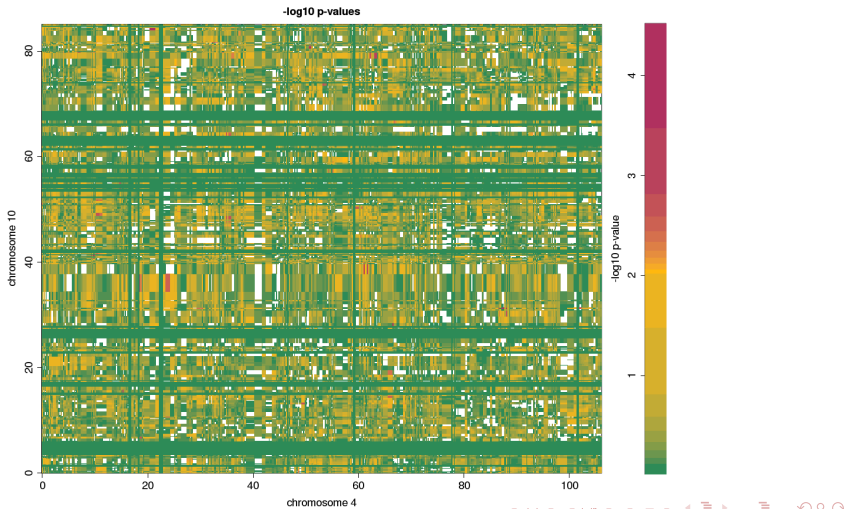


# Data

- ▶ Solberg, L.C. et al. (2006). A protocol for high-throughput phenotyping, suitable for quantitative trait analysis in mice. *Mammalian Genome*, **17**, 129-146.
- ▶ genotype data from more than 2000 outbred mice consisting of  $\sim 12,000$  markers
- ▶ only consider interactions on two different chromosomes

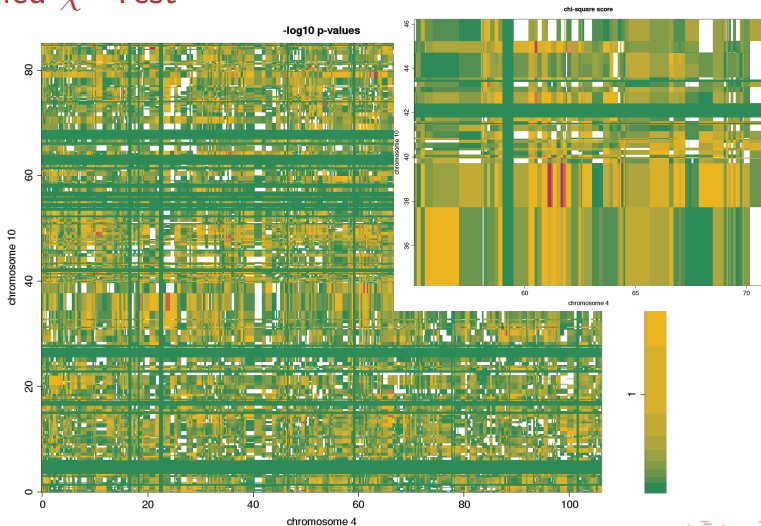


# Modified $\chi^2$ -Test





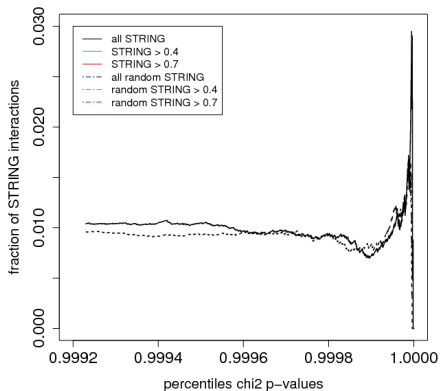
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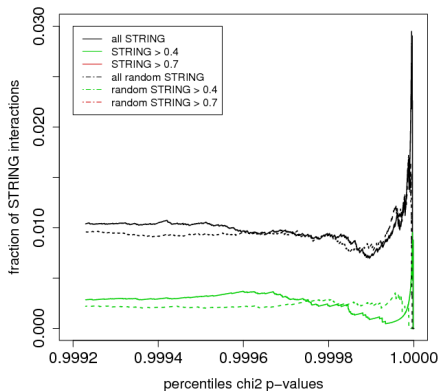
## Confirmation with STRING

- ▶ fraction of SNP pairs with a low  $\chi^2$  p-value that lie close to interacting genes
- ▶ proportion of confirmed interactions should increase with increasing  $\chi^2$  score



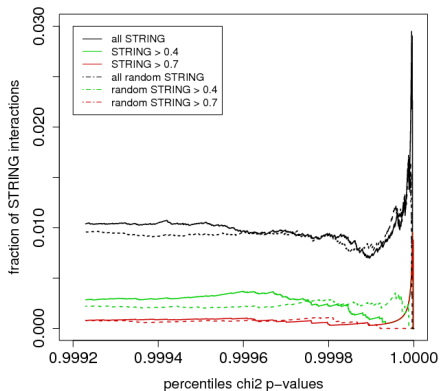
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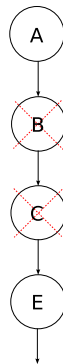
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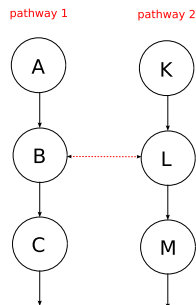
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- ▶ interactions in one pathway can be crucial, e.g. when signal weakened by two consecutive dysfunctional pathway members



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- ▶ interactions between pathways indicate common endpoint



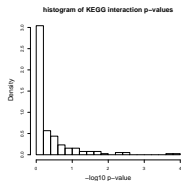
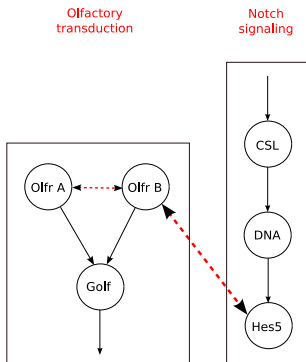


## Example: KEGG Pathway

KEGG: database of signaling and metabolic pathways

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indicates importance of olfactory receptors in embryonic development and interplay with notch

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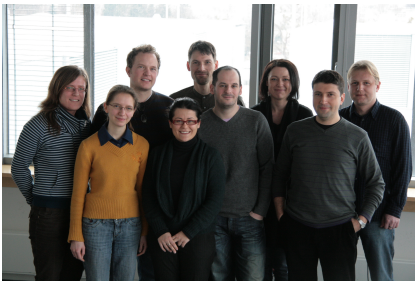
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- ▶ works on a genome-wide scale
- ▶ population structure explicitly taken into account
- ▶ other confounding factors readily included
- ▶ data integration from different sources increases power and facilitates biological interpretation of results



# Acknowledgements



- ▶ Dr. Andreas Beyer
- ▶ my colleagues in the Cellular Networks and Systems Biology group

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