

ESHG workshop (Barcelona 2016)



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Integrate network resources to optimize genetic association studies

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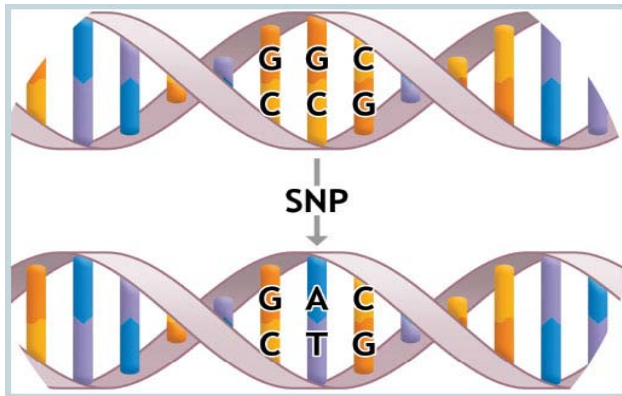
Outline

- 1. GWAS**
- 2. Integrate GWAS with gene network**
- 3. Practical issue**
- 4. A successful example**
- 5. Conclusion**

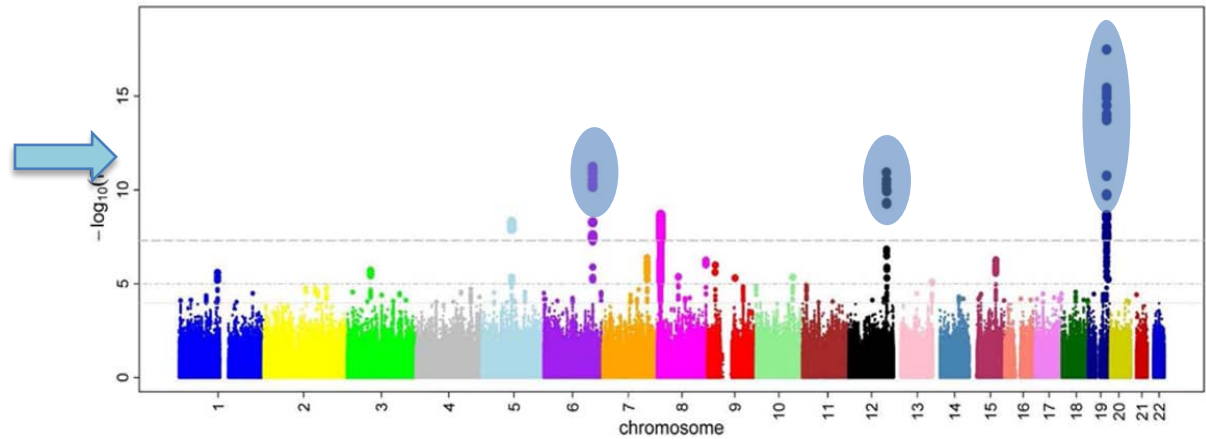
1. GWAS

□ Genome-wide association study

- Association test for each SNP(single nucleotide polymorphisms)



P-value threshold: 5×10^{-8}



2. GWAS + gene network

- **Why GWAS + network knowledge**

2. GWAS + gene network

□ **Why GWAS + network knowledge**

- GWASs Identified small amount of loci
→ explain a small part of genetic risk
- GWASs focused on single SNPs analysis
→ underpowered to detect SNPs with small marginal effect

2. GWAS + gene network

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Need to analyze multiple genetic variants jointly

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Network-based analysis:

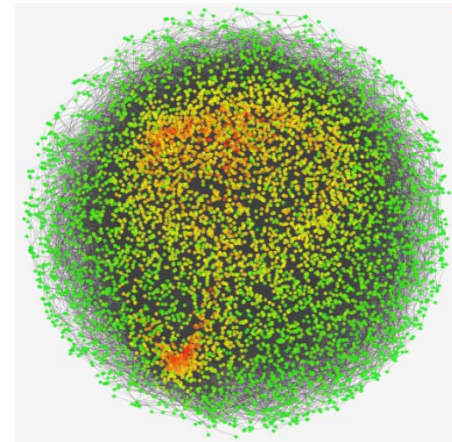
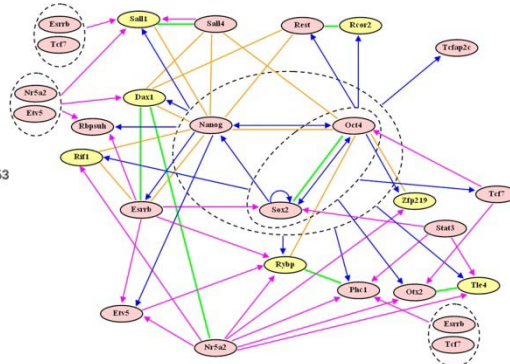
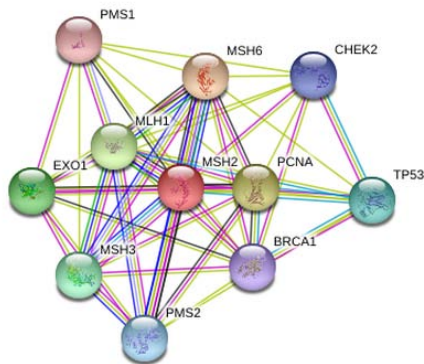
**GWAS results + Gene Network
= Disease-associated gene module**

2. GWAS + gene network

□ Abundance of network resources

Either experiment verified or computationally-predicted

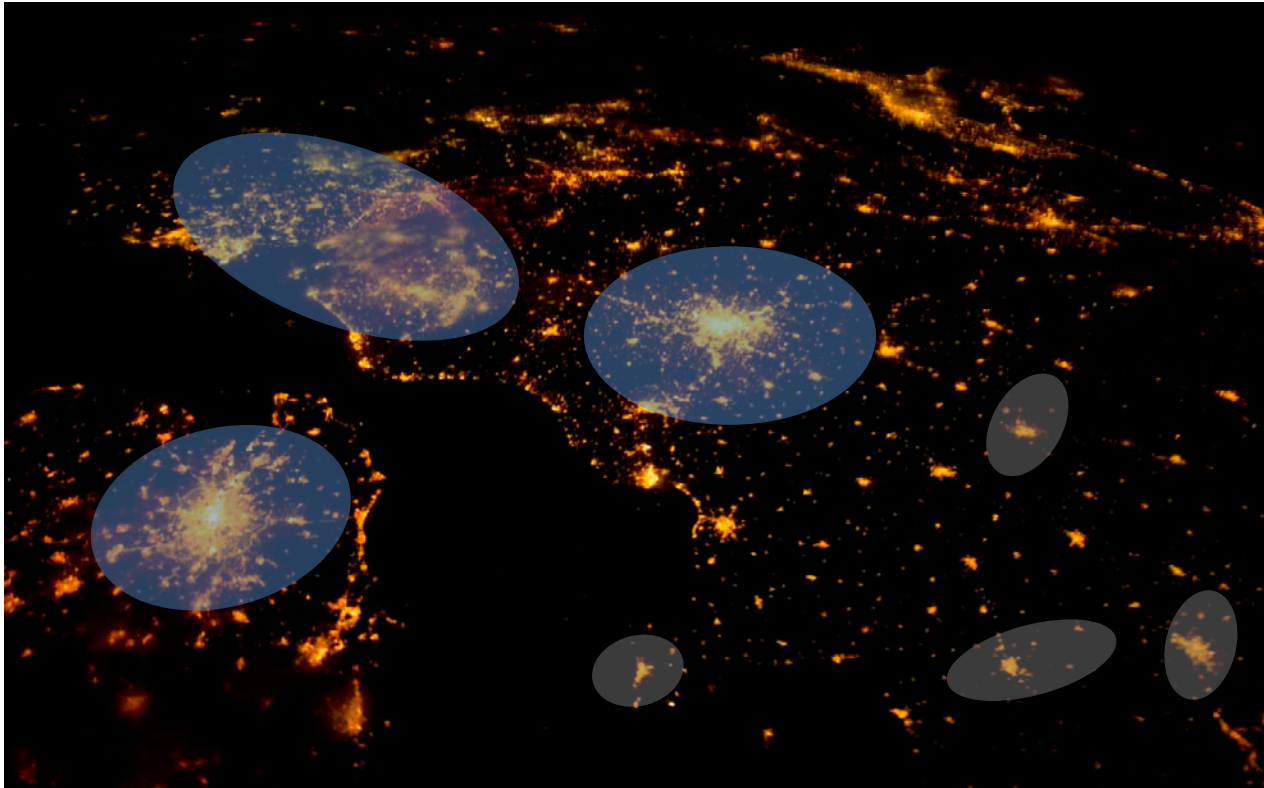
- Gene regulatory network
- Protein (gene product) interaction network
- Gene co-expression network
-



2. GWAS + gene network

□ General objective

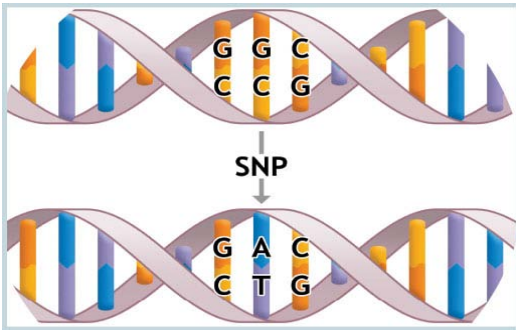
- Find gene modules (connected genes) enriched in high association signals



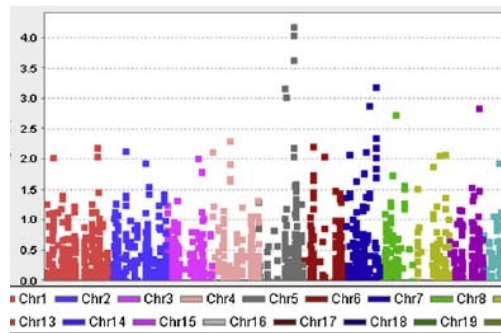
2. GWAS + gene network

□ Workflow

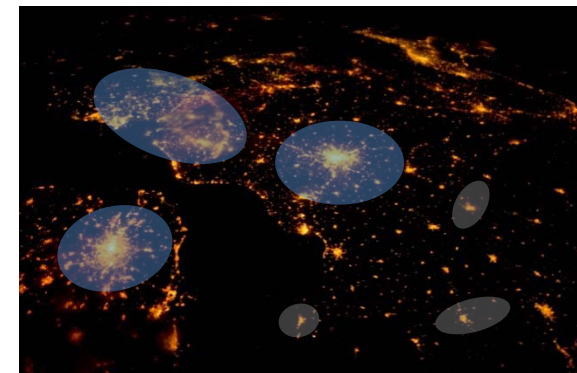
Genome-wide
raw data



Gene-disease
association signal



Signal weighted GeneNet

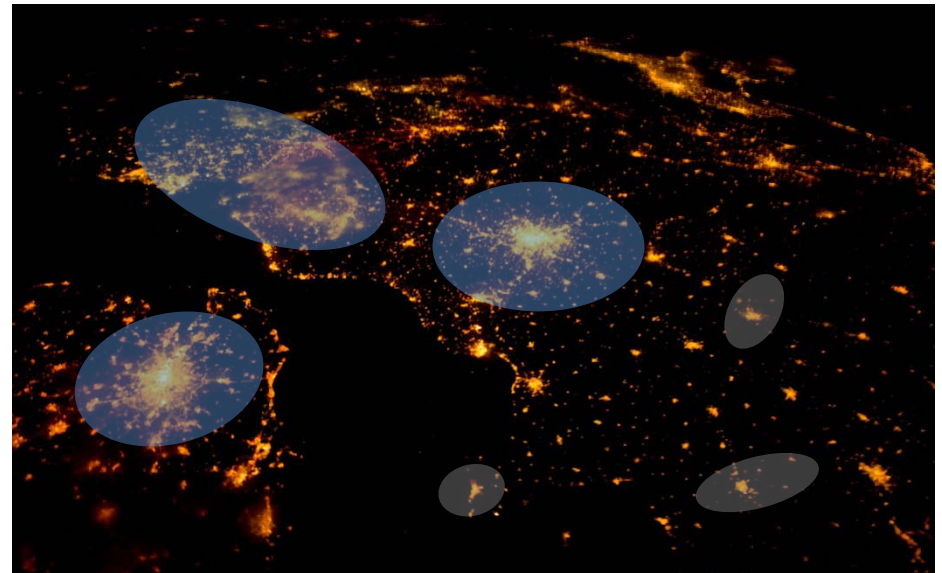


Gene modules

3. Practical issues

□ Practical issues

- **Big Data:** genome-wide study ~1M SNPs, 20K genes
- **Noise** in association signals / network information
- Method **efficiency**
- Result **replicability** / **Interpretability**
-



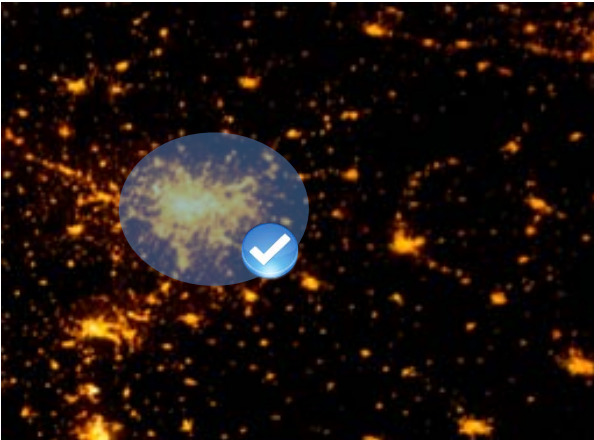
4. A successful example

- **Find gene module associated with asthma**
 - asthma is an inflammatory disease of the airways of the lungs
 - caused by genetic + environmental factors



4. A successful example

- ❑ **Specific objective: find gene module:**
 - Enriched in high signals
 - **Strongly** interconnected



4. A successful example

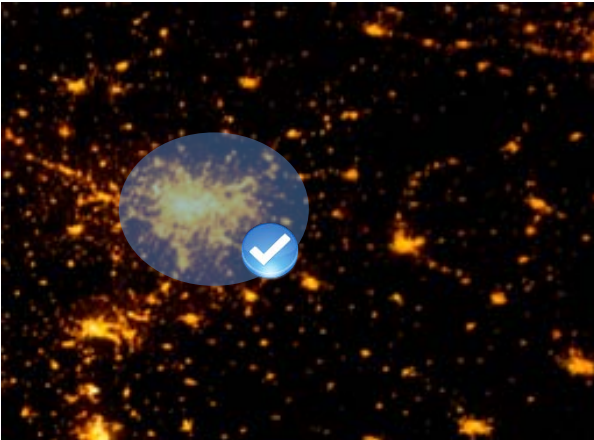
□ Specific objective: find gene module:

- Enriched in high signals

- **Strongly** interconnected →

- ✓ Robust to noise → high replicability

- ✓ Functional closely related → high interpretability



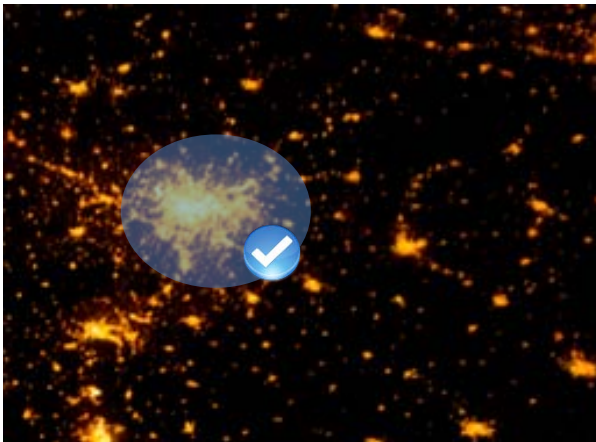
4. A successful example

□ Specific objective: find gene module:

- Enriched in high signals

- **Strongly** interconnected

- ✓ Robust to noise → high replicability
- ✓ Functional closely related → high interpretability



Method: select module by solving maximizing:

$$g(u) = z^T u + \lambda u^T A u - \eta \|u\|_0; \lambda \geq 0, \eta \geq 0$$

Encourage high signal Encourage interconnection Encourage sparsity

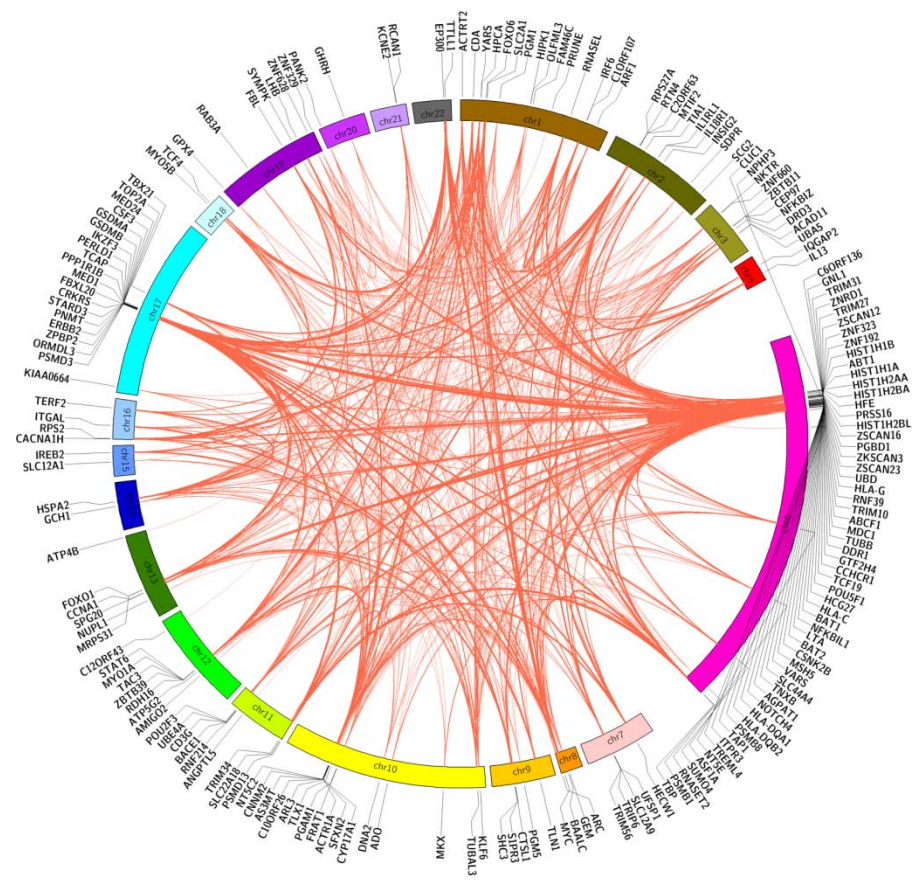
4. A successful example

Result: identified a gene module associated with asthma

- ✓ Strongly interconnected
- ✓ Enriched in high/consistent signals
- ✓ Enriched in known asthma genes
- ✓ Genes function closely related
- ✓ Functional relevance to asthma



- **190 genes / 1250 connections**
- **All genes** have P-value < 0.05
- Include **7 reported loci** of asthma
- **9 functionally related** gene groups
- **15 KEGG pathways** related to asthma



Conclusion

- Single marker-based test is underpowered to detect moderate effect markers
- Network-based analysis helps analyze genetic variants jointly and systematically
- **Robustness, efficiency, interpretability** are important aspects to be considered by network-based analysis

Acknowledgements

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Thanks!