

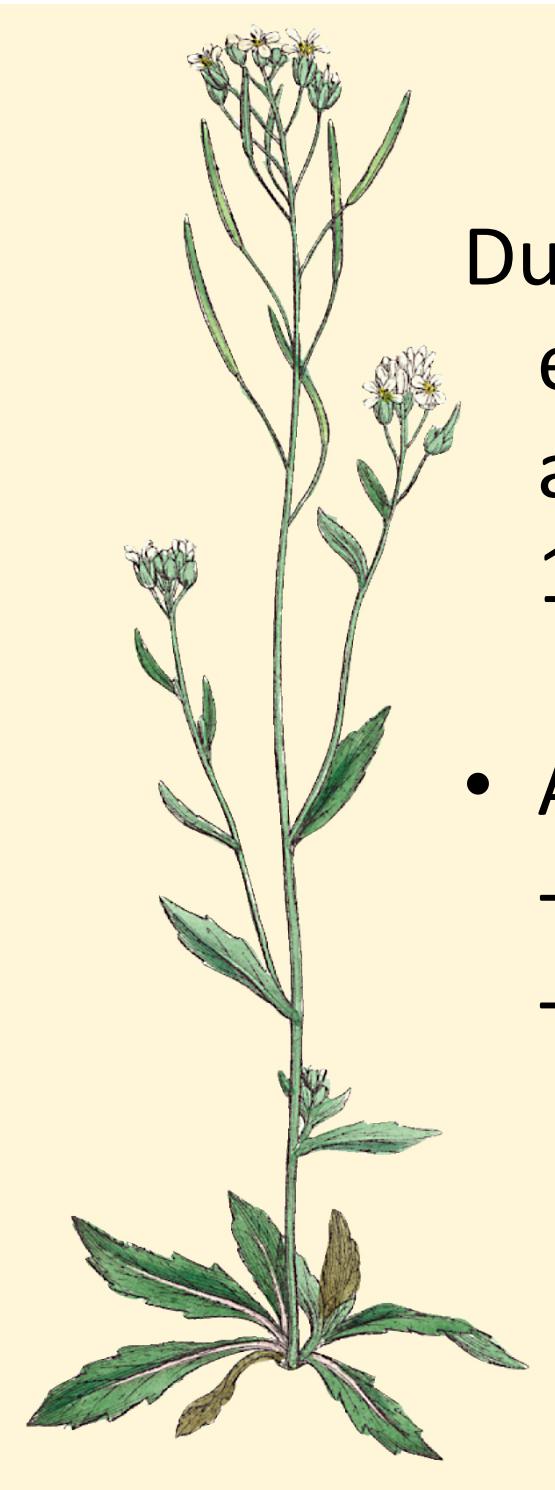
What can biological networks tell us about the fate of duplicate genes in *Arabidopsis thaliana*?

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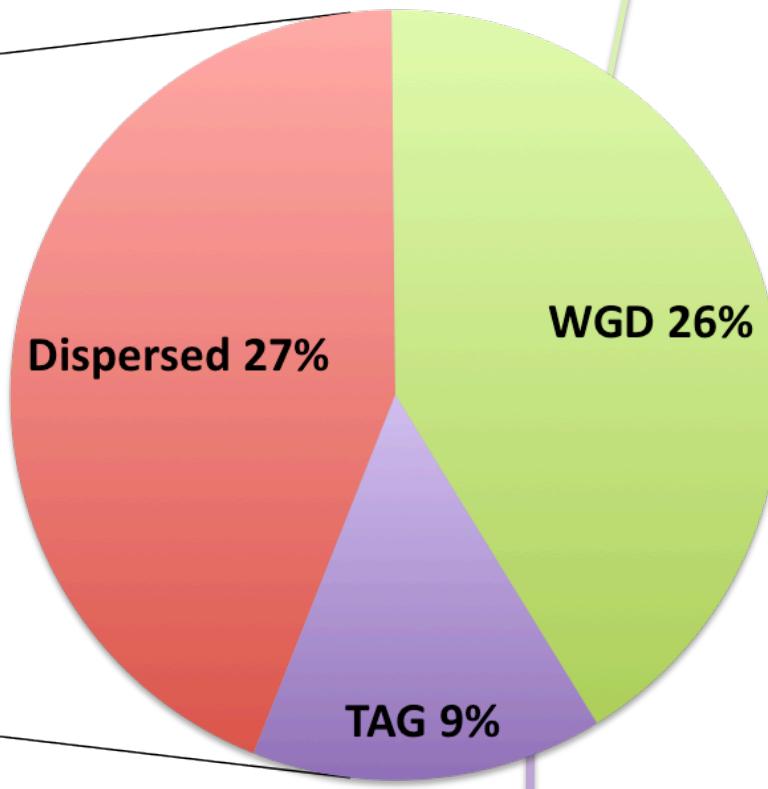
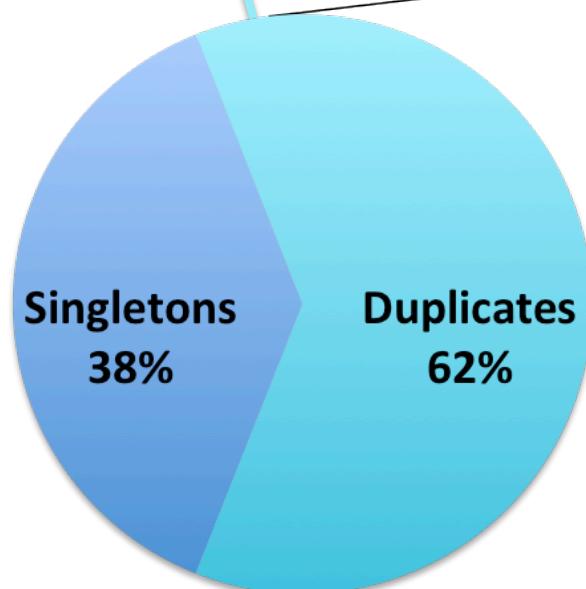
Duplicated genes are very important in evolution as they are implied in the appearance of new functions (Ohno, 1970).

- Analysis of expression data:
 - using standard methods
 - using inferred networks

Arabidopsis thaliana

Types of Duplication

The 16977 duplicate genes are clustered into 5253 families.



Whole Genome Duplicates are defined from the literature (Bowers 2003).

Tandemly Arrayed Genes



Expression Data

Using an Affymetrix array

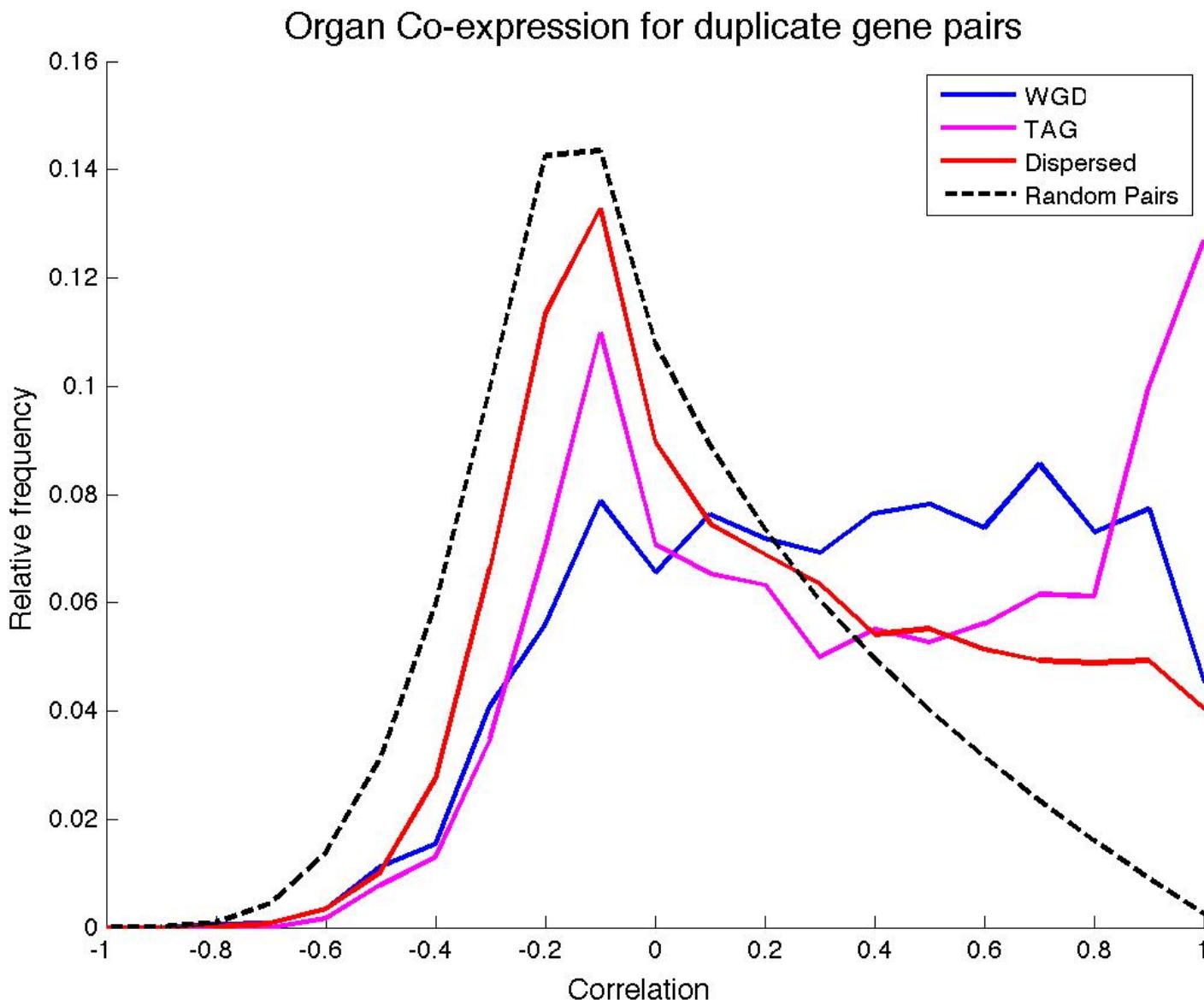
Left with **26 601 genes** with gene expression under 33 different conditions, grouped in 3 categories: **Organ, Stress and Light.**



From Kousuke Hanada, Riken Plant Centre, Japan

Gene Co-expression

using Spearman's rank correlation coefficients



Inferring Networks from Gene Expression data

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APPLICATIONS NOTE

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Systems biology

SIMoNe: Statistical Inference for MOdular NEtworks

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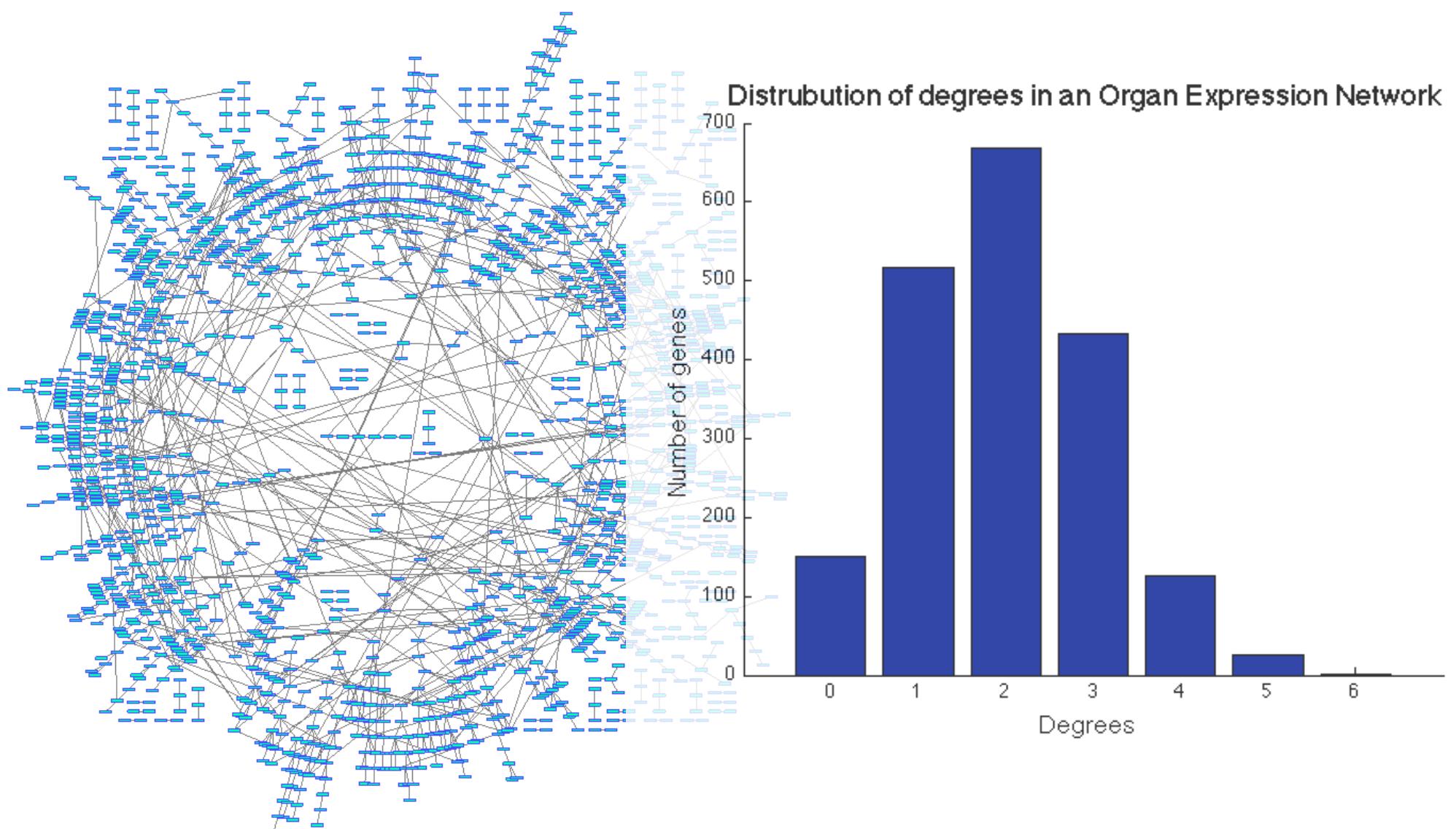
ABSTRACT

Summary: The R package SIMoNe (Statistical Inference for MOdular NEtworks) enables inference of gene-regulatory networks based on partial correlation coefficients from microarray experiments. Modelling gene expression data with a Gaussian graphical model (hereafter GGM), the algorithm estimates non-zero entries of the concentration matrix, in a sparse and possibly high-dimensional setting. Its originality lies in the fact that it searches for a latent modular structure to drive the inference procedure through adaptive penalization of the concentration matrix.

may be computed using the inverse covariance matrix (hereafter concentration matrix). Detecting non-zero entries in this matrix is in fact equivalent to reconstructing the graph of conditional dependencies.

Ideally, the concentration matrix can be estimated by inverting the empirical covariance matrix. However, in the high-dimensional setting the latter is not invertible. Moreover, such a procedure does not lead to a sparse estimate, whereas biological evidence advocates for sparse networks. Sparsity means that the concentration matrix has a large number of zero entries. In this context, several estimation

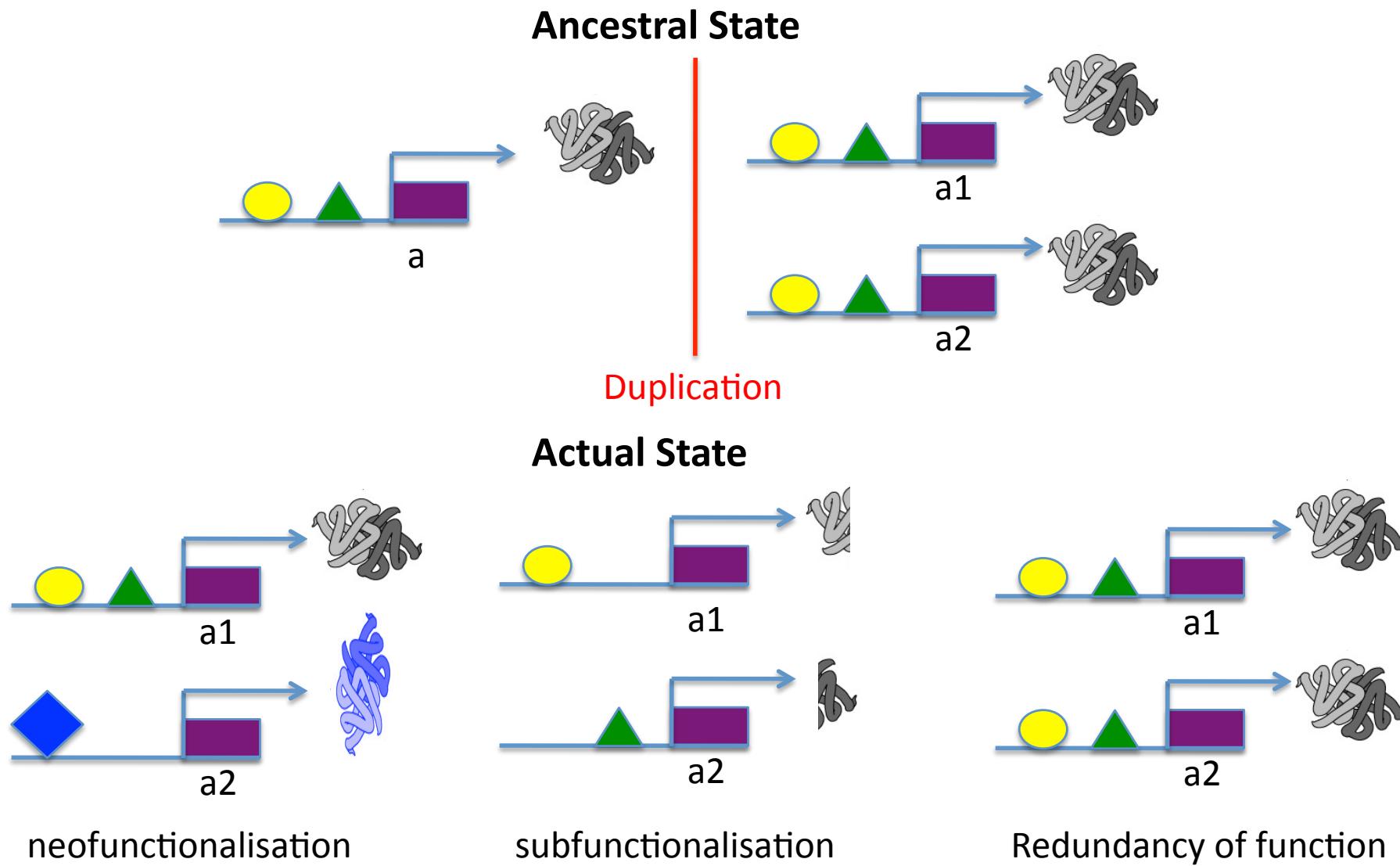
The Inferred Network



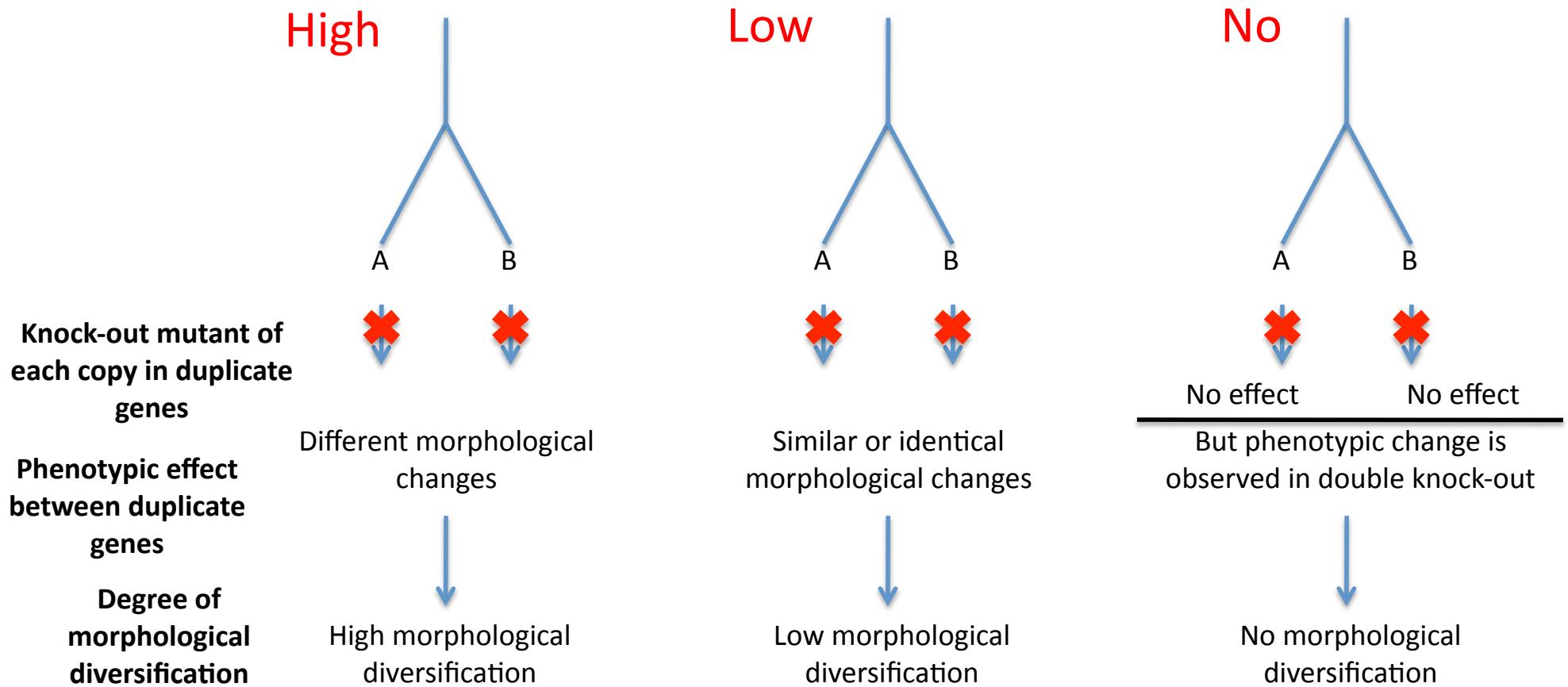
A Sparse network

Metric for measurement: shortest path between genes.

How duplicated genes are maintained

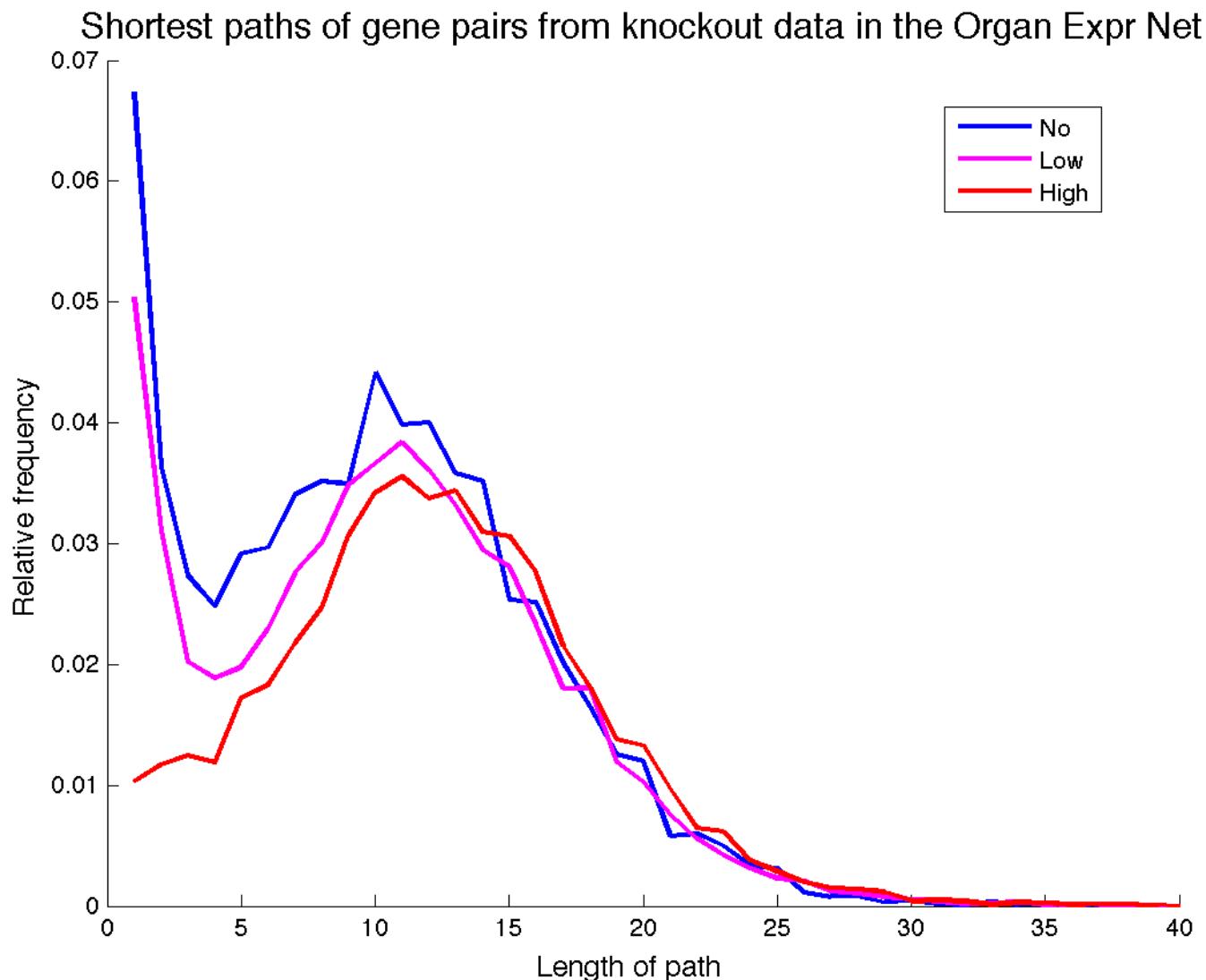


Morphological Knockout Data

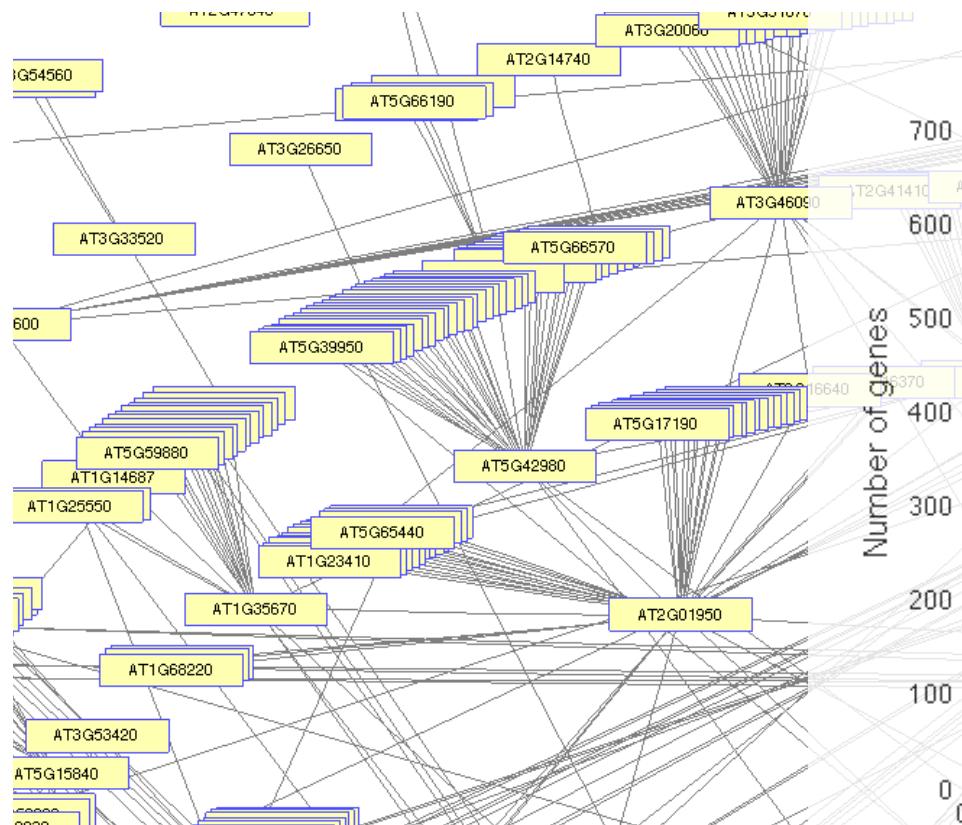


Increased expression and protein divergence in duplicate genes is associated with morphological diversification
Hanada K. et al (2009) PLOS

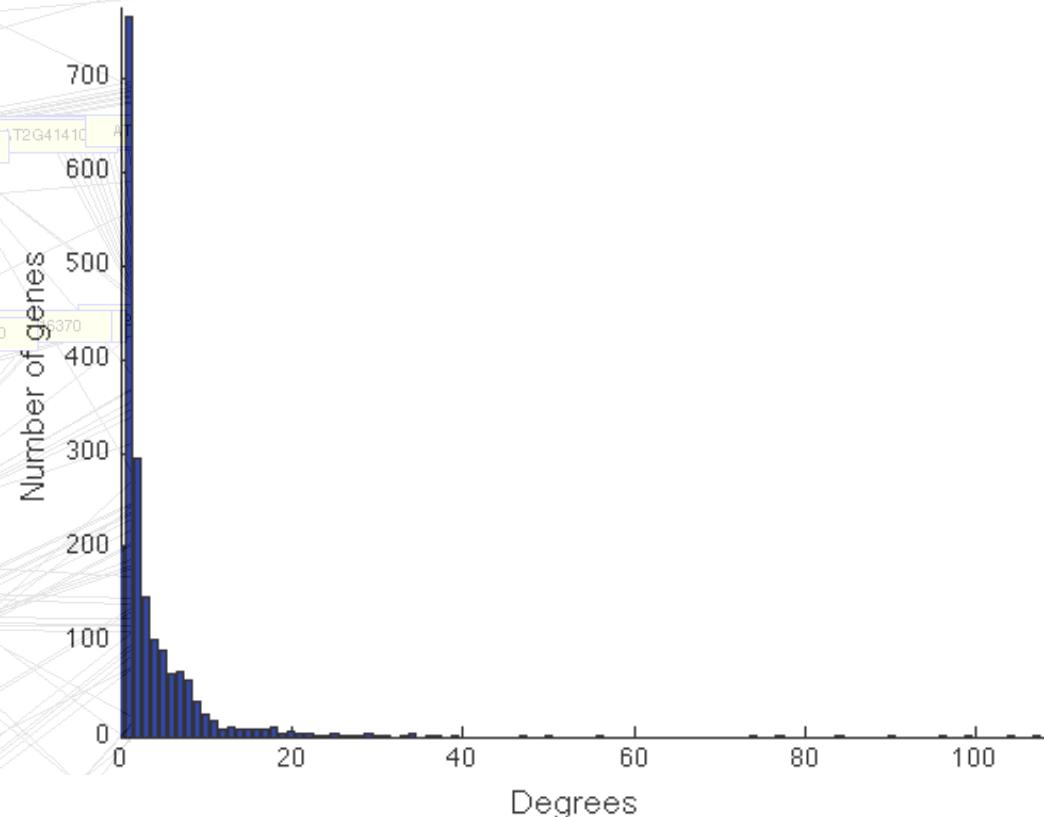
Network inferred for morphological knockout data



Protein-protein interaction networks



Distrubution of degrees in the PPI Network



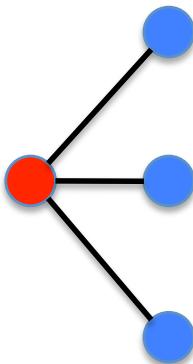
PPI Network:
Scale Free

Metric used: Jaccard index = shared_neighbours/all_neighbours

2946 genes
Of which we
have
expression
data for: 2749

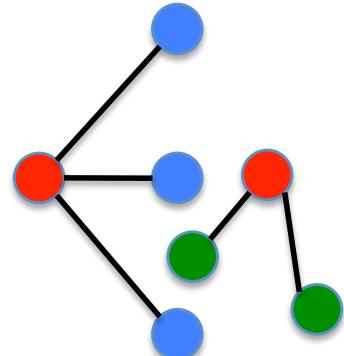
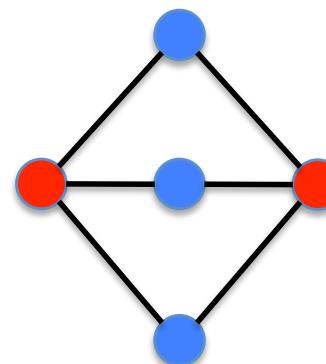
Duplication in PPI networks

Ancestral State



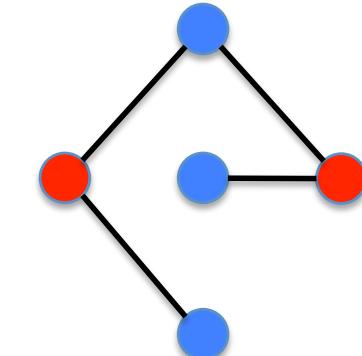
Duplication event

Actual State



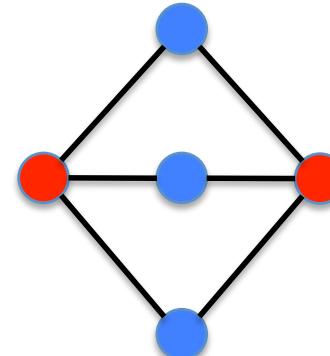
neofunctionalisation

Jaccard = 0



subfunctionalisation

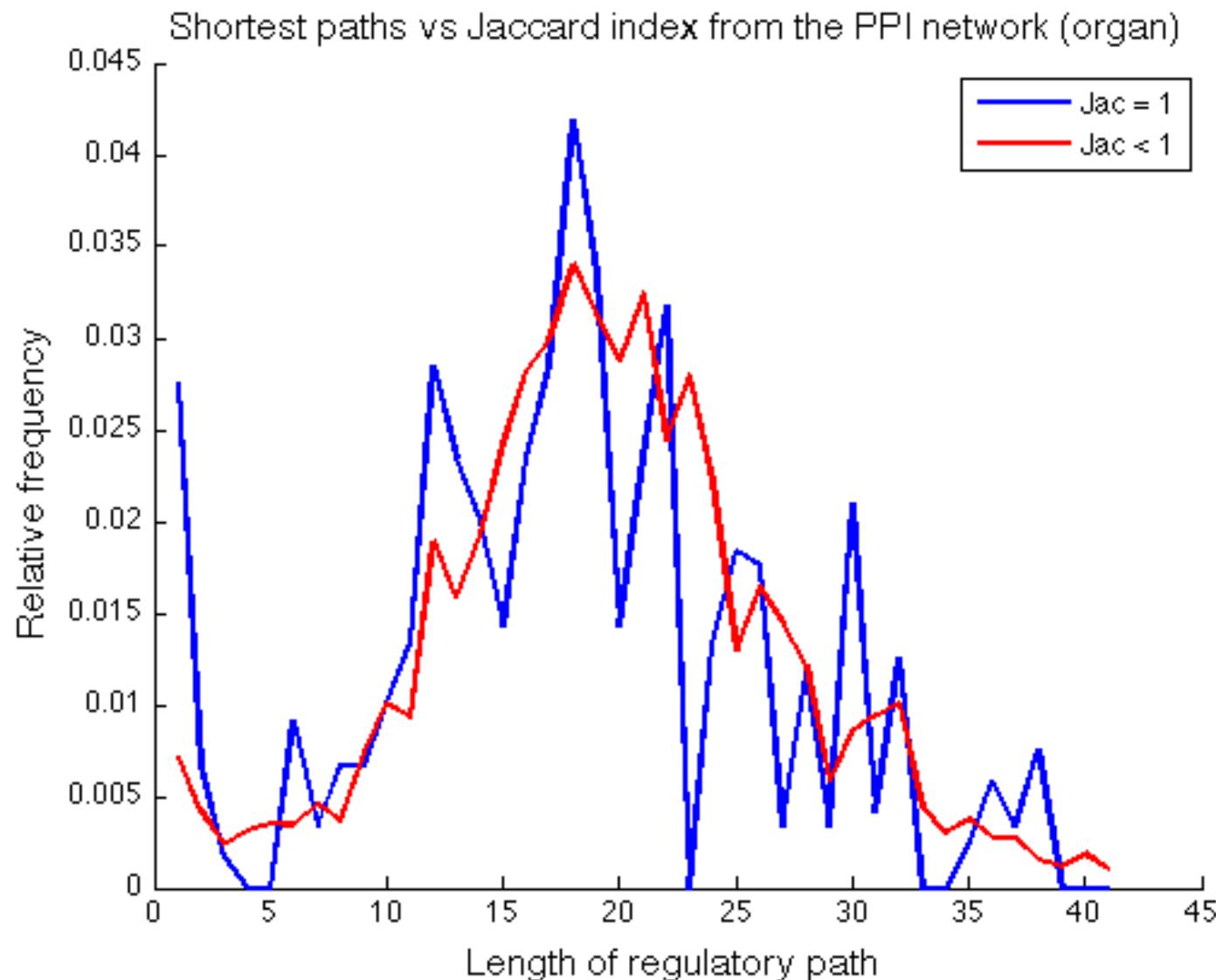
Jaccard = 1/3



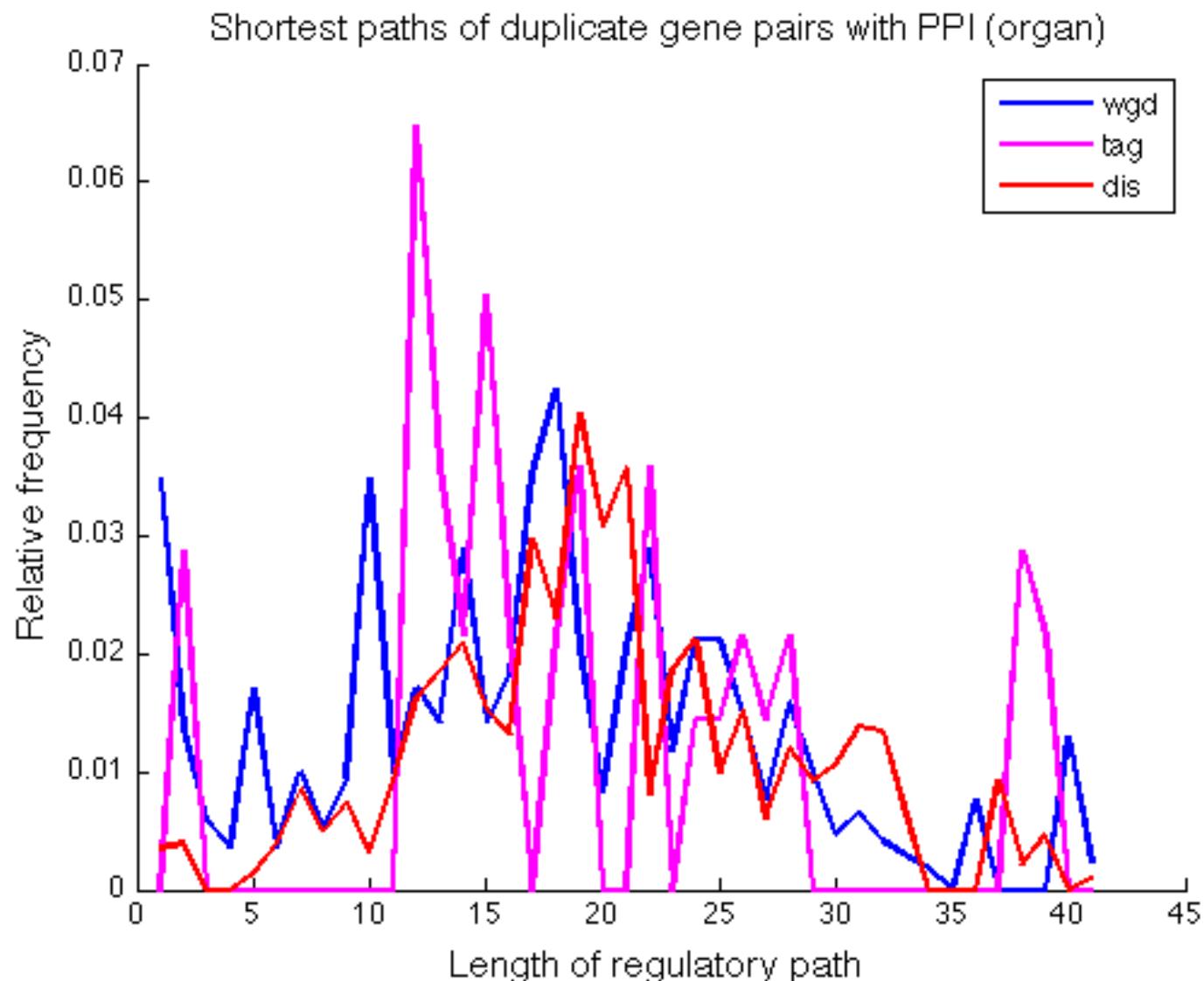
Redundancy of function

Jaccard = 1

Organ expression network analysed for Jaccard index in the PPI



Organ expression network analysed for type of duplication





Discussion

- Biological Results – Type of Duplication matters.
- Inferred networks provide us more detailed analysis than standard methods.

Further research

- How do genes within families relate to each other on the inferred network?

Arabidopsis thaliana

Thank you!



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Defining Gene Families

Finding homologous genes:

BLASTP search found 978 880 gene pairs (**23 386 genes**)

Sequence A



Sequence B

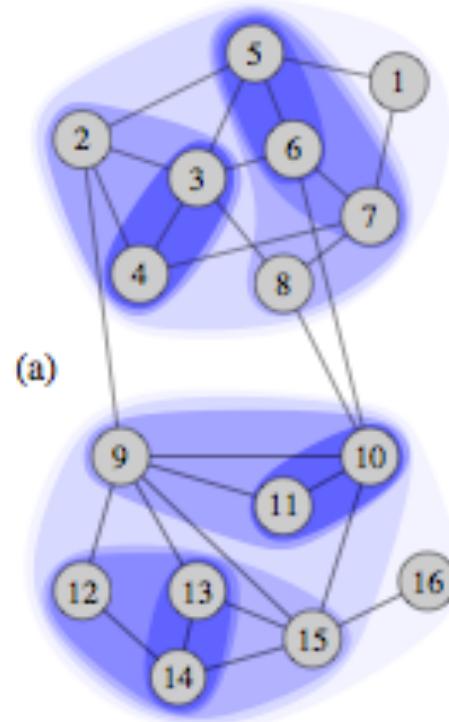


Sequence C



Defining Gene Families

Clustering with the Walktrap Algorithm



From:

Fig 1a in Computing Communities in Large Networks Using Random Walks

Pascal Pons and Matthieu Latapy

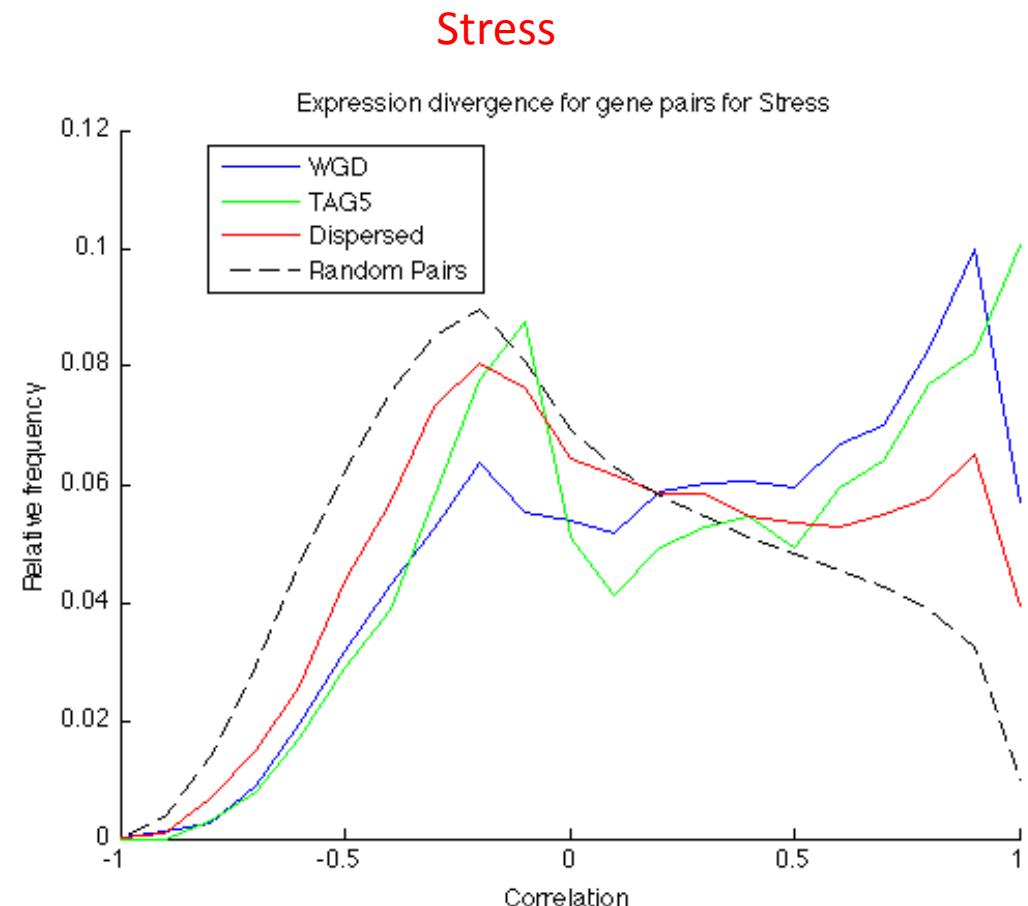
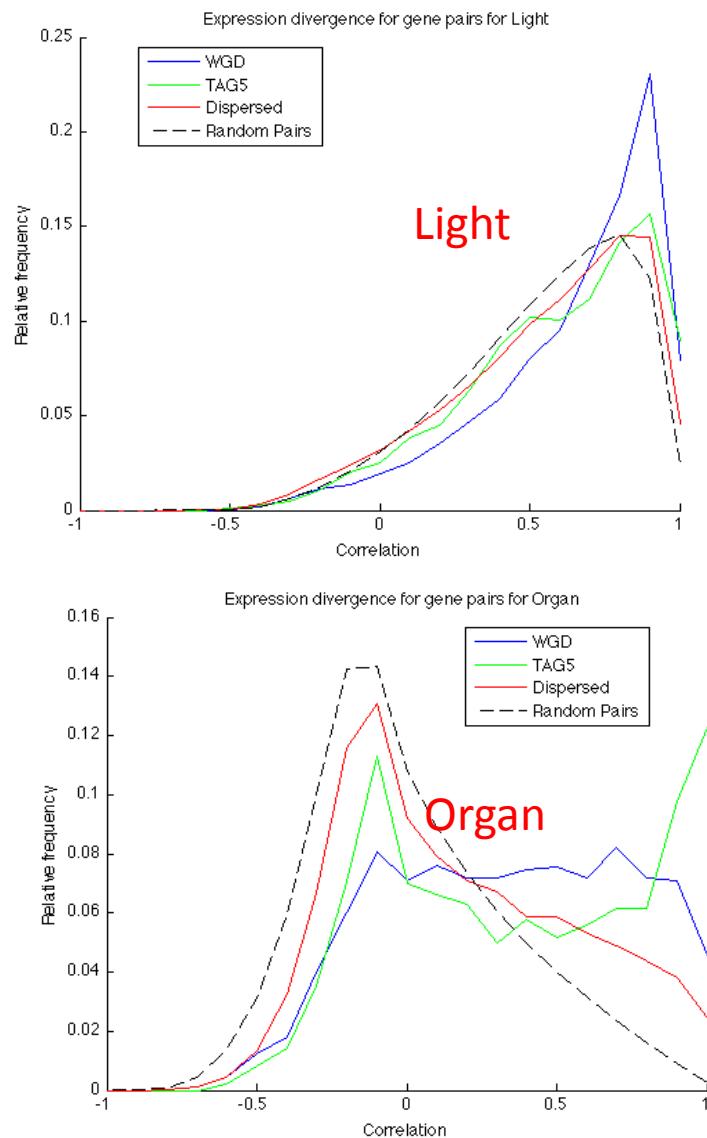
Computer and Information Sciences - ISCIS 2005

Lecture Notes in Computer Science, 2005, Volume 3733/2005, 284-293

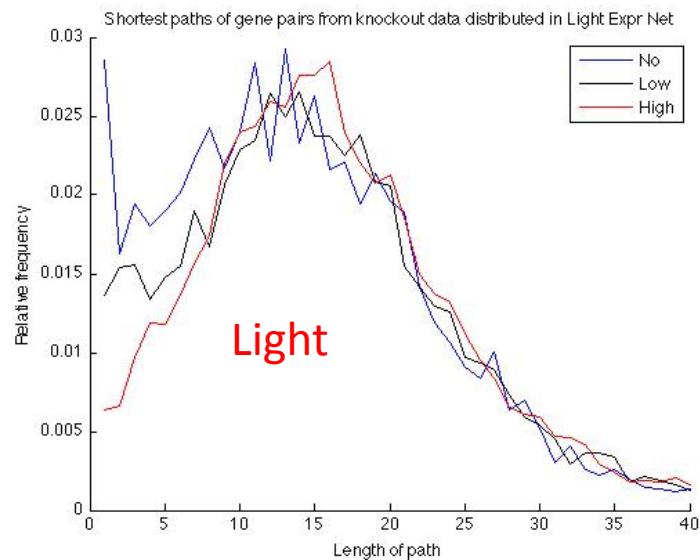
Clustering methods compared

	Walktrap 30	Walktrap 50	Walktrap 70	MCL	Ensembl
No. of duplicate genes	17085	16977	16619	17693	9438
No. of families	5092	5253	5690	6598	4116
Maximum family size	61	61	48	194	12
Families that are the same as a Walktrap 30		4949	4574	798	0
Families that contain a Walktrap 30 family		0	0	4908	1852
Families that are contained by Walktrap 30 family		302	1109	1587	277
Families that have no match with a Walktrap 30 family		2	7	1145	3271

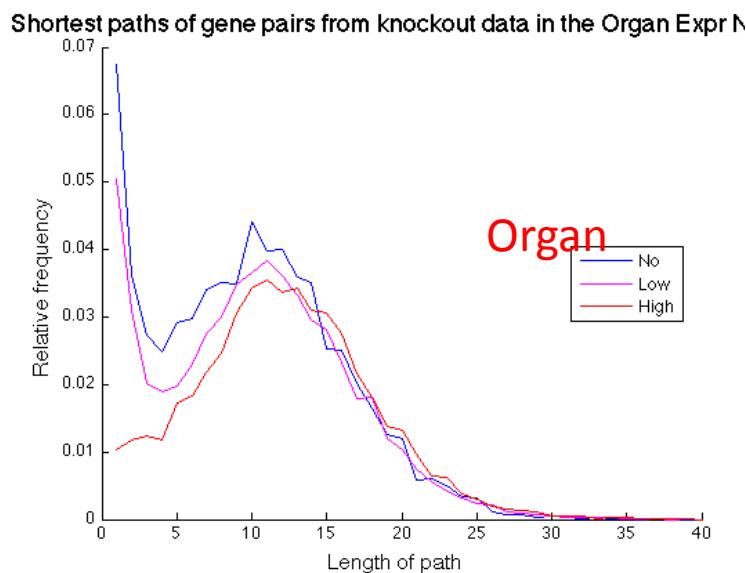
Gene Co-expression for all duplicates



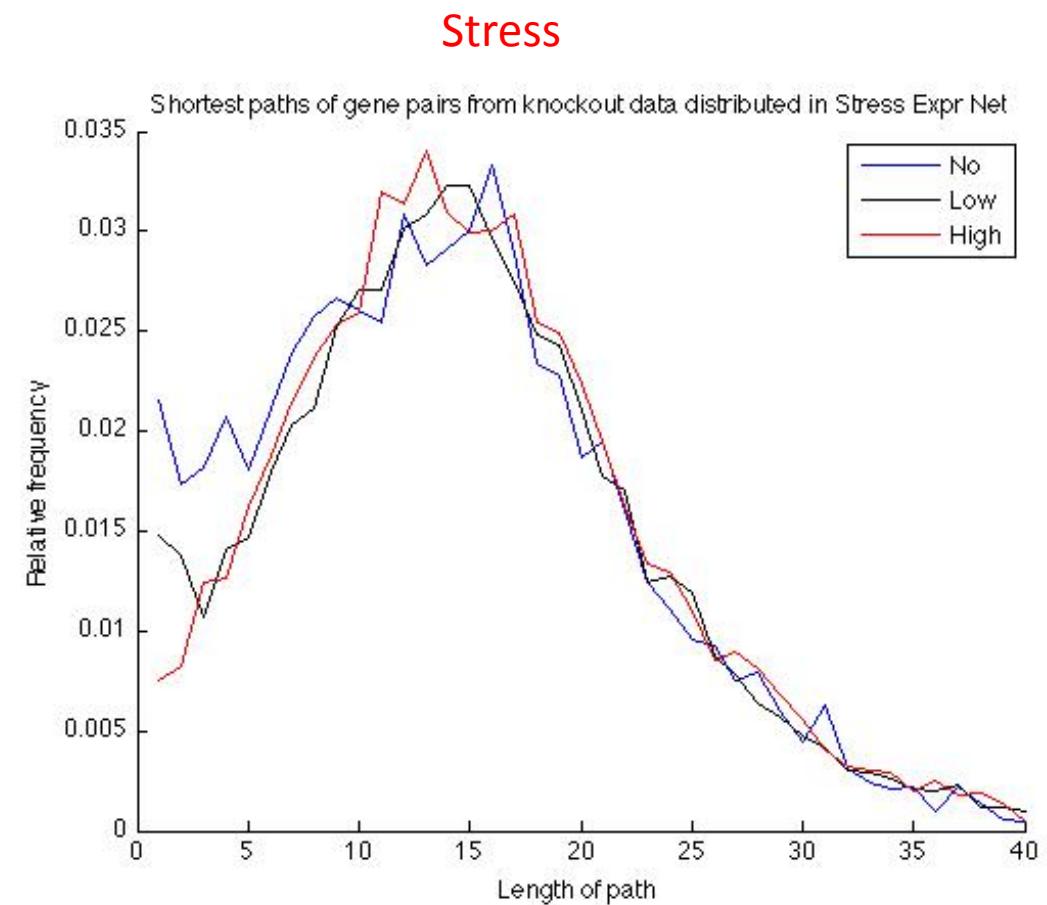
Gene Regulatory networks for Morphological Knockout data



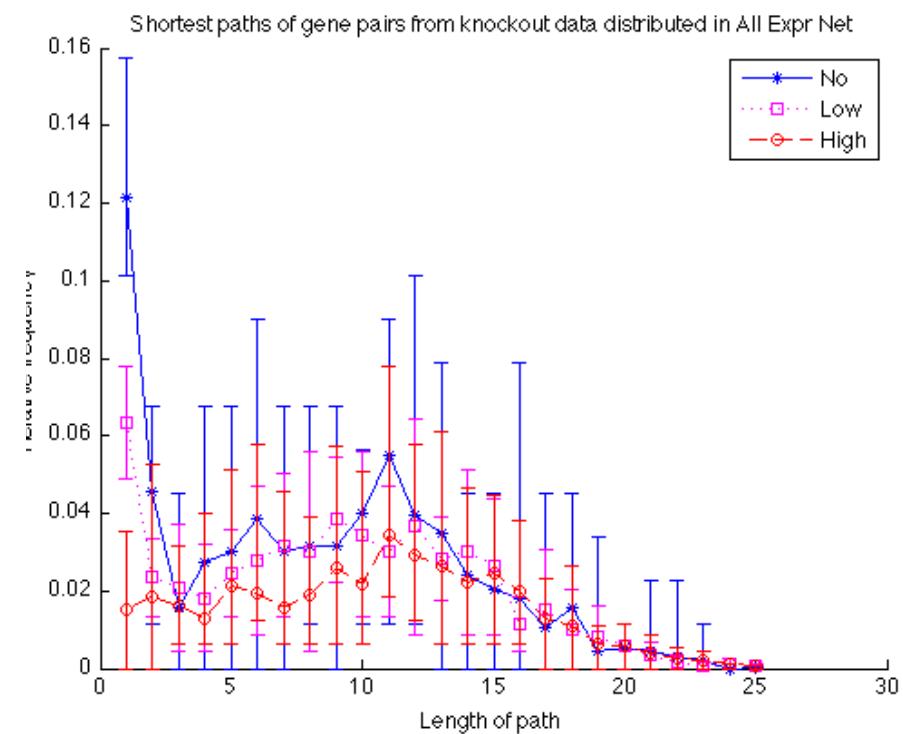
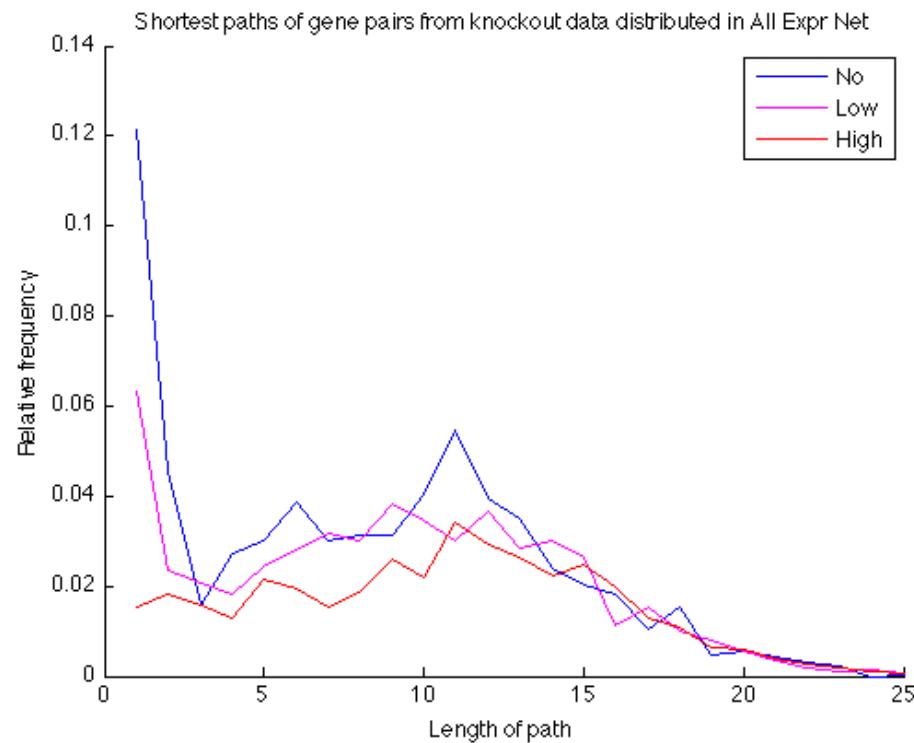
Light



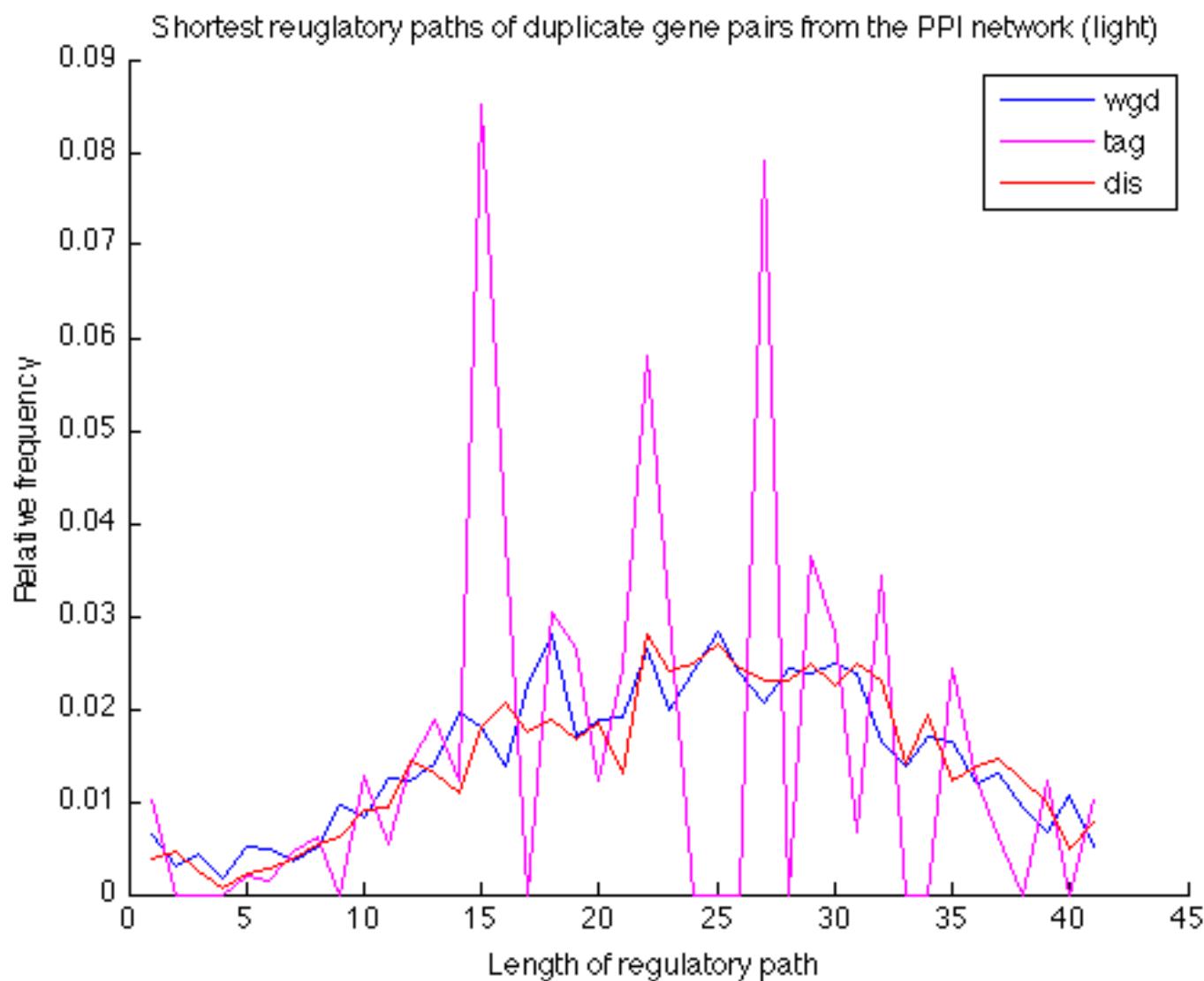
Organ



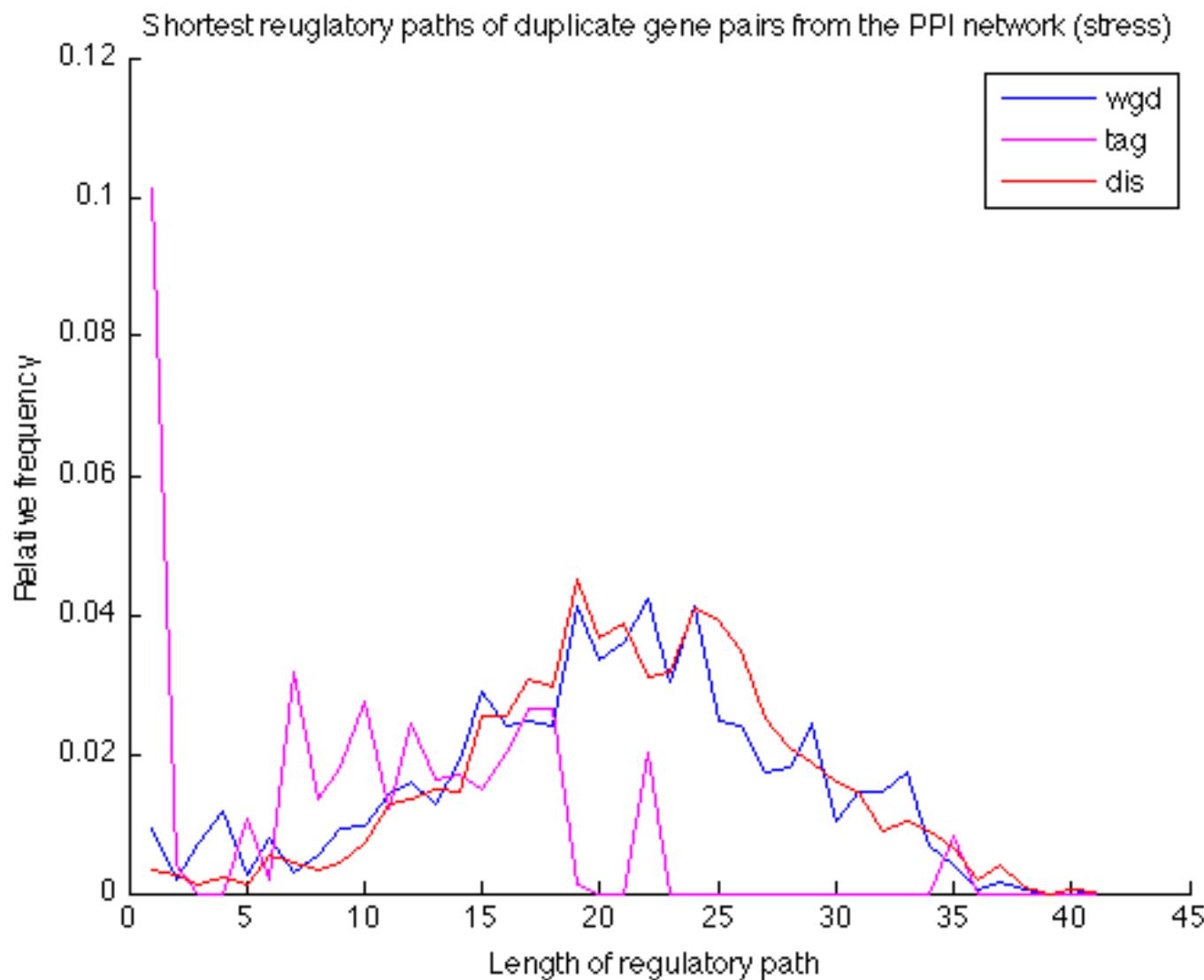
Gene Regulatory networks for Morphological Knockout data



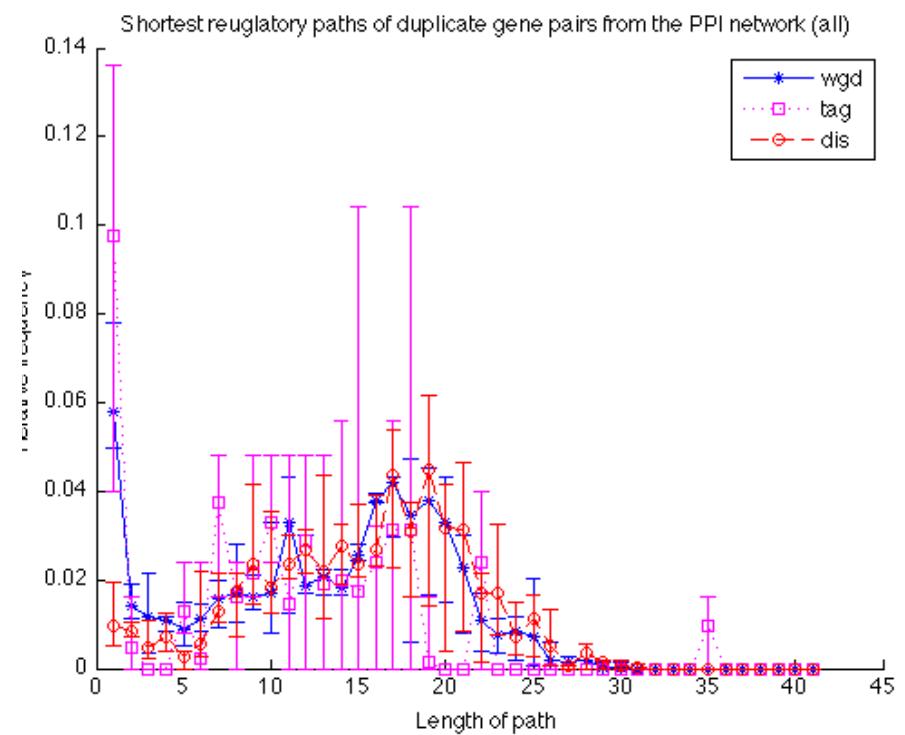
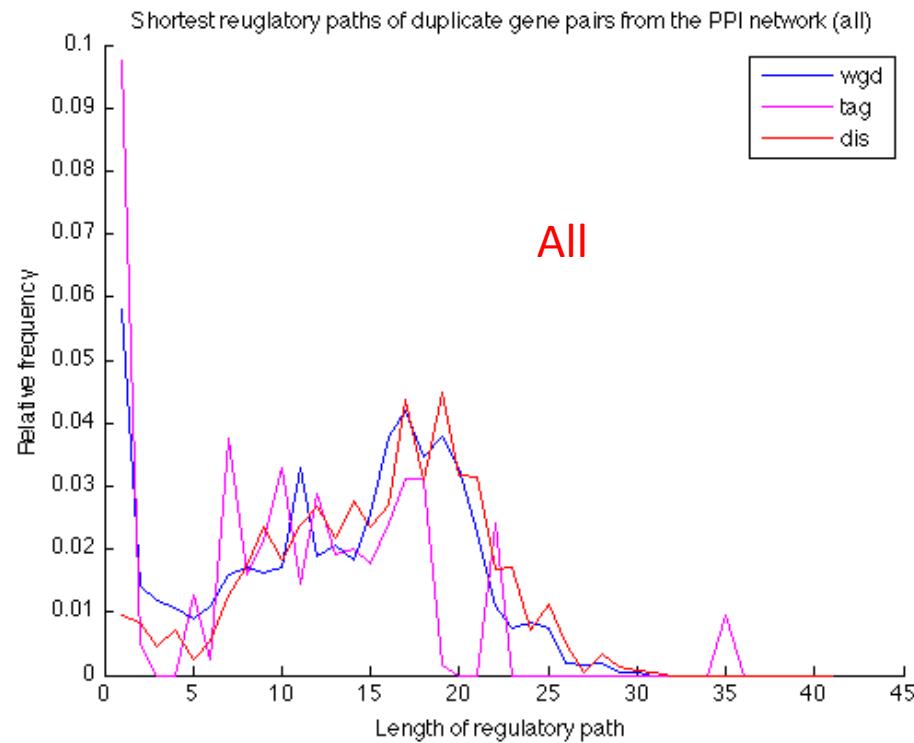
Light co-regulation for PPI genes



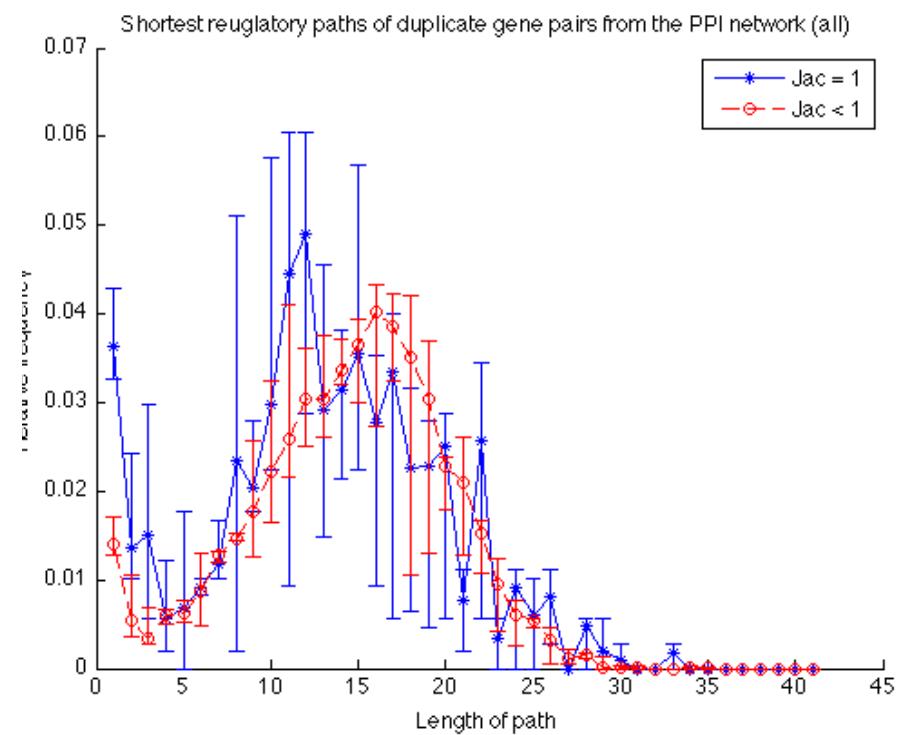
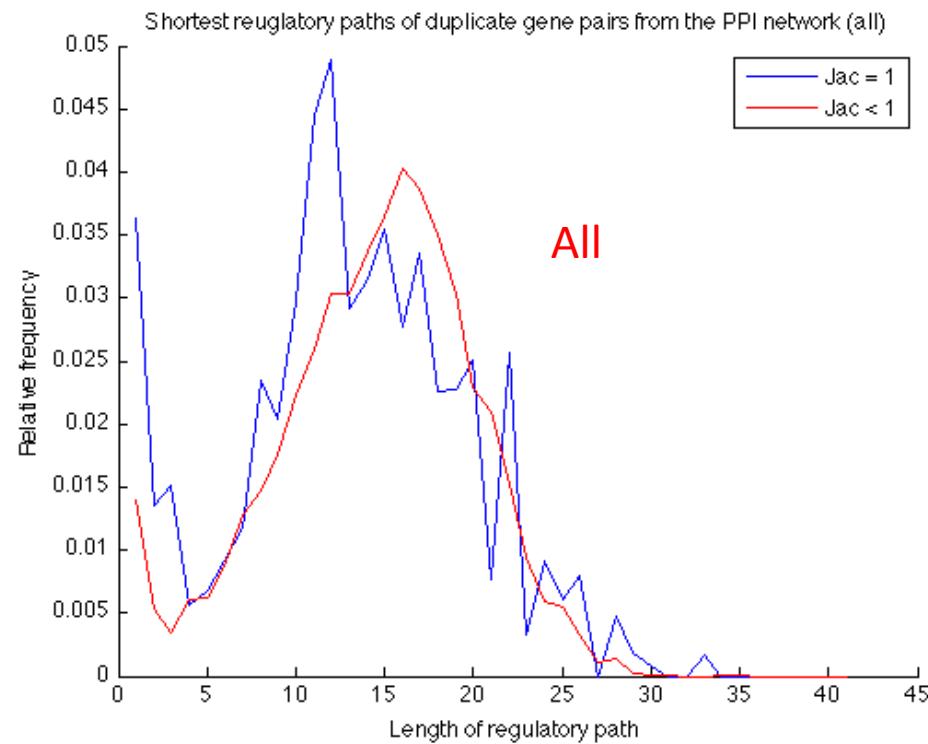
Stress co-regulation for PPI genes



Gene Regulatory networks for duplicates in the PPI

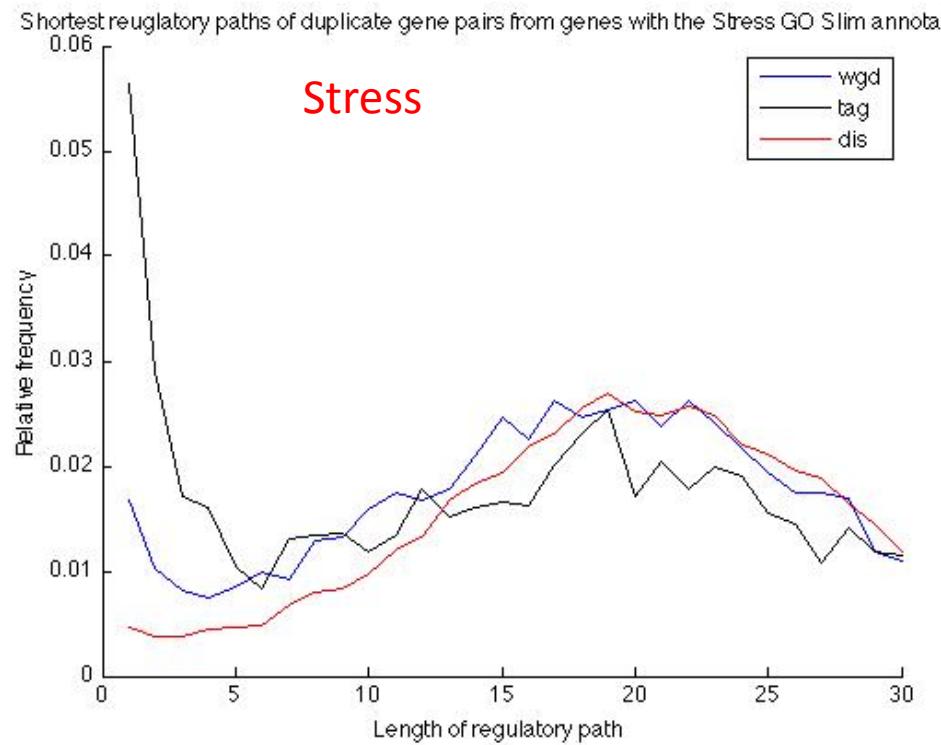


Jaccard index (PPI) vs. Shortest Path (Gene Regulatory Network)



Gene Regulatory networks found using:

Stress GO slim



Response to Light Stimuli GO annotation

