### (Supplementary material) An improved method for scoring protein-protein interactions using semantic similarity within the Gene Ontology

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### Supplementary figure S1 - Mutually exclusive sub-graphs

(a) Transitive reduction - suppose a, b, c, and d are the nodes in graph G with directed edges as shown in figure (a). Let the number of genes annotated to each node is 1. Then the total annotation of node a in G (annotation of a and its descendants) is 4. Transitive reduction of G will result in G' without edge  $d \rightarrow a$  and with same total annotation of 4 as a can still be reached from d. (b) Replication - suppose term d is common to both the sub-graphs B and C then term d will be copied to both the sub-graphs.



### Supplementary figure S2 - ROC curves for S. cerevisiae PPI dataset (IEA-)

ROC evaluations of semantic similarity measures at different cutoffs based on the *S. cerevisiae* PPI dataset derived from DIP are shown. The evaluation was performed using cellular component, biological process, molecular function ontologies of GO. Best-match average (bma) approach for combining multiple annotations was used on dataset without (IEA–) electronic annotations. TCSS & Resnik show best ROC profiles for all three ontologies.



### Supplementary figure S3 - F-score curves for *S. cerevisiae* PPI dataset (IEA-)

 $F_1$  score (harmonic mean of precision and recall) evaluations of TCSS and Resnik semantic similarity measures at different cutoffs based on the *S. cerevisiae* PPI dataset derived from DIP are shown. The evaluation was performed using cellular component, biological process, molecular function ontology of GO. Best-match average (bma) approach for combining multiple annotations was used on dataset without (IEA–) electronic annotations.  $F_1$  score reaches its best value at 1 and worst at 0. TCSS does better than Resnik for semantic similarity cutoff scores in all three ontologies.



### Supplementary figure S4 - ROC curves for *S. cerevisiae* PPI dataset (IEA+)

ROC evaluations of semantic similarity measures at different cutoffs based on the *S. cerevisiae* PPI dataset derived from DIP are shown. The evaluation was performed using cellular component, biological process, molecular function ontology of GO. Bestmatch average (bma) and maximum (max) approaches for combining multiple annotations are used on dataset with (IEA+) electronic annotations. TCSS & Resnik show best ROC profiles for all three ontologies.



### Supplementary figure S5 - F-score curves for *S. cerevisiae* PPI dataset (IEA+)

 $F_1$  score (harmonic mean of precision and recall) evaluations of TCSS and Resnik semantic similarity measures at different cutoffs based on the *S. cerevisiae* PPI dataset derived from DIP are shown. The evaluation was performed using cellular component, biological process, molecular function ontology of GO. Best-match average (bma) and maximum (max) approaches for combining multiple annotations was used on dataset with (IEA+) electronic annotations.  $F_1$  score reaches its best value at 1 and worst at 0. TCSS does better than Resnik for semantic similarity cutoff scores in all three ontologies.



### Supplementary figure S6 - ROC curves for *H. sapiens* PPI dataset (IEA-)

ROC evaluations of semantic similarity measures at different cutoffs based on the *H. sapiens* PPI dataset derived from DIP are shown. The evaluation was performed using cellular component, biological process, molecular function ontology of GO. Best-match average (bma) and maximum (max) approaches for combining multiple annotations were used on dataset without (IEA-) electronic annotations. TCSS & Resnik show best ROC profiles for all three ontologies.



### Supplementary figure S7 - F-score curves for *H. sapiens* PPI dataset (IEA-)

 $F_1$  score (harmonic mean of precision and recall) evaluations of semantic similarity measures at different cutoffs based on the *H. sapiens* PPI dataset derived from DIP are shown. The evaluation was performed using cellular component, biological process, and molecular function ontologies of GO. Best-match average (bma) and maximum (max) approaches for combining multiple annotations were used on dataset without (IEA–) electronic annotations.  $F_1$  score reaches its best value at 1 and worst at 0. TCSS does better than Resnik for semantic similarity cutoff scores in all three ontologies.



### Supplementary figure S8 - ROC curves for *H. sapiens* PPI dataset (IEA+)

ROC evaluations of semantic similarity measures at different cutoffs based on the *H. sapiens* PPI dataset derived from DIP are shown. The evaluation was performed using cellular component, biological process, molecular function ontology of GO. Bestmatch average (bma) and maximum (max) approaches for combining multiple annotations were used on dataset with (IEA+) electronic annotations. TCSS & Resnik show best ROC profiles for all three ontologies.



### Supplementary figure S9 - F-score curves for *H. sapiens* PPI dataset (IEA+)

 $F_1$  score (harmonic mean of precision and recall) evaluations of semantic similarity measures at different cutoffs based on the *H. sapiens* PPI dataset derived from DIP are shown. The evaluation was performed using cellular component, biological process, and molecular function ontologies of GO. Best-match average (bma) and maximum (max) approaches for combining multiple annotations were used on dataset with (IEA+) electronic annotations.  $F_1$  score reaches its best value at 1 and worst at 0. TCSS does better than Resnik for semantic similarity cutoff scores in all three ontologies.



# Supplementary figure S10 - Effect of topology cutoff on (ROC) AUC and F-score for *S. cerevisiae* PPI dataset (IEA-)

Change in AUC (TPR/FPR ROC) values and average F-scores with respect to topology cutoff under different settings. BMA stands for best-match average approach of combining multiple annotations and MAX stands for maximum approach. Test was conducted separately for cellular component (CC), biological process (BP), and molecular function (MF) ontologies without IEA (IEA-) annotations.



# Supplementary figure S11 - Effect of topology cutoff on (ROC) AUC and F-score for *S. cerevisiae* PPI dataset (IEA+)

Change in AUC (TPR/FPR ROC) values and average F-scores with respect to topology cutoff under different settings. BMA stands for best-match average approach of combining multiple annotations and MAX stands for maximum approach. Test was conducted separately for cellular component (CC), biological process (BP), and molecular function (MF) ontologies with IEA (IEA+) annotations.



#### Supplementary figure S12 - Topology cutoff for S. cerevisiae PPI dataset

Topology cutoffs for cellular component (CC), biological process (BP), and molecular function (MF) ontologies were determined by evaluating AUC values and average F-scores at different cutoffs. The topology cutoff where both the AUC and average F-score maximized under different conditions is picked. Test was done with best-match average (bma) and maximum (max) approaches of combining multiple annotations on datasets with (IEA+) and without (IEA-) electronic annotations. Topology cutoff value chosen for CC is 2.4, BP is 3.6, and MF is 3.2 (marked by "×").



# Supplementary figure S13 - Effect of topology cutoff on (ROC) AUC and F-score for *H. sapiens* PPI dataset (IEA-)

Change in AUC (TPR/FPR ROC) values and average F-scores with respect to topology cutoff under different settings. BMA stands for best-match average approach of combining multiple annotations and MAX stands for maximum approach. Test was conducted separately for cellular component (CC), biological process (BP), and molecular function (MF) ontologies without IEA (IEA-) annotations.



# Supplementary figure S14 - Effect of topology cutoff on (ROC) AUC and F-score for *H. sapiens* PPI dataset (IEA+)

Change in AUC (TPR/FPR ROC) values and average F-scores with respect to topology cutoff under different settings. BMA stands for best-match average approach of combining multiple annotations and MAX stands for maximum approach. Test was conducted separately for cellular component (CC), biological process (BP), and molecular function (MF) ontologies with IEA (IEA+) annotations.



#### Supplementary figure S15 - Topology cutoff for H. sapiens PPI dataset

Topology cutoffs for cellular component (CC), biological process (BP), and molecular function (MF) ontologies were determined by evaluating AUC values and average F-scores at different cutoffs. The topology cutoff where both the AUC and average F-score maximized under different conditions is picked. Test was done with best-match average (bma) and maximum (max) approaches of combining multiple annotations on datasets with (IEA+) and without (IEA-) electronic annotations. Topology cutoff value chosen for CC is 3.0, BP is 4.0, and MF is 3.6 (marked by "×").



### Supplementary figure S16 - Correlation with gene expression

Pearson correlation between gene expression similarity and semantic similarity on *S. cerevisiae* dataset are shown. The evaluation was performed for cellular component, biological process, and molecular function ontologies of GO. Best-match average (bma) approach for combining multiple GO annotations was used. TCSS showed best correlation with gene expression in all three ontologies.



### Supplementary figure S17 - Correlation with CESSM dataset

Correlation between semantic similarity and sequence, enzyme commission (EC), protein family (Pfam) similarity using online CESSM tool. The evaluation was performed for cellular component (CC), biological process (BP), and molecular function ontologies (MF) of GO. Best-match average (bma) approach for combining multiple GO annotations was used on the dataset without (IEA–) electronic annotations. TCSS showed best correlation with EC & Pfam similarity for CC ontology and same as Resnik's for MF and BP ontologies.



### Supplementary table S1 - Area under ROC curves for *H. sapiens* PPI dataset

Area under ROC curves for *H. sapiens* PPI dataset. The tests were performed separately for cellular component (CC), biological process (BP), and molecular function (MF) ontologies. *Best-match average* and *maximum* approaches were used for datasets "with (IEA+) and without (IEA-)" electronic annotations.

		IEA-			IEA+		
		CC	BP	MF	CC	BP	MF
TCSS	max	0.80	0.89	0.80	0.82	0.92	0.85
	bma	0.78	0.87	0.79	0.79	0.90	0.84
Resnik	max	0.80	0.89	0.80	0.81	0.92	0.84
	bma	0.77	0.87	0.79	0.79	0.90	0.84
Lin	max	0.78	0.88	0.74	0.76	0.91	0.78
	bma	0.76	0.86	0.73	0.75	0.88	0.78
Jiang	max	0.76	0.86	0.70	0.71	0.88	0.70
	bma	0.73	0.83	0.67	0.65	0.80	0.65
Schlicker	max	0.74	0.84	0.71	0.72	0.87	0.75
	bma	0.72	0.83	0.70	0.70	0.85	0.74
SimGIC		0.70	0.68	0.63	0.68	0.74	0.68

Supplementary table S2 - Gene expression datasets from GeneMANIA Gene Expression Omnibus (GEO) identifiers, series title, pubmed ids of gene expression datasets downloaded from GeneMANIA.

GEO id	Series title	Pubmed id	
GSE1311	YDRseries1, Yeast desiccation / rehydration time course	16332871	
GSE1312	YDRseries2, Yeast desiccation / rehydration time course	16332871	
GSE1313	YDRseries3, Yeast desiccation / rehydration time course	16332871	
GSE1639	Rpd3 and histone H3 and H4 deletions/mutations	15456858	
GSE1693	A novel response to microtubule perturbation in meiosis	15899877	
GSE1723	Two-dimensional transcriptome analysis in chemostat cultures of S. cerevisiae	15496405,17241460,12414795	
GSE1814	Transcriptional effects of the TOR2-controlled signaling function	15476558, 16959779	
GSE1938	Phosphomannose isomerase gene (PMI40) deletion strain cultivated in varying initial	15520001	
	mannose concentrations		
GSE1975	Simultaneous genotyping, gene expression measurement, and detection of allele-specific	15687292	
	expression		
GSE2076	leu3p dependent transcription	15949974	
GSE2224	Experimental condition	15878181	
GSE2343	TFIIH mutants treated with methyl methanesulfonate	15837426	
GSE3076	Impact of Nonsense-mediated mRNA Decay on the Global Expression Profile of Budding	17166056	
	Yeast		
GSE3431	Logic of the yeast metabolic cycle	16254148	
GSE3806	Histone H2B^3-32, H2B K->G, H2B^3-37, and H2B^30-37 mutations	16648479	
GSE3821	Short term perturbation	16969341	
GSE4135	Wild type yeast and H3del(1-28) and H4del(2-26) yeast grown in complete synthetic media	16461773	
GSE4669	Response of yeast to saponin treatment	16870766	
GSE5238	SFP1 dependent transcription	18174152	
GSE5301	Expression data from yeast treated with enediynes compared to gamma radiation	17163986	
GSE6073	Rap1 and Abf1 DNA-binding ts mutants and wild type after 1 hr at 37 C	17158163	
GSE6190	Temperature-dependent transcriptional response under anaerobic C and N limitations in Yeast	17928405	
GSE6405	Transcriptional responses of yeast to preferred and non-preferred nitrogen sources in C-lim	17419774	
GSE7660	Sch9 Is a Major Target of TORC1 in Saccharomyces cerevisiae	17560372	
GSE7820	Transcript and Proteomic Analyses of Wild-Type and GPA2 Mutant Saccharomyces	17700863	
GSETCEC	cerevisiae Strains	11100000	
GSE8187	Adaptation of S. cerevisiae to fermentative conditions	18304306	
GSE8536	The response of Saccharomyces cerevisiae to stress throughout a 15-day wine fermentation	18215224	
GSE8761	Transcriptional profiling of ribosomal protein knockouts	17981122	
GSE8825	Coordination of Growth Rate, Cell Cycle, Stress Response and Metabolic Activity in Yeast	17959824	
GSE8895	Role of Transcriptional Regulation in Controlling Fluxes in Central Carbon Metabolism of	14630934	
	Saccharomyces cerevisiae		
GSE8900	Genome-wide transcriptional responses of Saccharomyces cerevisiae to high carbon dioxide	15780657	
	concentrations		
GSE9217	Transcriptomes for different level of glucose	18679056	
GSE9302	A perturbation in the system leads to period doubling	17043222	
GSE9423	The Oxidative Stress Response of a Lager Brewing Yeast Strain during Industrial	18373683	
	Propagation and Fermentation		
GSE9482	GAL-NMD2	18087042	
GSE9590	Saccharomyces cerevisiae TPP 2-oxo acid decarboxylases	18281432	
GSE9644	Glucose Pulse to sfp1delta continuous cultures	18524923	
GSE11452	Saccharomyces cerevisiae chemostat steady state microarray compendium	19173729	
GSE12890	Xylose metabolism in recombinant Saccharomyces cerevisiae	18533012	