Protein-protein interactions: evolution, prediction and regulation

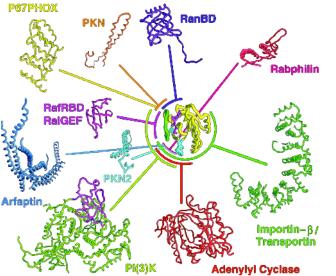
Anna Panchenko, National Center for Biotechnology Information, NIH, USA

Outline

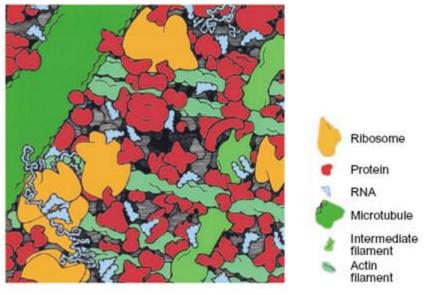
- Types of protein-protein interactions and their role in cell function.
- Physico-chemical properties of complexes and interfaces, binding hot spots.
- Experimental methods to identify interactions.
- Computational methods to predict PPIs.
- Evolution of protein interactions.
- Regulation of protein-protein binding.

Proteins function while interacting with other partners

- Many cellular processes are regulated through protein-protein interactions, distortions may cause diseases
- Proteins provide specific binding interfaces to interact with ligands.
- Binding selectivity and affinity is determined by physico-chemical properties of binding interfaces.
- Binding interfaces share common properties: conservation of certain amino acids, hot spots, geometry.



Vetter & Wittinghofer, Science 2001



Different types of protein-protein interactions.

- Permanent/obligatory subunits might not be stable in isolation and transient – subunits might fold independently.
- External are between different chains; internal are within the same chain.
- Homo- and hetero-oligomers depending on the similarity between interacting subunits.

Types of protein-protein interactions (PPI)

Obligate PPI

usually permanent

the protomers are <u>not</u> found as stable structures on their own *in vivo*



Obligate heterodimer Human cathepsin D

Permanent

(many enzyme-inhibitor complexes)

dissociation constant K_d=[A][B] / [AB]

10⁻⁷ - 10⁻¹³ M



Non-obligate permanent heterodimer

Thrombin and rodniin inhibitor

Weak

Non-obligate PPI

(electron transport complexes)

 $K_d mM-\mu M$

Intermediate

Transient



Non-obligate transient homodimer, Sperm lysin (interaction is broken and formed continuously)

(antibody-antigen, TCR-MHC-peptide, signal transduction PPI), K_d μM-nM

Strong

(require a molecular trigger to shift the oligomeric equilibrium)

K_d nM-fM

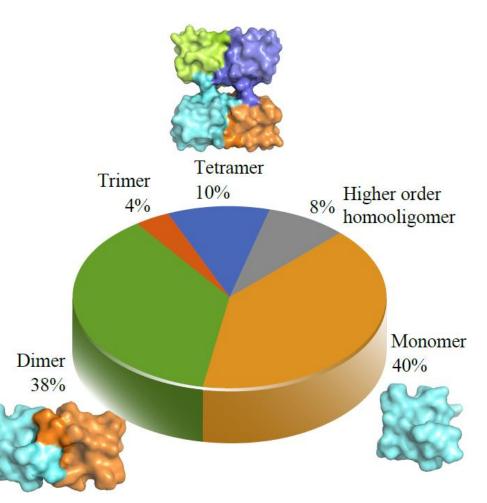
Bovine G protein dissociates into $G\alpha$ and $G\beta\gamma$ subunits upon GTP, but forms a stable trimer upon GDP

Role of homooligomers in a cell

• Mediate and regulate gene expression, activity of enzymes, ion channels, receptors and cellcell adhesion processes.

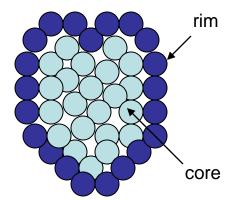
• Provide sites for allosteric regulation, new binding sites at interfaces to increase specificity.

• Provides stability, protection against denaturation.



Common properties of proteinprotein interactions.

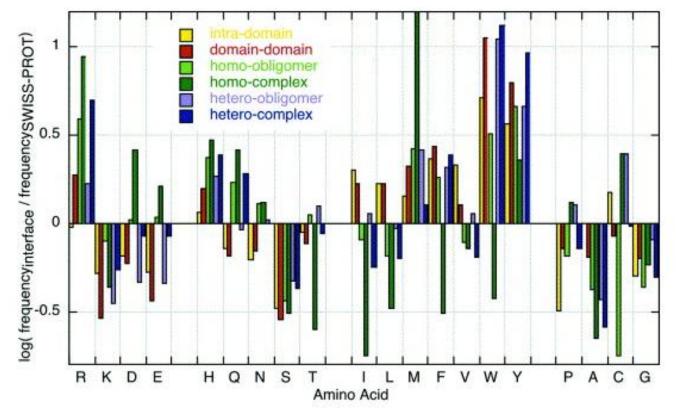
- Majority of protein complexes have a buried surface area ~1600±400 Å^2 ("standard size" patch).
- Complexes of "standard size" do not involve large conformational changes while large complexes do.
- Protein recognition site consists of a completely buried core and a partially accessible rim.



Top molecule

Bottom molecule

Amino acid composition of different types of complexes



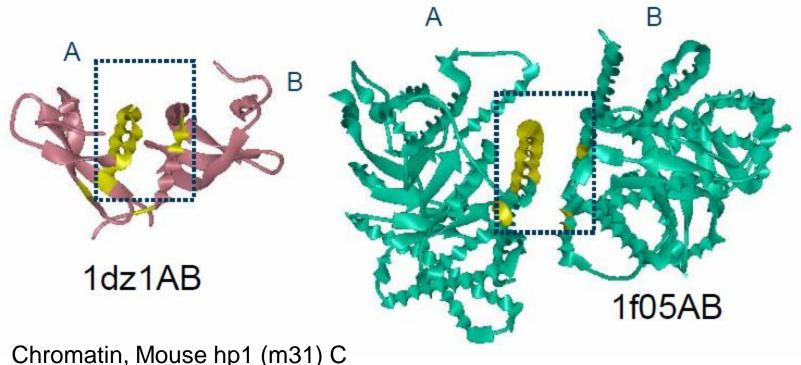
Ofran & Rost, JMB, 2003

Properties of different types of interfaces

- Non-obligate complexes tend to be more hydrophilic.
- Hydrophobic groups tend to be burried upon complex formation.
- Electrostatics, hydrogen bonds, salt bridges confer specificity.
- Permanent interfaces tend to be larger, less planar, and tightly packed.

Classification of interfaces

Similar interfaces- dissimilar functions



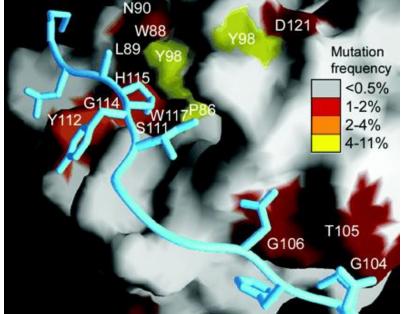
Chromatin, Mouse hp1 (m31) C terminal domain

Human transaldolase

Keskin, Gursoy, Nussinov, PRISM

Binding hot spots

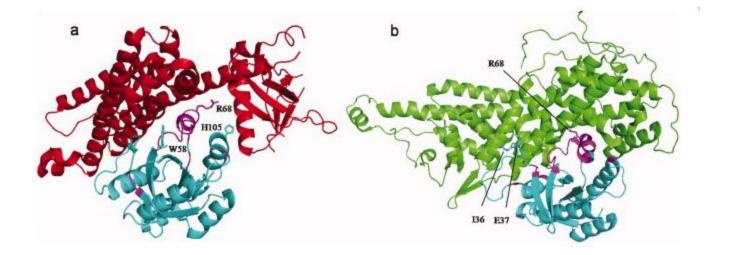
- Interface sites which contribute the most to binding energy (>2kcal/mol).
- Amino Acid composition: aromatic, Thr, Ser, Cys.
- Structurally and sequence conserved



Why do we need to identify binding hotspots?

 To understand how proteins bind to different partners – "binding promiscuity"

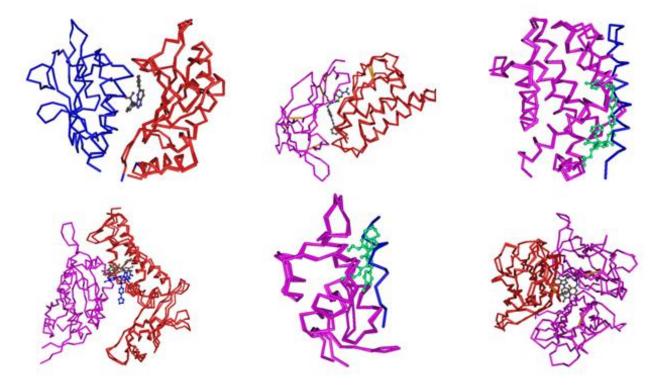
Interaction between GTPase domain and GEF



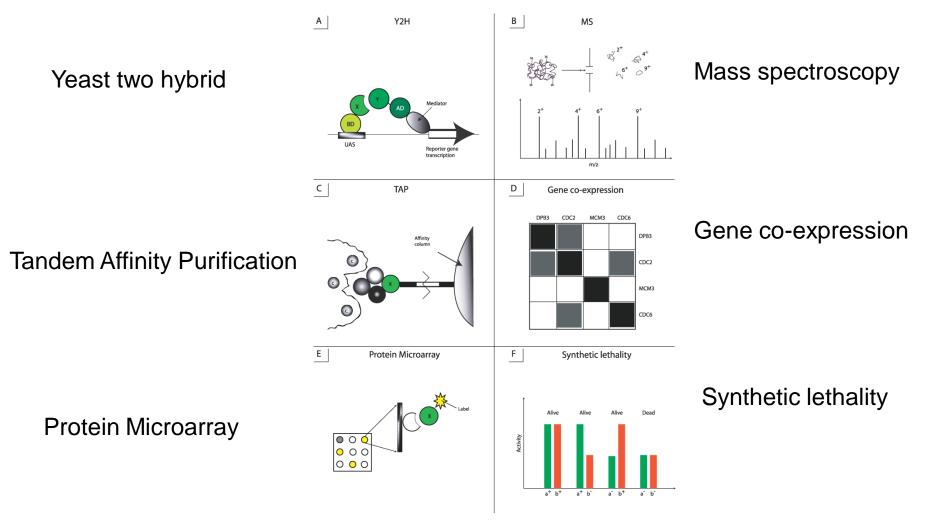
Tyagi et al, Protein Science 2009

Why do we need to identify binding hotspots?

 To target protein-protein interfaces by small molecule drugs



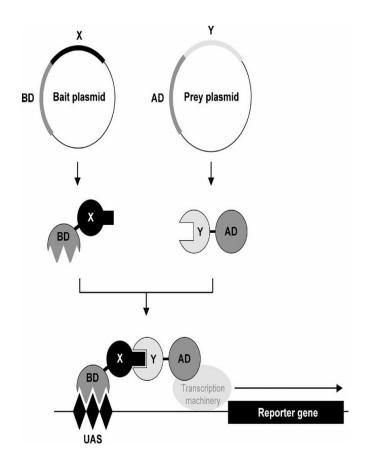
High-throughput methods to detect proteinprotein interactions



Shoemaker & Panchenko, PloS Comp Biol, 2007

Yeast two-hybrid experiments.

- Many transcription factors (ex: Gal4, LexA) have two distinct domains; one that directs binding to a promoter DNA sequence (BD) and another that activates transcription (AD).
- Fields and Song (1989) demonstrated that DNA-binding domain can not activate transcription at a promoter unless physically (not necessarily covalently) associated with an activating domain.

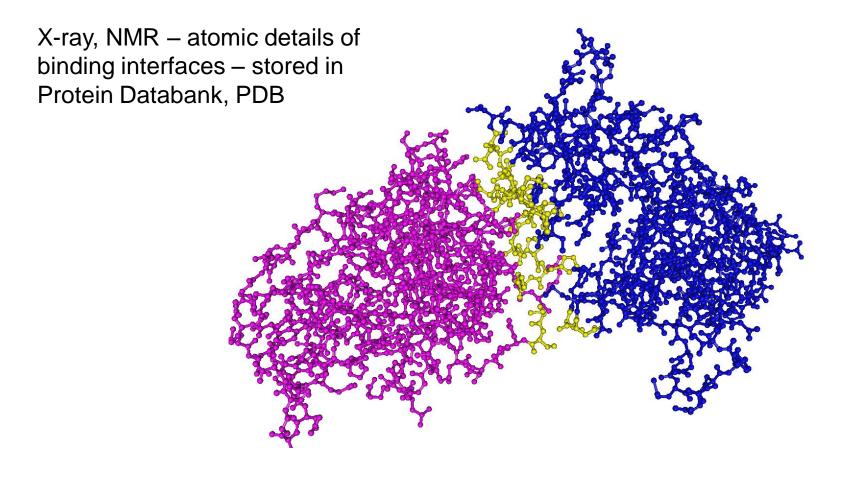


Causier, Mass Spectroscopy Reviews, 2004

Low-throuput biophysical methods

- X-ray crystallography, NMR
- Fluorescence resonance energy transfer (FRET)
- Surface plasmon resonance (SPR)
- Isothermal titration calorimetry (ITC)
- Atomic force microscopy

Resolving atomic details of interaction interfaces



Prediction of protein-protein interactions

Methods of prediction of functional associations and protein interactions

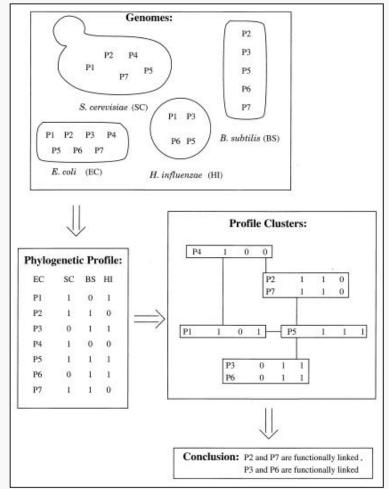
Method Name		Physical Interaction/ Functional Association
Gene co-expression	Р	F
Synthetic lethality	Р	F
Gene cluster and gene neighbor	Р	F
Phylogenetic profile	P, D	F
Rosetta Stone	Р	F
Sequence co-evolution	P, D	F
Classification	P, D	Ρ
Integrative	P, D	Ρ
Domain association	D	Ρ
Bayesian networks	P, D	F, P
Domain pair exclusion	D	Ρ
<i>p</i> -Value	D	Ρ

Phylogenetic profile method.

Functionally linked and putative interacting proteins should have orthologs in the same subset of fully sequenced organisms (Pellegrini et al, *PNAS* 1999).

Drawbacks:

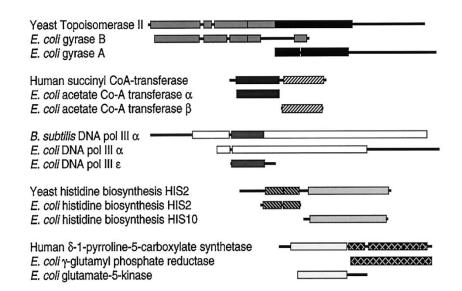
- high computational cost;
- dependence on homology detection between distant organisms;
- ubiquitous unlinked proteins present in all genomes – false positives;
- shared phylogenetic history between two proteins – false positives.



Rosetta Stone approach.

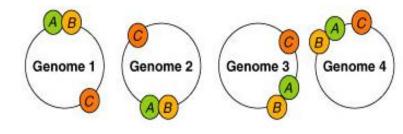
Some pairs of interacting domains have homologs which are fused into one protein chain – "Rosetta Stone" protein (Marcotte et al, *Science*, 1999).

• In *E.coli* ~ 6809 pairs of nonhomologous proteins; both proteins from each pair could be mapped to a single protein from some other genome.



Gene neighborhood method.

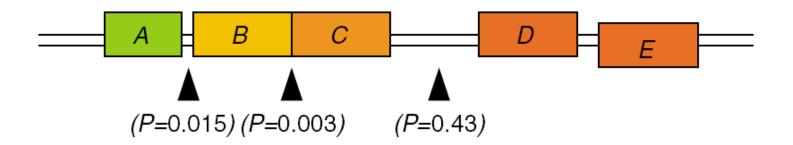
- Gene pairs from conserved gene clusters encode proteins which are functionally related and possibly interact.
- Conservation of gene order can be used to predict gene function.
- Analysis of gene order conservation : 65%–75% of coregulated genes interact physically (prediction of archael exosome by comparing GN in archaea, Koonin et al, Genome Res 2001)



Bowers et al, Genome Biology, 2004

Gene cluster method.

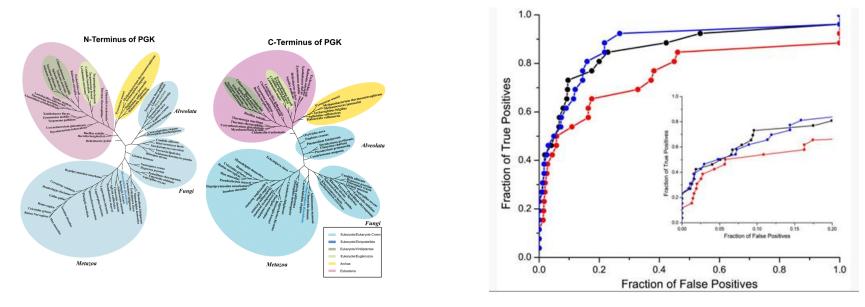
- Bacterial genes of related function are often transcribed simultaneously operon.
- Identification of operons is based on intergenic distances.



Bowers et al, Genome Biology, 2004

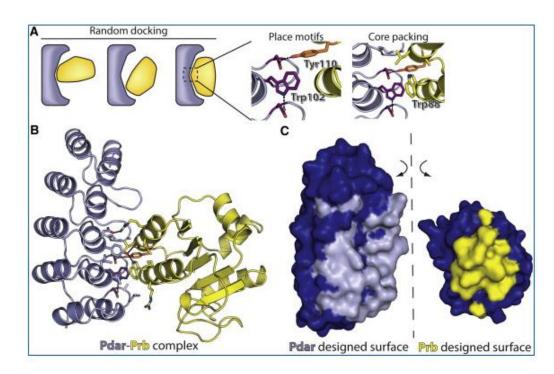
Coevolution of interacting proteins – "mirrortree" methods.

- Interacting proteins may co-evolve and their phylogenetic trees show similarity (*Goh et al, J.Mol.Biol.,2000*).
- Similarity between phylogenetic trees is measured by correlation coefficient between distance matrices.
- Signal comes from both correlated evolution of binding sites and whole protein sequence (*Kann et al, JMB 2009*).



Interface design

- Computationally alter interface to modify function
- Create useful properties
- Alter oligomeric state
- Alter specificity
- Novel interactions



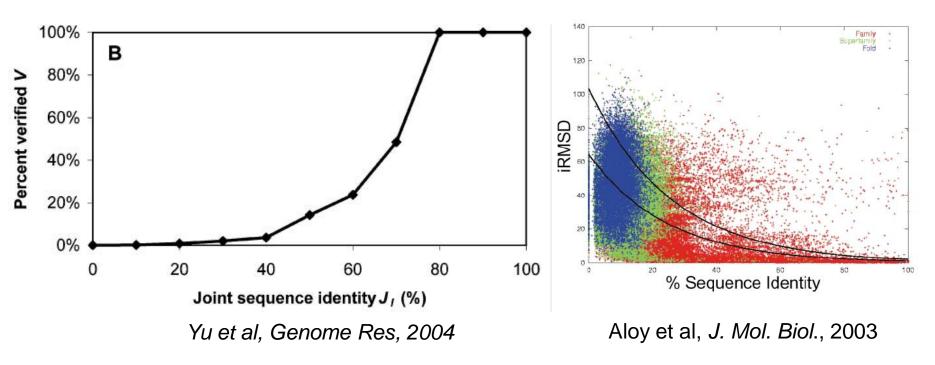
Karanicolas et al, Mol Cell, 2011

Evolution of protein interactions

Conservation of protein-protein interactions.

Conservation of interologs

Conservation of binding interfaces



Mechanisms of evolution of novel interfaces

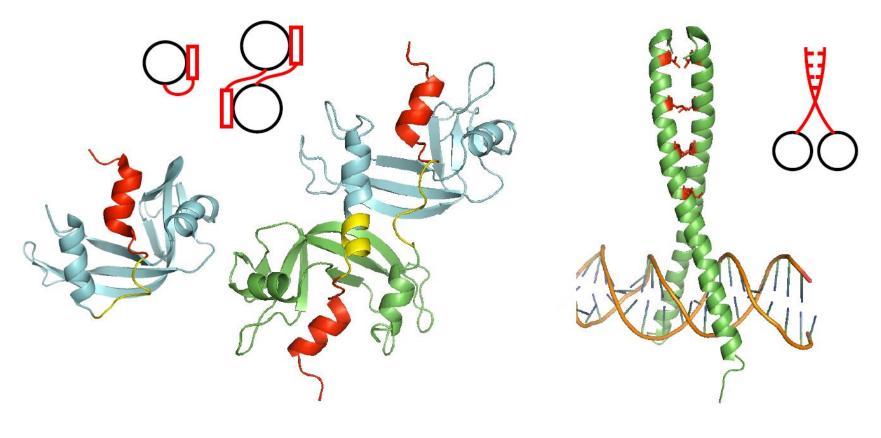
 Gene duplication with subsequent diversification (Pereira-Leal and Teichmann, Genome Res, 2005; Reid et al, BMC Genomics, 2010)

Partial Complete GGTase I FTase RabGGTase

Types of duplication

- Domain shuffling
- Point mutations on interfaces
- Insertions and deletions
- Other

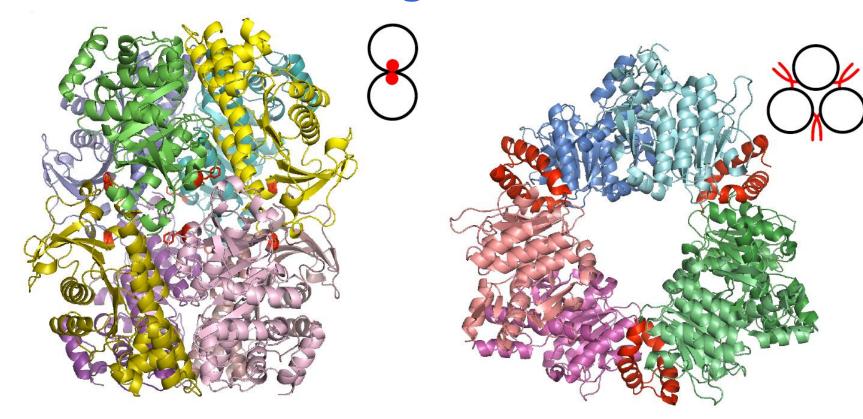
Evolutionary mechanisms to form oligomers



Domain swapping

Leu zipper

Evolutionary mechanisms to form oligomers

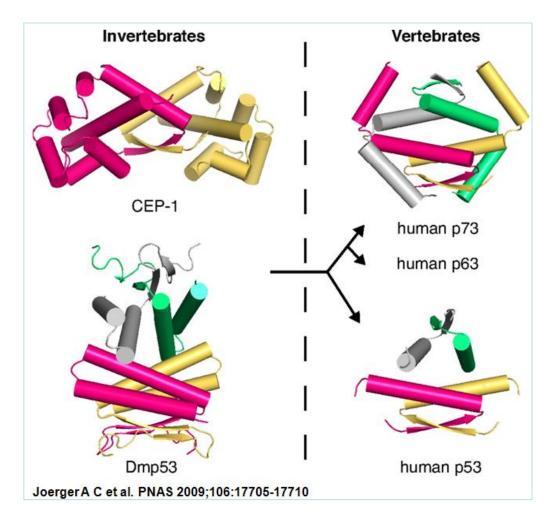


Mutations on interface

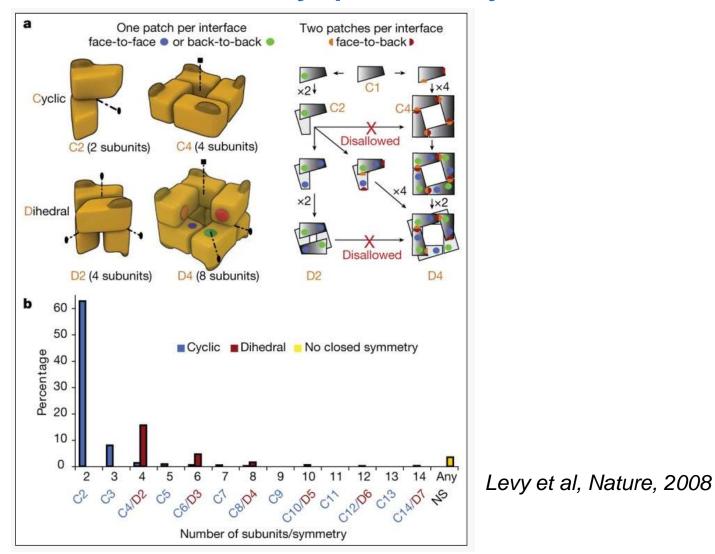
Insertions/deletions

Evolution of new specificity through oligomerization

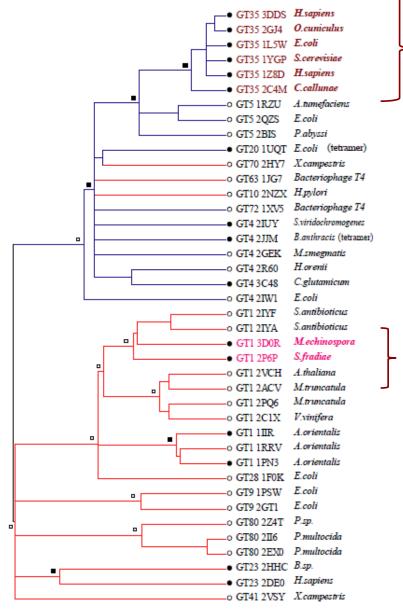
Stabilization of p63/p73 tetramer leads to separation of their pathways from p53 pathway

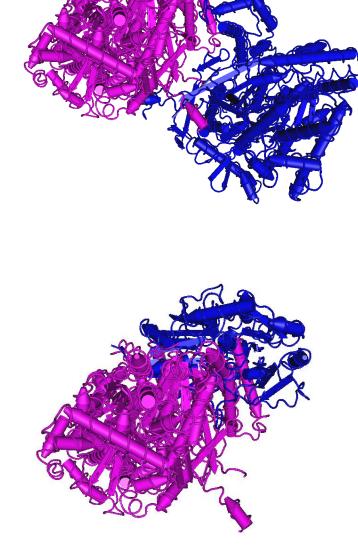


Assembly pathway mimics the evolutionary pathway



Evolution of homooligomeric binding modes: Glycosyltransferase

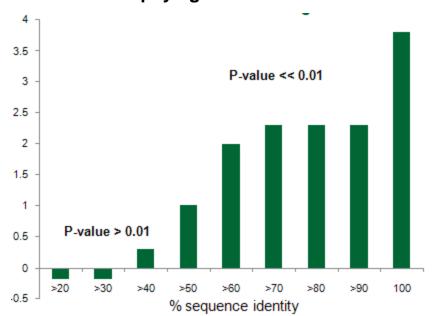




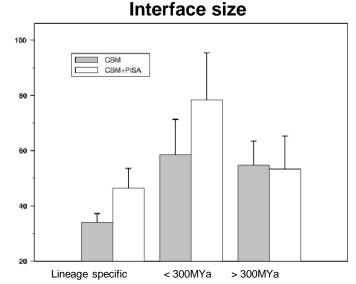
Hashimoto et al, J Mol Biol, 2010

Conservation of binding modes in evolution

Logarithm of probability ratio for finding the same or different binding modes on phylogenetic tree



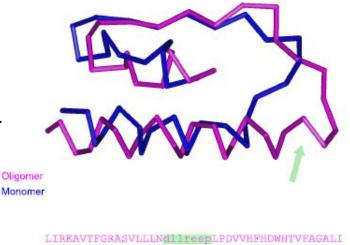
Dayhoff et al, JMB 2010



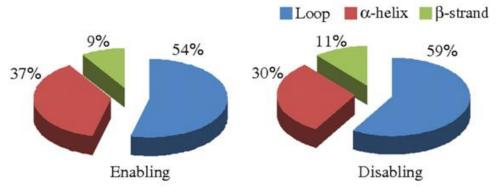
- binding modes are well conserved within phylogenetic clades sharing more than 50% sequence identity
- lineage-specific binding modes are smaller, less stable. Newer interactions are weaker

Role of Insertions and deletions in formation of oligomers

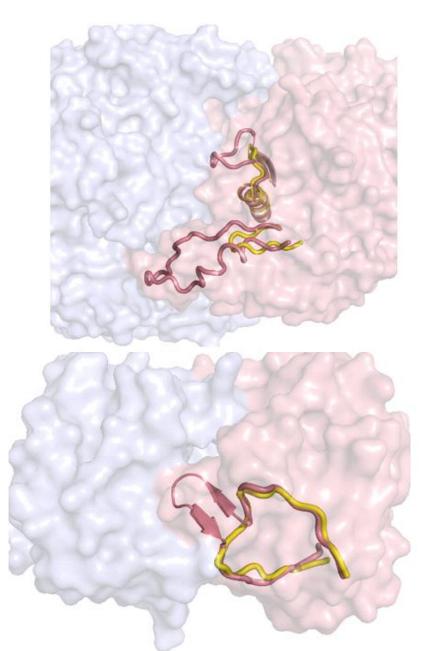
- "enabling" and "disabling" loops (Akiva et al, PNAS 2008)
- Insertions/deletions occur more frequently or the interface than on the surface (*P*value << 10e-7) – "enabling" and "disabling regions"
- 25% homooligomers have enabling and disabling regions;
- they contain more polar and charged residues, Gly and Pro than "conventional interfaces"



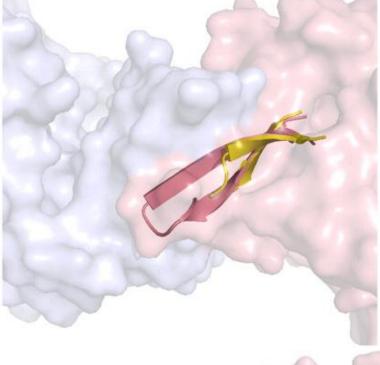
YLRVNALLADKLLPLLO-----DDDITWIHDYHLLPFAHEL

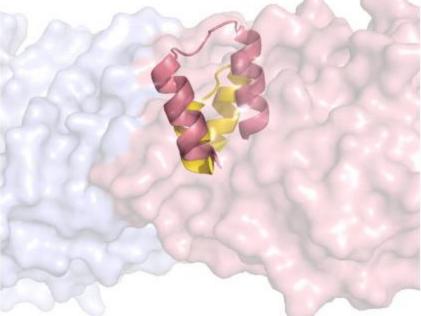


Hashimoto & Panchenko, PNAS 2010



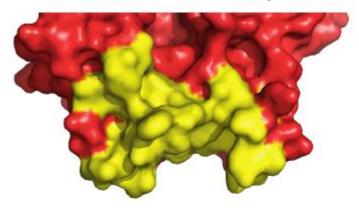
Enabling region in homodimer (1P3C) Aligned region in monomer (1FQ3)



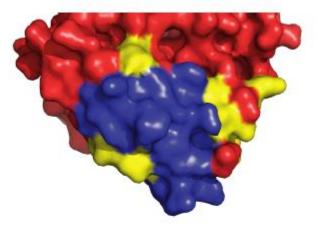


Disabling regions

Glycogenin glucosyltransferase, disrupting features



Eukaryotes, dimer



Bacteria, monomer

1LL2: 122 aap	158
1G9R: 123 nwlgasidlfverqegykqkigxadgeYYFNAGVLLINLKKWRrHDIFKXSSEWVeqykdv 1	183
1LL2: 159 FDGGDQGLLNTFFNSWattdirKHLPFIYNLSSISIYSY1pafkafgaNAK 2	
1G9R: 184 XQYQDQDILNGLFKGGvCYANSRFNFXPTNYAFXanwfasrhtdplyrdrtntvxPVA 2	241

Prediction of oligomeric states from sequence

	Sensitivity	Specificity	Precision	Error rate
Enabling/disabling features	0.70	0.74	0.94	0.36
% identity	0.71	0.62	0.91	0.38
RMSD	0.72	0.60	0.90	0.40
GSAS	0.81	0.57	0.91	0.43
BLAST	0.74	0.53	0.89	0.47

Regulation of protein-protein binding

Mechanisms of regulation

- Availability/abundance
 - Gene expression, translation
 - Translocation of proteins or
 - substrates
 - Turnover
- Proteolytic activation
- Inteins

Mechanisms of protein regulation

State R

active

Effectors, inhibitors, PTM)

State T *inactive*

- ation by another protein or small males
- Regulation by another protein or small molecule
 Reversible covalent post-translational modifications
- Allosteric activation and inhibition

Mechanisms of protein regulation

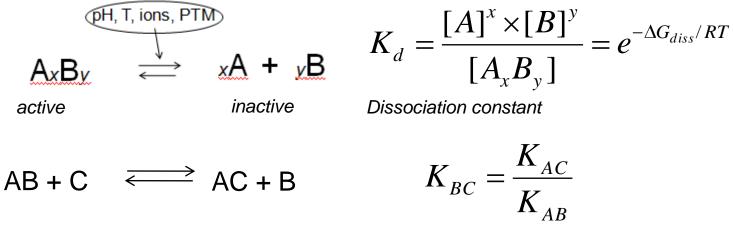
State T inactive



State R active

- Regulation by another protein or small molecule
- Reversible covalent post-translational modifications
- Allosteric activation and inhibition

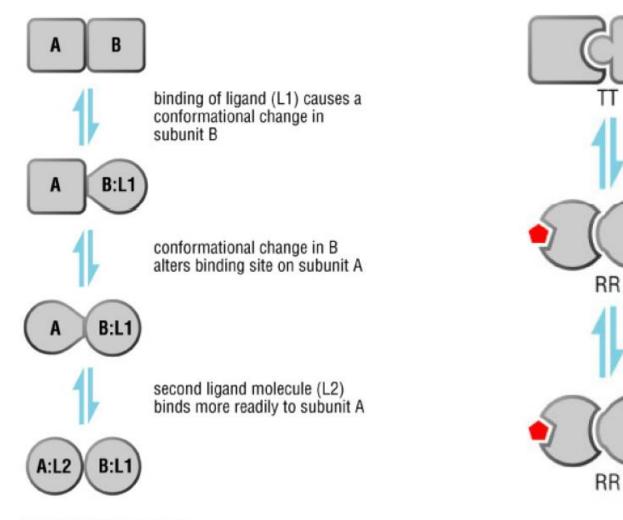
Transitions between different oligomeric states



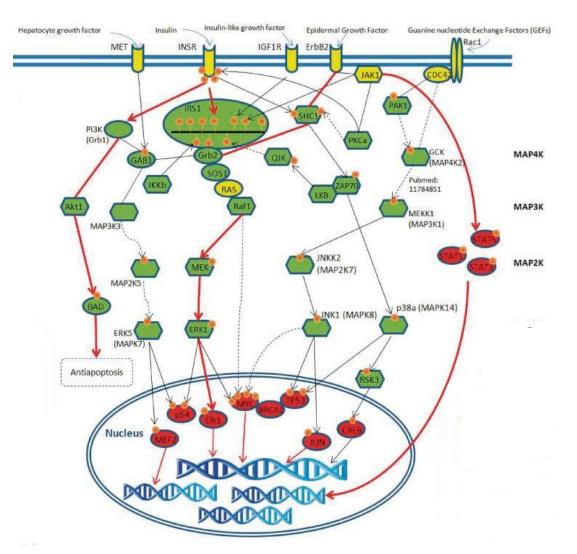
Binding selectivity constant

Allosteric regulation

+X

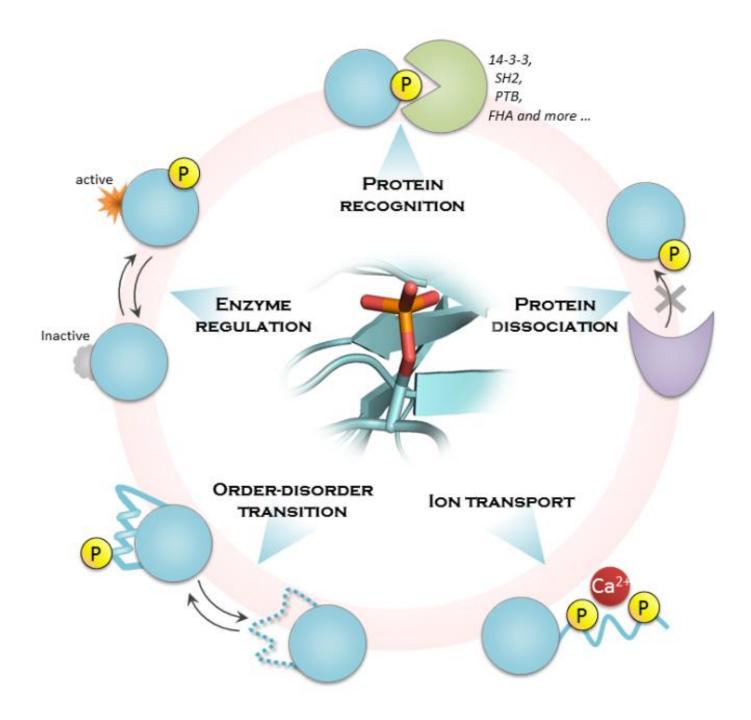


Signal transduction through protein-protein interactions and post-translational modifications

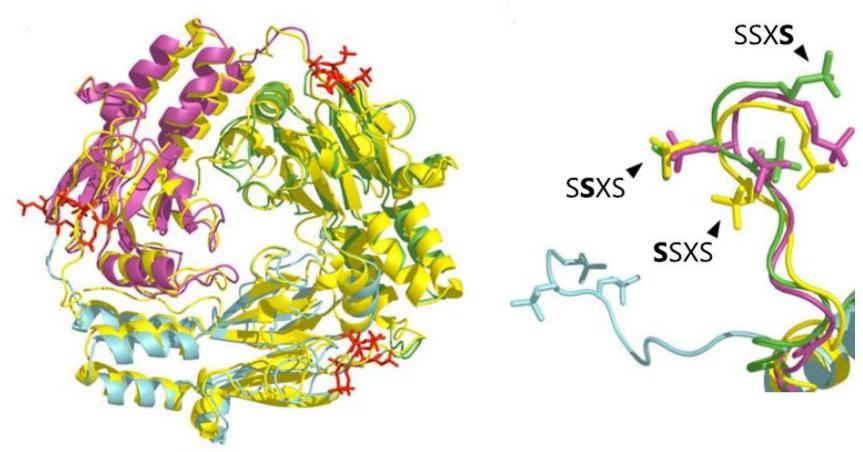


Protein control through covalent modifications

- 50-90% of human proteins is posttranslationaly modified
- over 40 different modifications have been described
- most important: phosphorylation, glicosylation, lipidation, methylation, Nacetylation, S-nitrosylation, SUMOilation



Smad2-MH2 trimer: phospho-group promotes the trimer formation



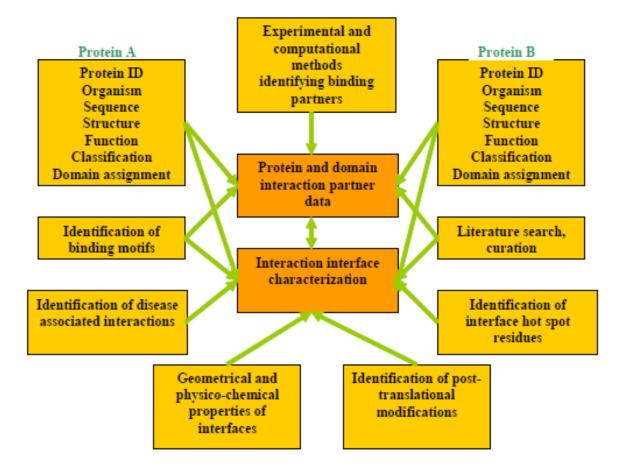
Nishi et al, submitted

Effect of phosphorylation/dephosphorylation on Smad complex formation

Protein	Site	pSite	AB	BC	AC	Average ΔΔΔG, kcal/mol
Smad2 (1khx)	<u>s</u> sxs	S->pS	0.59	0.59	0.59	0.59
	S <u>S</u> XS	pS->S	0.88	0.88	0.88	0.88
	SSX <u>S</u>	pS->S	1.53	1.53	2.86	1.97
Smad1 (1khu)	<u>s</u> sxs	S->pS	-2.11	1.58	-0.22	-0.25
	S <u>S</u> XS	S->pS	-0.9	1.49	-1.74	-0.38
	SSX <u>S</u>	S->pS	-1.45	-1.87	-1.08	-1.47

Protein-protein interaction databases

Data flow in protein interaction databases



Protein interaction databases

Database	Proteins/Domains	Туре	Number of Interactions
DIP ^a , LiveDIP	Ρ	E,S	55,733
BIND ^a	Р	E,C,S	83,517
MPact/MIPS ^a	Р	E,C,F	15,488 (4,300) ^b
STRING	Ρ	E,P,F	730,000 (proteins)
MINT ^a	Р	E,C	71,854
IntAct ^a	Ρ	E,C	68,165
BioGRID ^a	Р	E,C	116,000 (30,000) ^b
HPRD	Р	E,C	33,710
ProtCom	P,D	S,H	1,770
3did, Interprets	D	S,H	3,304
Pibase, ModBase	D	S,H	2,387
CBM	D	S	2,784
SCOPPI	D	S	3,358
iPfam	D	S	3,019
InterDom	D	Р	30,037
DIMA	D	F,S	_
Prolinks	Р	F	_

BioGRID, Stark et al, NAR 2011

Organism	Experiment Type	Raw Interactions	Non- Redundant Interactions	Unique Proteins	Unique Publications
Human	PHYSICAL	60570	39635	10259	12411
	GENETIC	513	489	525	198
	COMBINED	61083	39938	10448	12470
All	PHYSICAL	204613	140813	31754	20369
	GENETIC	184715	132714	9420	8827
	COMBINED	389328	267879	33563	26894

<u>IBIS</u> – NCBI server to analyze and infer interactions and binding sites http://www.ncbi.nlm.nih.gov/Structure/ibis/ibis.cgi

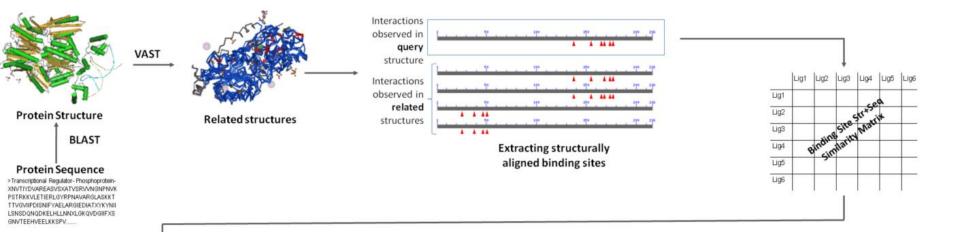


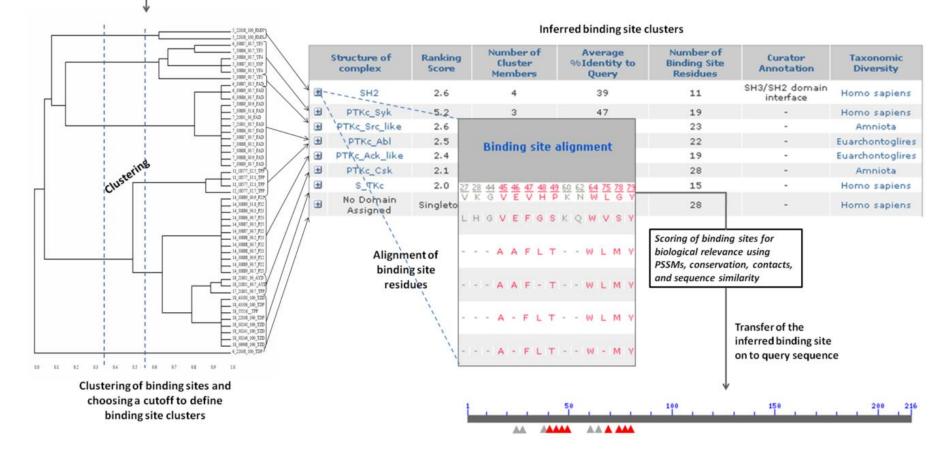
"Observed" interactions – from structures "Inferred" interactions – from homologous structures with observed interactions

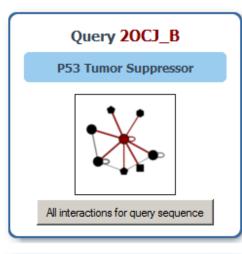
Biological relevance of binding sites:

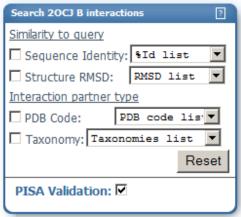
- occurs in several non-redundant homologs;
- structurally and sequence conserved;
- binds biologically active molecules;
- validated by PISA algorithm (Krissinel & Henrick, 2007);
- overlaps with the curated binding site

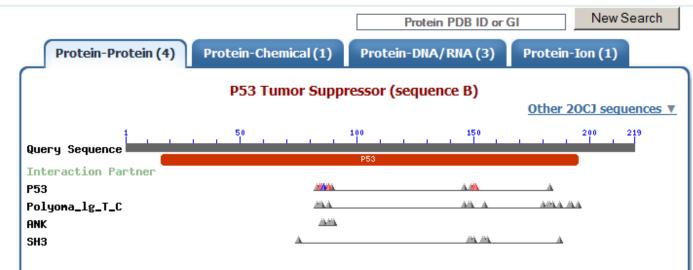
Shoemaker et al, NAR 2010 Thangudu et al, BMC Bioinformatics, 2010





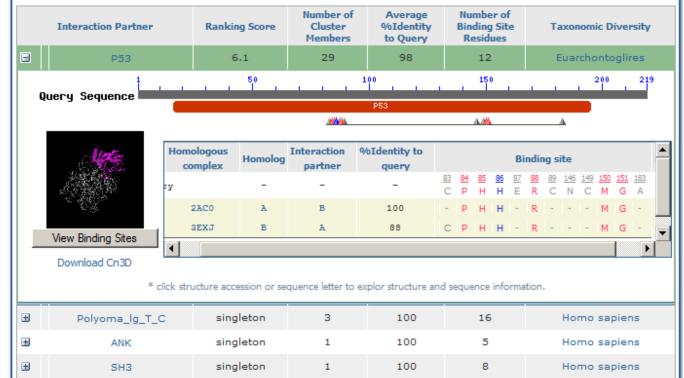


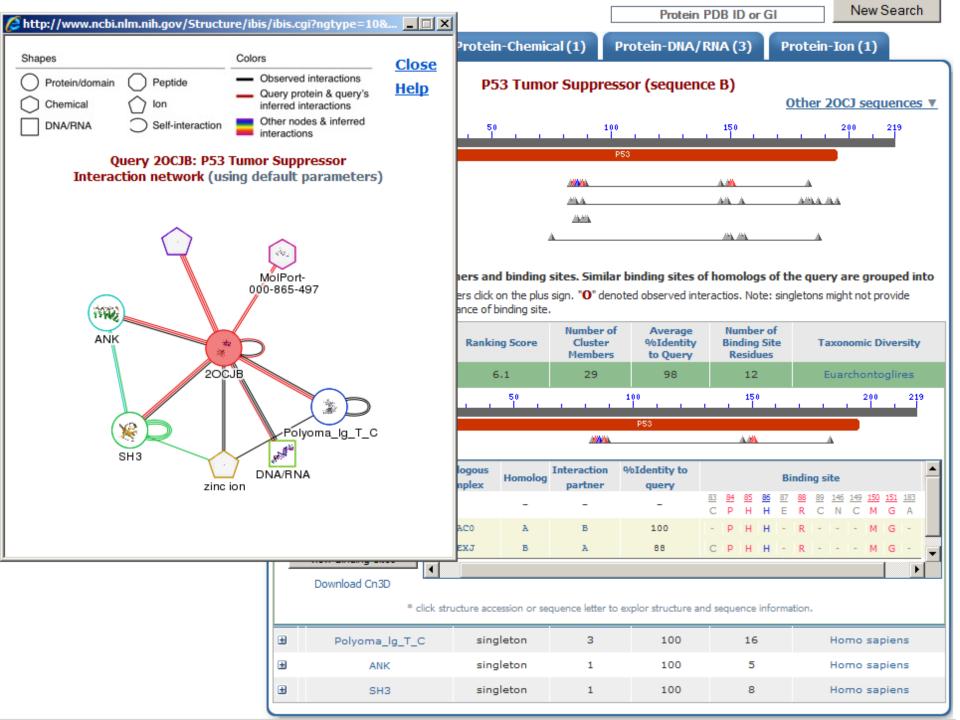




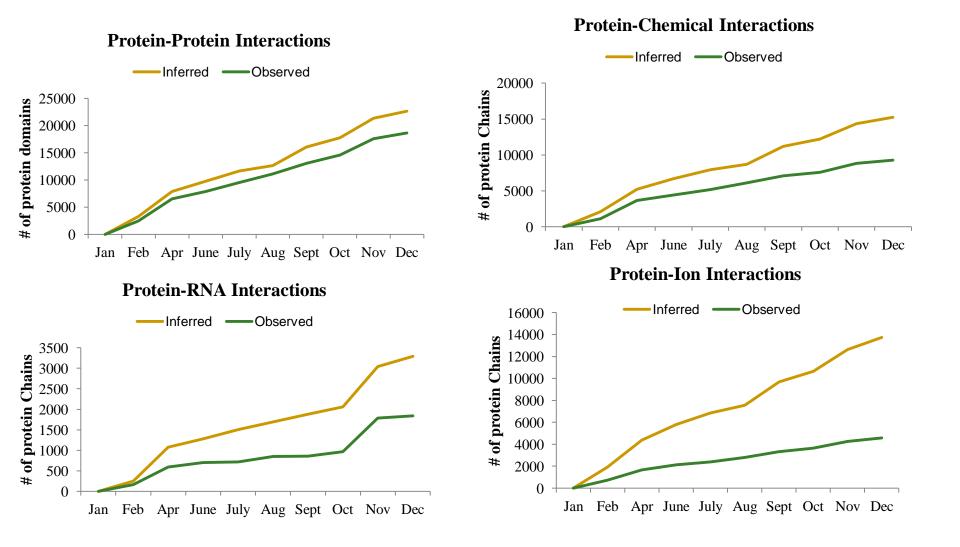
List of protein interaction partners and binding sites. Similar binding sites of homologs of the query are grouped into

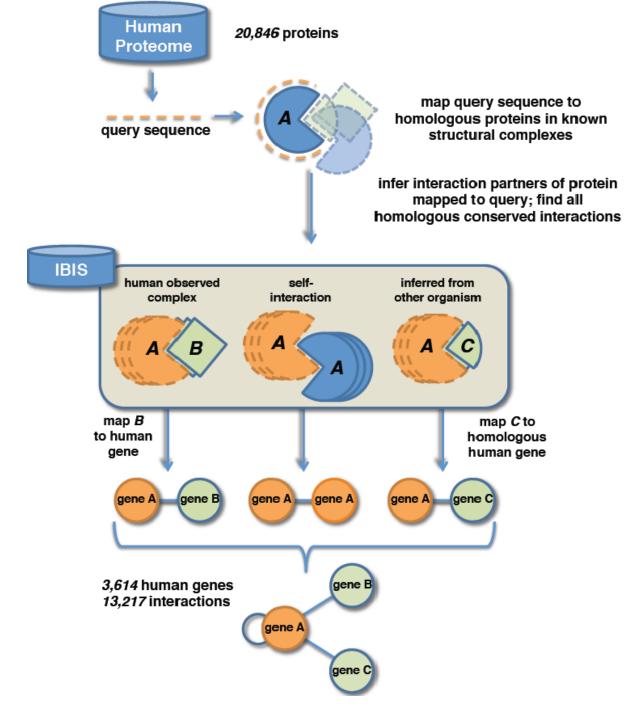
clusters. To view the cluster members click on the plus sign. "O" denoted observed interactios. Note: singletons might not provide enough evidence for biological relevance of binding site.



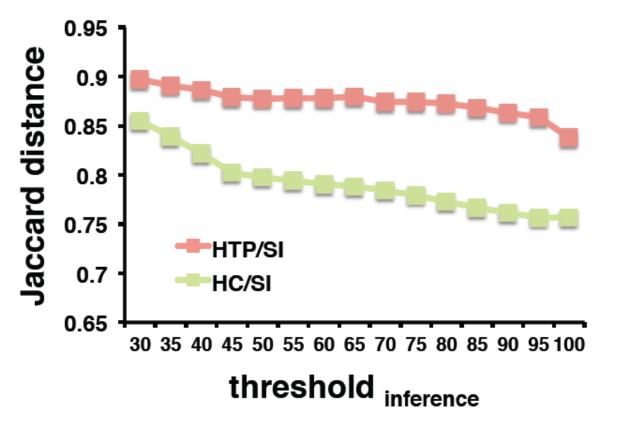


Growth of IBIS data over 2010





Comparison of structurally inferred (SI), high-throughput (HTP) and high confidence HTP (HC) networks

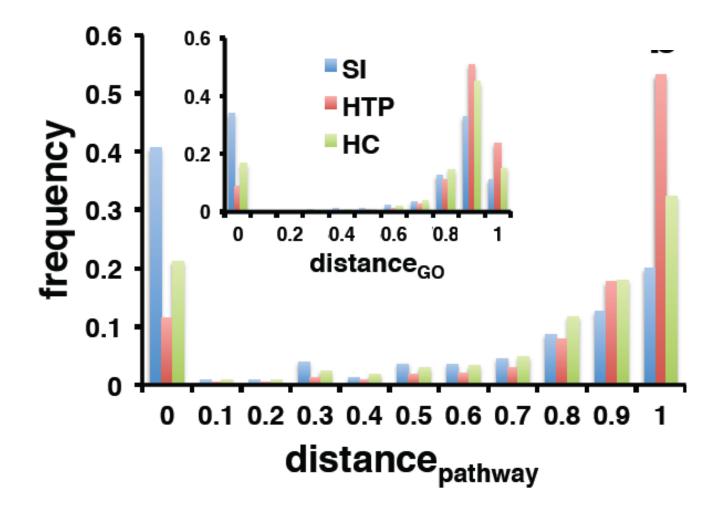


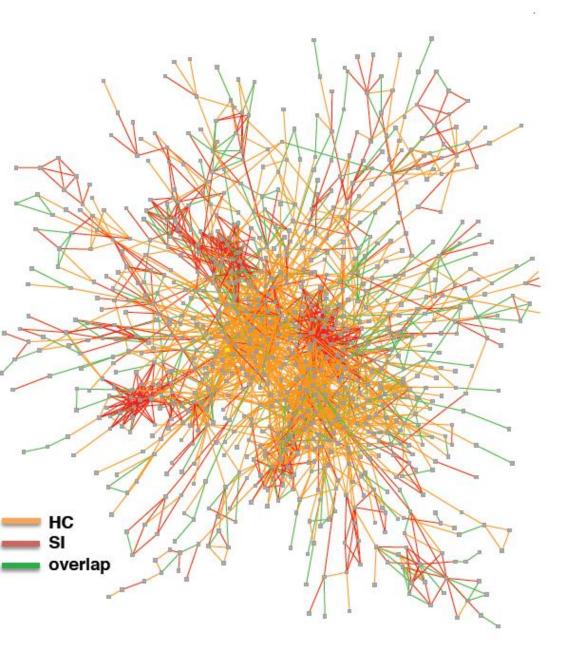
Inference threshold – similarity between query protein and

closest homolog with known complex

Tyagi et al, submitted

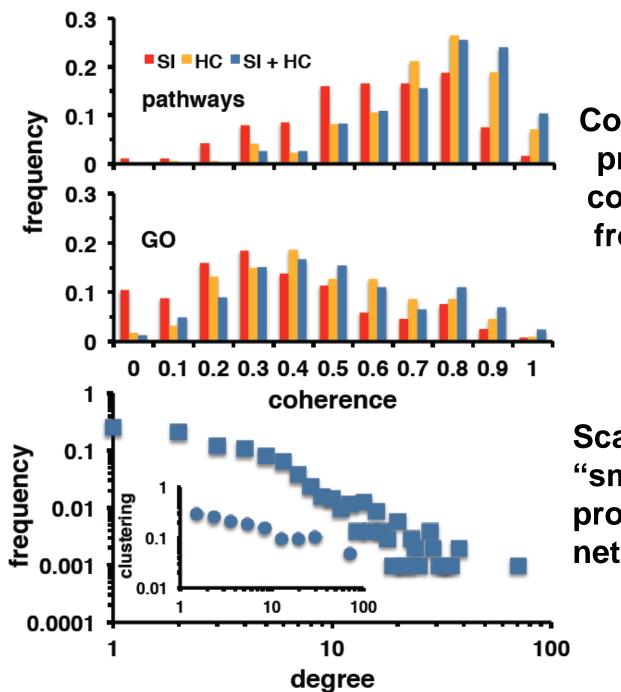
Structurally inferred networks are more functionally coherent than high-throughput networks





"Merged" networks = structurally inferred (SI) + high confidence highthroughput (HC), ~5500 proteins and ~17000 interactions

SI and HC complement each other; ~20% of HC interactions are observed in SI and ~50% SI interactions are observed in HC



Coherence – fraction of proteins in a network composed by proteins from a given pathway

Scale-free, modularity, "small-world" properties` of merged network

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Benjamin Shoemaker

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Dachuan Zhang