

Robust and accurate prediction of residue–residue interactions across protein interfaces using evolutionary information

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Abstract Do the amino acid sequence identities of residues that make contact across protein interfaces covary during evolution? If so, such covariance could be used to predict contacts across interfaces and assemble models of biological complexes. We find that residue pairs identified using a pseudo-likelihood-based method to covary across protein–protein interfaces in the 50S ribosomal unit and 28 additional bacterial protein complexes with known structure are almost always in contact in the complex, provided that the number of aligned sequences is greater than the average length of the two proteins. We use this method to make subunit contact predictions for an additional 36 protein complexes with unknown structures, and present models based on these predictions for the tripartite ATP-independent periplasmic (TRAP) transporter, the tripartite efflux system, the pyruvate formate lyase-activating enzyme complex, and the methionine ABC transporter.

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Introduction

Recent work has demonstrated the accuracy of coevolution-based contact prediction for monomeric proteins using a global statistical model (Thomas et al., 2008) to distinguish between direct and indirect couplings (Marks et al., 2011; Morcos et al., 2011; Hopf et al., 2012; Nugent and Jones, 2012; Jones et al., 2012; Lapedes et al., 2012; Marks et al., 2012; Sulikowska et al., 2012; Kamisetty et al., 2013). While early approaches relied on estimating an inverse covariance matrix (Marks et al., 2011; Morcos et al., 2011; Jones et al., 2012), more recent studies have shown that a pseudo-likelihood-based approach (Balakrishnan et al., 2011) results in more accurate predictions (Ekeberg et al., 2013; Kamisetty et al., 2013) for a range of alignment sizes and protein lengths.

In contrast to this rich body of work for monomeric proteins, relatively little is known about the utility of such statistical models in predicting protein–protein interactions. The more general problem of predicting if two proteins interact with each other has been studied extensively using a wide variety of approaches (de Juan et al., 2013; Hosur et al., 2012; Zhang et al., 2012; Shoemaker and Panchenko, 2007; Valencia and Pazos, 2002; Ochoa and Pazos, 2010). Amino acid residue coevolution has been used to predict residue–residue interactions across interfaces with local statistical models (Pazos et al., 1997; Halperin et al., 2006). As noted above, the accuracy of these models is reduced by the confounding of direct and indirect correlations (Lapedes et al., 1999; Weigt et al., 2009); the application of global statistical models to coevolution-based contact prediction across interfaces has been limited to the case of the histidine-kinase/response-regulator two component system (Burger and van Nimwegen, 2008; Weigt et al., 2009; Schug et al., 2009; Dago et al., 2012).

eLife digest Proteins are considered the ‘workhorse molecules’ of life and they are involved in virtually everything that cells do. Proteins are strings of amino acids that have folded into a specific three-dimensional shape. Proteins must have the correct shape to function properly, as they often work by binding to other proteins or molecules—much like a key fitting into a lock. Working out the structure of a protein can, therefore, provide major insights into how the protein does its job.

Two or more proteins can bind together and form a complex to perform various tasks; and solving the structures of these complexes can be challenging, even if the structures of the protein subunits are known. Now, Ovchinnikov, Kamisetty, and Baker have developed a method for predicting which parts of the proteins make contact with each other in a two-protein complex.

Different species can have copies of the same proteins; but a copy from one species might have different amino acids at certain positions when compared to a related copy from another species. As such, when pairs of interacting proteins from different species are compared, there will be many positions in the two proteins that vary. However, if the amino acid at a position in one protein (let's call it ‘X’) varies, and the amino acid at, say, position ‘Y’ in the other protein also varies such that for any given amino acid at position Y there is often a specific amino acid at position X; positions X and Y are said to ‘co-vary’. Ovchinnikov et al. noticed that when a pair of amino acids (one from each protein in a two-protein complex) co-varied, these two amino acids tended to make contact with each other at the protein–protein interface.

Ovchinnikov et al. used the new method to make predictions about the protein–protein interfaces in 28 protein complexes found in bacteria, and also to make a prediction about the interface between protein subunits in the bacterial ribosome. When these predictions were checked against the actual structures, which were all known beforehand, they were found to be accurate if the number of copies of each protein being compared is greater than the average length of the two proteins.

Ovchinnikov et al. went on to predict the amino acids on the protein–protein interfaces for another 36 bacterial protein complexes with unknown structures, and to present models for four larger complexes. The next challenge is to extend the method to protein complexes that are found only in eukaryotes (i.e., not in bacteria). Since the number of related copies for eukaryotic proteins tends to be smaller, there are fewer proteins to compare and it is therefore harder to detect ‘covariation’ when it occurs.

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In this study, we examine residue–residue covariation across protein–protein interfaces using a pseudo-likelihood-based statistical method. In a large set of complexes of known structure, we find that covarying pairs of positions are almost always in contact in the three-dimensional structure, provided there are sufficient aligned sequences. We find further that significant residue–residue covariation occurs frequently between physically interacting protein pairs but very rarely between non-interacting pairs, and hence should be useful for predicting whether two proteins interact. We use the pseudo-likelihood method to predict contacts across protein–interfaces for 36 evolutionarily conserved complexes of unknown structure and present structure models for four of the complexes particularly well constrained by these data.

Results

For a single protein family, it is straightforward to generate a multiple sequence alignment and subsequently identify covarying residue pairs. To identify covarying residue pairs between two proteins A and B is not as easy: only organisms that contain an ortholog of protein A and protein B contribute, and in generating the alignments the protein A and protein B sequences for each organism must be properly paired. To simplify the ortholog identification problem, we focus on pairs of genes with conserved chromosomal locations separated in the genome by fewer than 20 other annotated genes. We then build GREMLIN global statistical models for sequences in the paired protein families. The models have ‘one-body’ parameters for each amino acid at each position in the two proteins, and ‘two-body’ parameters for each pair of amino acids at each pair of positions in the two proteins. These parameters are obtained by maximizing the pseudo-likelihood of the observed sequence pairs,

rather than their likelihood, which makes the quite formidable estimation tractable. In the following sections, we investigate the structural contexts of residue pairs with large values of these two-body coupling parameters

Residue–residue covariation in the bacterial 50S ribosomal unit

We began by studying residue–residue coupling parameters in the bacterial 50S ribosomal subunit—the largest evolutionarily conserved bacterial multiprotein complex with an atomic resolution structure. For each individual protein in the complex, we constructed multiple sequence alignments by querying the UniProt sequence database (Wu *et al.*, 2006) for homologous sequences. For every pair of proteins in the complex, we then constructed a paired multiple sequence alignment ('Materials and methods'). For each such paired alignment, we built a GREMLIN global statistical model, computed normalized coupling strengths from the two body coupling parameters, and ranked inter-protein residue pairs based on these scores ('Materials and methods'). A coupling strength larger than one indicates higher than average coupling between two residues.

We find that in the 50S ribosomal subunit only a small fraction of residue pairs coevolve, as indicated by coupling strengths (y axis of **Figure 1A**) greater than 1.5. Remarkably, the two residues in

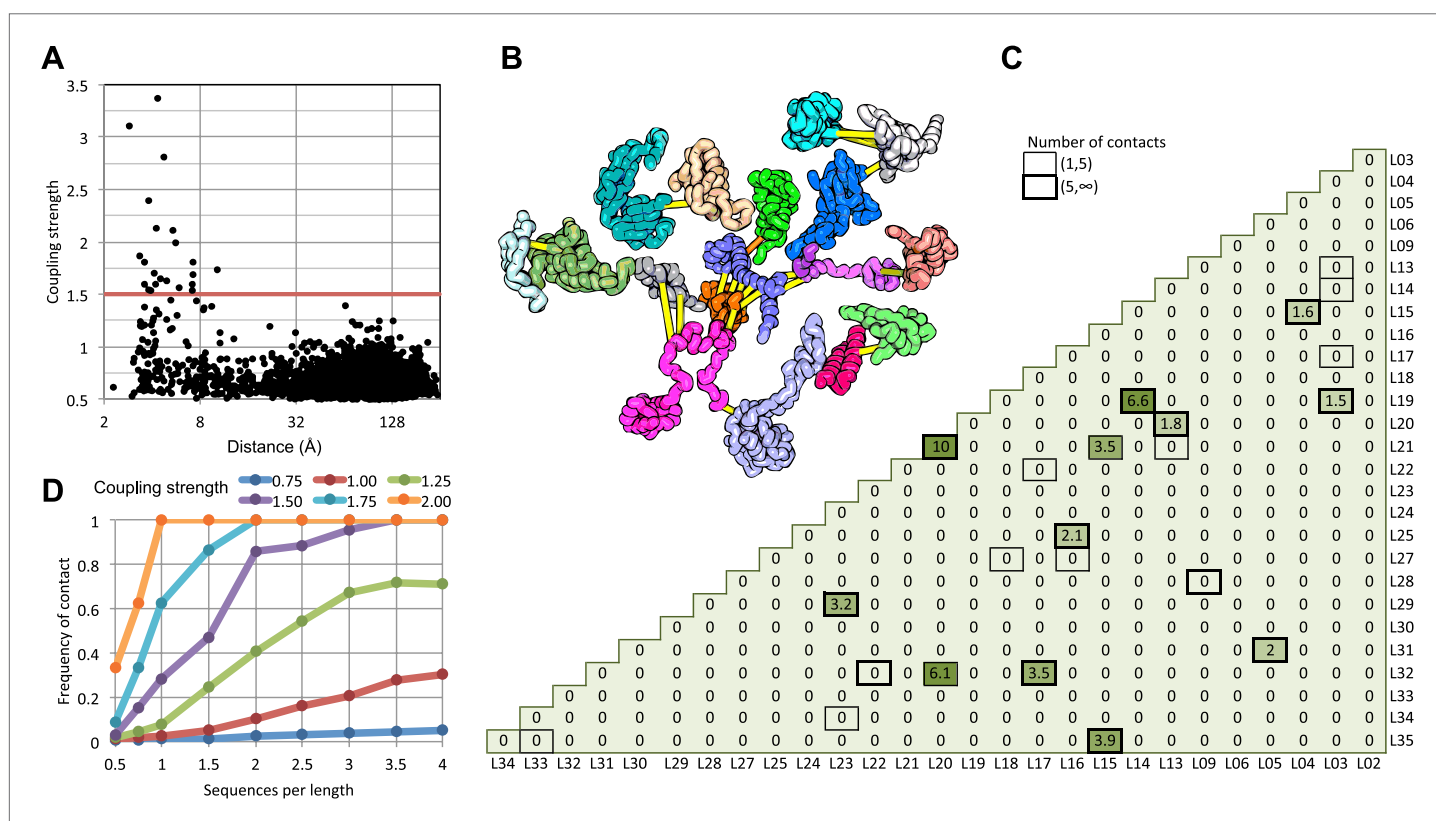


Figure 1. Residue pairs with high normalized coupling strengths are in contact in the 50S ribosomal subunit. **(A)** Coupling strengths and inter-residue distances for each residue pair in the 50S subunit (black dots). Residue pairs with coupling strength greater than 1.5 are nearly always less than 8 Å apart. **(B)** Locations of coevolving (high coupling strength) residue pairs in the protein component of the 50S subunit. The monomers have been pulled apart slightly for clarity. Lines connect residue pairs with coupling strength greater than 1.5; yellow, distance less than 8 Å; orange, distance less than 12 Å. **(C)** Protein pairs with strong inter-residue covariation (colors) make contact in the three-dimensional structure (black boxes). For each protein pair, the sum of the coupling strength greater than 1.5 for each pair of 50S subunit proteins is indicated; black boxes indicate contacts in the crystal structure. **(D)** Dependence of contact prediction accuracy on coupling strength and the number of sequences in the alignments. For each of the indicated coupling strength cutoffs (colors), the frequency of contact in the 50S structure (y axis) was computed for sub-alignments with different sequence depths (x axis). DOI: 10.7554/eLife.02030.003

The following figure supplements are available for figure 1:

Figure supplement 1. Determining GREMLIN scores from normalized coupling strengths.

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each of these pairs are almost all within 8 Å of each other in the 50S crystal structure (**Figure 1A**) and all are within 12 Å. The locations of the covarying residue pairs in the 50S structure (with the individual proteins pulled apart for clarity) are shown in **Figure 1B**; yellow lines indicate distances less than 8 Å and orange lines, distances less than 12 Å. For the 50S ribosome, the GREMLIN model was built using sequence data from ~1500 non-redundant genomes; **Figure 1D** suggests that for complexes with such large numbers of aligned sequence, residue–residue interactions across interfaces can be predicted with quite high confidence based on amino acid sequence covariation.

For a large protein–protein complex, can the sum of the coupling strengths between pairs of proteins in the complex be used to distinguish directly interacting and non-interacting protein pairs? In the 50S subunit, every pair of proteins with summed coupling strengths (numbers in **Figure 1C**) greater than 1.5 interacts with each other (boxes in **Figure 1C**). There are, however, several instances of protein pairs that contact in the 50S subunit for which no covariance is observed; clearly not every interaction will be identified by the sum of the coupling strengths, for example between two proteins that are held together primarily by the ribosomal RNA.

How many aligned sequences are required for accurate contact prediction? To assess the dependence on alignment depth, we generated paired sub-alignments with varying numbers of sequences for every pair of 50S proteins and recomputed coupling strengths for each sub-alignment. For each alignment depth, we calculated the fraction of residue pairs within 12 Å for different ranges of coupling strengths. We find that the greater the number of aligned sequences, the lower the value of the coupling strength above which residue pairs are likely to be in contact in the structure (**Figure 1D**). For example, if the number of aligned sequences is greater than the sum of the lengths of the two proteins, residue–residue contact predictions are likely to be accurate if the coupling strength is 2 or greater (**Figure 1D**: orange dots), while if there are twice as many sequences, contact predictions are accurate above a coupling strength of 1.5 (the cutoff shown in **Figure 1A**). A sigmoidal function of the coupling strength and the number of sequences per position in the complex accurately fits the observed contact frequency data ('Materials and methods' and **Figure 1—figure supplement 1**); we refer to the fitted values as GREMLIN scores for the remainder of the paper.

Bacterial complex benchmark

We next generated paired-alignments for all *E. coli* gene-pairs that had conserved intergenic distances across genomes deposited in the UniProt ('Materials and methods'). As the 50S results (**Figure 1D**) suggested that alignment depths greater than the average of the lengths of the two proteins were required for accurate prediction, we focused on paired alignments with at least this number of sequences—1126 gene pairs in total excluding the ribosomal proteins. For each of these 1126 pairs, we generated GREMLIN global statistical models and determined the coupling strength for each residue pair.

For 64 of the 1126 gene pairs, at least one pair of residues had GREMLIN score >0.85. For 28 of the 64 pairs three-dimensional structures have been determined experimentally, and the locations of the residue pairs with GREMLIN score >0.6 for several of these complexes are shown in **Figure 2A** (pairs within 8 Å are in yellow, between 8 Å and 12 Å in orange, and greater than 12 Å, in red). Almost all pairs with GREMLIN scores greater than 0.6 are in contact in the complex structures, with the notable exception of the NADH dehydrogenase subunits (**Figure 2B**). The complex is thought to undergo a cascade of conformational changes during electron transfer (**Baradaran et al., 2013**); the high GREMLIN score contacts not made in the solved structure may provide insight into the nature of these changes. As observed for the 50S complex (**Figure 1C**), the existence of one or more high GREMLIN scores between two proteins provides evidence that the proteins interact: 44% (28/64) of the protein pairs with high GREMLIN scores form a complex which has been solved crystallographically compared to 8% (78/1126) over the whole set.

Contact predictions for complexes of unknown structure

The results with the 50S ribosome and the protein pairs in the benchmark suggest that interactions can be accurately predicted across protein–protein interfaces given a sufficient number of aligned sequences. In **Figure 3**, we provide residue–residue contact predictions for the 36 of the 64 complexes with currently unknown structure (the *E. coli* gene sequences were clustered, and hence each complex may represent multiple *E. coli* gene pairs). These predictions should contribute to the determination of the structures of these biologically important complexes.

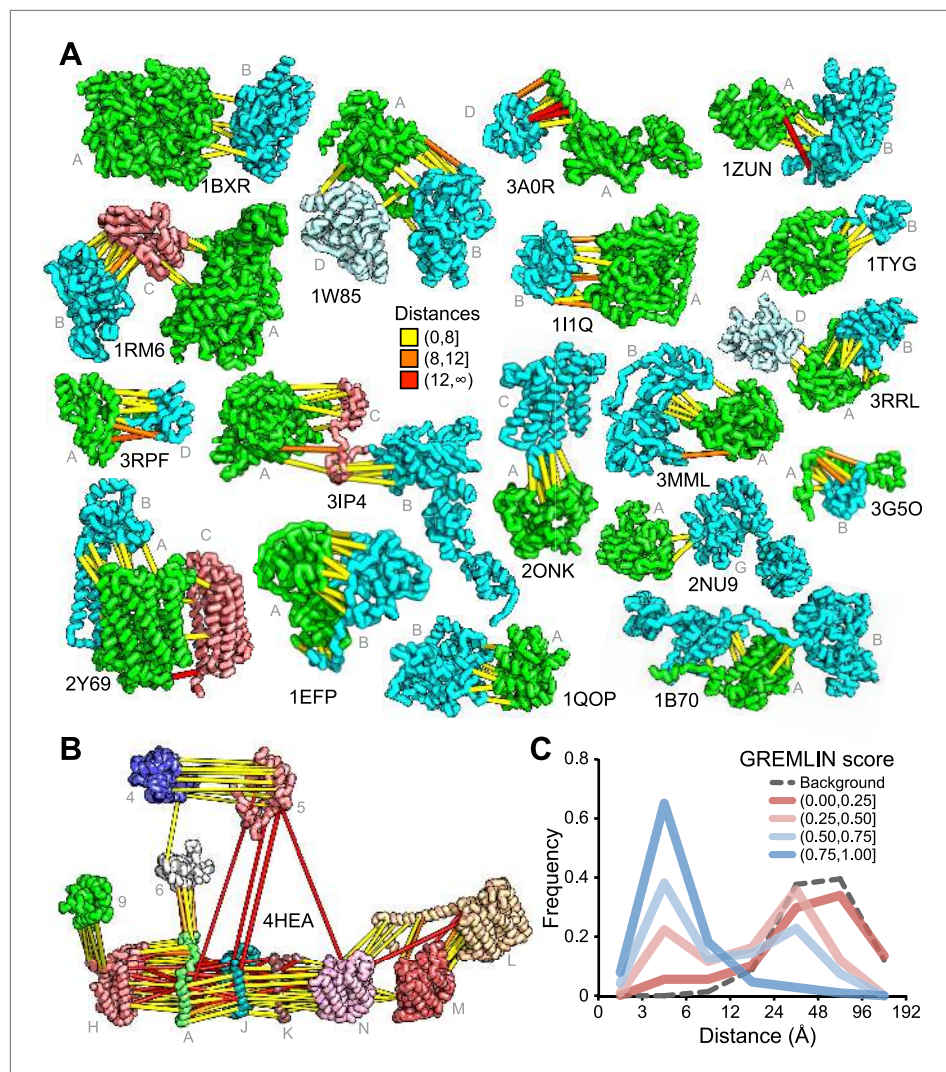


Figure 2. Residue covariation in complexes with known structures. **(A)** Residue-pairs across protein chains with high GREMLIN scores almost always make contact across protein interfaces in experimentally determined complex structures. All contacts with GREMLIN scores greater than 0.6 are shown; the structures are pulled apart for clarity. Labels are according to chains in the PDB structure. **(B)** Complex I of the electron transport chain has an unusually large number of highly co-varying inter residue pairs not in contact in the crystal structure of 4HEA; these contacts may be formed in different state of the complex. Residue pairs within 8 Å are in yellow, between 8 Å and 12 Å in orange, and greater than 12 Å, in red. Distances are the minimal distances between any side chain heavy atom. Labels are according to chains in 4HEA. **(C)** Dependence of inter-residue distance distributions on GREMLIN score. All residue–residue pairs between subunits in the benchmark set were grouped into four bins based on their GREMLIN score (colors), and the distribution of residue–residue distances (x axis) within each bin computed from the three-dimensional structures. See [Figure 2—source data 1](#) for the table of all the interfaces used in the calculation.

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The following source data are available for figure 2:

Source data 1. PDB benchmark set.

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From contacts to structural models

Are the predicted contacts useful in assembling models of the protein complex from models of each component? We evaluated this on a docking test set containing 18 protein complexes from the benchmark set where at least one component (or a close homolog) had a known structure in the apo form ('Materials and methods', docking test-set). We developed a docking protocol that used the predicted contacts as distance restraints and sampled the space of physically plausible

structures to generate models of the protein–protein complex. The model with the best restraint score had an interface that was within 4 Å (in root mean square deviation) of the native interface in 14 of the 18 cases and had more than half the native contacts in 16 of the 18 cases (**Figure 4A**, **Figure 4—figure supplement 1**). Two of the cases in which the iRMSD (interface root-mean-square deviation) was the highest (bottom of table in A) are illustrated in **Figure 4B–C**: the high iRMSD is due to large changes in the conformation of one of the monomers upon binding; despite these changes the binding interface is reasonably accurately identified. Conformational changes that hinder the rigid-body docking protocol from sampling the bound conformation also occurred for thiazole synthase/sulfur carrier and phenylalanyl-tRNA synthase with iRMSD of 4.8Å and 4.3Å, respectively. In **Figure 4D**, a second energy minimum corresponds to a second interface in the complex with a different homooligomer subunit. In the absence of conformational changes, predicted contact guided docking is very accurate. The same protocol, on a positive control set of known bound structures of 41 protein-pairs (including 15 protein-pairs from the NADH electron transport complex), generated models that were within 2 Å of the native complex structure in 38 cases and within 4 Å in all but one case (**Figure 4—source data 1**, **Figure 4—figure supplement 2**).

Taken together, these results suggest that in cases with small conformational change, the docking protocol can recover the entire interface to high accuracy and in cases where binding is accompanied by a large conformational change, the protocol recovers the largest intact and/or unobstructed interface.

Of the complexes with unknown structure listed in **Figure 3**, we selected four cases with two or more high GREMLIN score (≥ 0.6) contact predictions across the interface that had experimentally determined structures for most of the subunits ('Materials and methods') and generated structural models of the complexes. These models provide the basis for formulating hypotheses about the structure/function of the complex, but we emphasize they are not experimentally determined structures; in particular the assumption in the modeling procedure that there are not large backbone rearrangements could be incorrect—in such cases the overall organization of the complex is still likely to be correct but the details of the interfaces could be considerably in error.

The TRAP complex

The tripartite ATP-independent periplasmic (TRAP) transporters are composed of three proteins: two integral membrane proteins YIAM and YIAN, and one periplasmic protein YIAO (**Mulligan et al., 2011**). The structure of the periplasmic domain is known, but the membrane portion is unknown. To generate a model of the three-dimensional structure of the complex, we built YIAM models using Rosetta de novo structure prediction (**Simons et al., 1999**; **Raman et al., 2009**) guided by the intramonomer predicted contacts, and models for YIAN and YIAO using RosettaCM comparative modeling. For YIAN the homologous structure of 4f35 (**Mancusso et al., 2012**) was used. The three monomer structure models were then assembled using PatchDock (**Duhovny et al., 2002**) and RosettaRelax (**Conway et al., 2014**) guided by the predicted intersubunit contacts ('Materials and methods'). In the resultant model of the complex (**Figure 5**), YIAO interacts with both of the membrane components; this is supported by a number of intersubunit contacts (yellow lines).

Tripartite efflux system

Tripartite efflux complexes span both the inner and outer membrane, and are widely used in bacteria to pump toxic compounds out of the cell. The mode of interactions between the outer membrane factor and the membrane fusion protein is unresolved, with reports suggesting either a tip-to-tip interaction, the insertion of one into the other, or a multistage interaction with an initial tip-to-tip interaction, followed by sliding one through the channel of the other (**Long et al., 2012**). We generated homology models for the subunits based on the alignments to 1yc9 (**Federici et al., 2005**) and 3fpp (**Yum et al., 2009**) and docked them to generate models of the multidrug resistance protein complex. The predicted residue–residue contacts for this family of complexes support the tip-to-tip interaction (**Figure 4**; yellow lines); the coevolution data did not provide any evidence to support the insertion model.

Pyruvate formate lyase-activating enzyme complex

Pyruvate formate-lyase (PFL) catalyzes the reaction of acetyl-CoA and formate from pyruvate and CoA in the Fermentation pathway. Formate acetyltransferase 1 or Pyruvate formate-lyase 1 (PFLB) is activated by Pyruvate formate-lyase 1-activating enzyme (PFLA). The structure of the complex is unknown, but the structures of the individual proteins have been solved (PDB ids: 3c8f [**Becker and Kabsch, 2002**] and 1h16 [**Vey et al., 2008**]). We carried out rigid body docking calculations with these two

YIAM_YIAN (3.9L)	YOHJ_YOHK (1.9L)	FLGB_FLGC (2.4L)	PTPC1_PTPD (1.2L)	APPB_APPC (1.2L)
21_F 246_F 1.00	64_C 75_Y 1.00	34_D 13_A 1.00	130_K 58_K 1.00	91_I 474_S 1.00
91_I 25_L 1.00	61_N 79_E 1.00	34_D 107_S 0.99	210_L 189_I 1.00	101_P 68_A 0.99
42_R 52_D 0.97	115_S 94_I 1.00	110_A 117_E 0.98	34_I 122_C 0.98	95_C 481_L 0.95
14_A 268_L 0.93	79_G 38_M 1.00	98_N 11_G 0.98	126_D 28_R 0.98	88_V 473_F 0.94
87_L 26_L 0.89	110_V 101_V 0.99	121_S 129_T 0.96	232_G 180_L 0.93	95_C 477_M 0.92
59_D 273_K 0.89	106_V 101_V 0.99	34_D 103_V 0.96	123_T 27_E 0.86	342_I 98_M 0.90
17_S 245_A 0.87	99_S 115_A 0.97	113_S 108_A 0.89	180_H 258_S 0.85	91_I 477_M 0.77
28_Y 234_L 0.84	83_M 42_I 0.96	27_A 114_A 0.88	125_A 32_G 0.85	332_S 90_D 0.70
43_Y 48_V 0.82	92_Q 116_S 0.93	117_Q 128_K 0.80	134_T 53_L 0.72	
10_A 268_L 0.79	104_T 184_A 0.89	111_D 122_V 0.78	125_A 61_L 0.66	RNFE_RNFG (1.8L)
64_Q 17_G 0.76	105_L 161_F 0.84	113_S 125_M 0.71	104_S 71_V 0.64	188_L 160_V 0.90
56_A 276_S 0.75	40_I 146_I 0.84	27_A 27_N 0.63	12_L 223_I 0.60	85_M 27_N 0.90
9_L 272_A 0.75	71_M 34_L 0.83	131_M 69_E 0.61		67_I 12_L 0.89
46_V 52_D 0.72	65_Y 79_E 0.81	23_Q 37_P 0.60	YADG_YADH (5.4L)	146_G 171_T 0.66
13_L 271_A 0.67	113_W 169_M 0.80	93_P 45_K 0.60	97_Q 22_R 0.94	
50_F 51_A 0.66	16_V 14_I 0.78		49_G 93_E 0.93	NRFC_NRFD (1.2L)
95_L 47_L 0.62	74_L 65_L 0.78	FLIP_FLIQ (2.2L)	106_Y 99_P 0.84	130_Y 92_S 0.93
46_V 51_A 0.60	82_V 42_I 0.76	185_I 84_L 0.99	90_N 15_K 0.70	129_Q 76_T 0.79
10_A 269_I 0.60	101_A 157_L 0.75	54_I 55_I 0.90	97_Q 18_H 0.68	124_L 87_H 0.69
YIAN_YIAO (3.6L)	39_S 70_V 0.72	229_V 77_V 0.89	82_L 97_V 0.67	154_F 55_T 0.63
168_S 196_Y 1.00	115_S 176_A 0.70	203_V 24_L 0.68	104_G 14_A 0.64	
296_E 62_K 0.89	106_V 105_T 0.68	205_M 66_G 0.67		TOLQ_TOLR (4.9L)
166_S 221_E 0.72	115_S 179_L 0.66	213_P 55_I 0.66	RBSA_RBSC (5.6L)	149_I 27_V 0.91
237_P 197_T 0.59	100_C 150_C 0.65	243_F 72_L 0.65	365_S 188_Y 1.00	177_A 27_V 0.91
YIAM_YIAO (5.0L)	18_I 72_A 0.64	149_L 67_P 0.64	112_K 188_Y 0.97	177_A 25_L 0.90
33_S 186_N 0.56	109_L 169_M 0.63	188_T 80_L 0.63	301_A 211_L 0.95	177_A 26_L 0.84
	110_V 97_I 0.60		93_Q 204_G 0.79	139_Y 19_V 0.83
	110_V 213_I 0.60	FLGH_FLGI (0.6L)	371_E 197_R 0.66	142_L 20_P 0.77
PFLA_PFLB (0.7L)		52_F 133_V 0.99	105_E 198_Y 0.64	179_A 24_V 0.76
66_F 650_T 0.92	CYOC_NUOK (3.8L)	82_L 257_S 0.59		167_V 27_V 0.76
59_E 643_K 0.67	49_V 80_L 0.96		APAG_PPIC (0.8L)	179_A 26_L 0.73
62_T 570_N 0.54	83_Y 29_L 0.68	FLHB_FLIR (1.1L)	68_V 73_E 0.97	177_A 18_I 0.69
70_S 688_G 0.27	75_L 73_A 0.68	181_A 188_L 0.98		133_V 102_N 0.67
	65_E 89_N 0.65	185_A 192_T 0.64	DDPB_DDPC (8.7L)	177_A 19_V 0.67
MDTP_MDTN (2.4L)	101_W 29_L 0.61		148_W 268_G 1.00	156_L 19_V 0.66
431_A 146_E 1.00	42_I 62_Y 0.60	FLIP_FLIR (1.5L)	279_L 231_T 1.00	142_L 22_L 0.66
223_S 146_E 0.96	ATP6_ATPF (2.3L)	78_L 187_A 0.90	20_G 135_I 0.99	200_N 11_D 0.63
227_H 142_P 0.90	74_K 34_E 1.00	97_T 13_L 0.71	327_Y 180_L 0.97	134_G 19_V 0.59
231_A 150_S 0.74	149_V 10_Q 0.93		327_Y 177_G 0.93	
431_A 148_F 0.45	77_T 33_I 0.92	MLAD_MLAE (1.4L)	233_E 292_K 0.91	ATKA_ATKC (0.7L)
435_R 142_P 0.44	53_G 13_A 0.86	14_L 13_G 0.93	221_R 283_D 0.91	190_I 30_G 1.00
	155_L 11_A 0.78	14_L 24_G 0.76	229_E 212_P 0.83	448_P 91_N 1.00
METI_METQ (2.0L)	50_V 13_A 0.72	8_I 6_L 0.65	28_I 147_A 0.77	249_A 175_V 0.99
80_R 193_D 0.99	255_I 139_S 0.68	58_V 6_L 0.65	222_Q 217_Q 0.75	186_A 26_T 0.99
80_R 192_D 0.91	243_I 20_F 0.64	65_V 219_F 0.63	13_G 127_A 0.73	183_L 26_T 0.84
87_I 182_L 0.70	49_S 10_Q 0.64		268_L 137_L 0.67	249_A 167_L 0.79
87_I 66_F 0.56	263_Y 21_C 0.64	CCMC_CCME (0.8L)	126_R 293_A 0.65	11_T 170_Y 0.78
174_Q 67_N 0.50	239_V 16_L 0.63	49_Q 104_R 0.95	143_S 279_N 0.65	198_F 24_L 0.63
87_I 75_A 0.43	111_W 10_Q 0.62		140_T 272_L 0.63	238_A 83_G 0.62
		CCMA_CCMB (1.3L)	24_I 135_I 0.62	244_F 55_I 0.60
UMUC_UMUD (1.0L)	FECL_FECR (3.3L)	95_E 16_R 0.87	233_E 185_Y 0.59	FEOA_FEOB (0.9L)
415_S 38_I 1.00	162_E 14_R 0.96		ENGB_NDK (1.0L)	41_R 378_F 0.94
421_V 54_V 0.86	133_Q 57_R 0.95	MREB_NIFU (0.6L)	193_L 60_F 0.87	
132_H 76_S 0.85	158_A 18_H 0.61	21_N 57_R 0.97		YEJA_YEJB (3.1L)
404_R 33_Y 0.72		149_I 108_I 0.72	CLPP_CLPX (1.3L)	466_D 325_R 0.90
183_D 127_V 0.66	FTSI_FTSW (2.0L)		96_F 271_G 0.95	
78_A 34_V 0.60	39_L 313_V 0.98	QMCA_YBBJ (1.5L)	40_E 267_G 0.72	
	42_V 312_V 0.94	145_E 115_R 0.91		
IF1_SECY (1.4L)	47_V 309_Y 0.80	36_R 113_H 0.72		
32_V 343_A 0.89	43_A 309_Y 0.74			
	34_A 357_L 0.62			

Figure 3. Predicted residue–residue interactions across protein interfaces of unknown structure. Strongly co-evolving residue pairs for complexes without known structure that had at least one prediction with GREMLIN score greater than or equal to 0.85. Each row shows the residue pairs, their sequence identity and the GREMLIN score. Structure models for complexes highlighted in red are shown in **Figure 5**. Full dataset is provided with the deposited data.

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proteins guided by GREMLIN predictions. Interestingly, the region that undergoes conformational change in the activating enzyme upon substrate binding (3c8f -> 3cb8 [**Becker and Kabsch, 2002**]) is in the region we predict to be in contact with PFL.

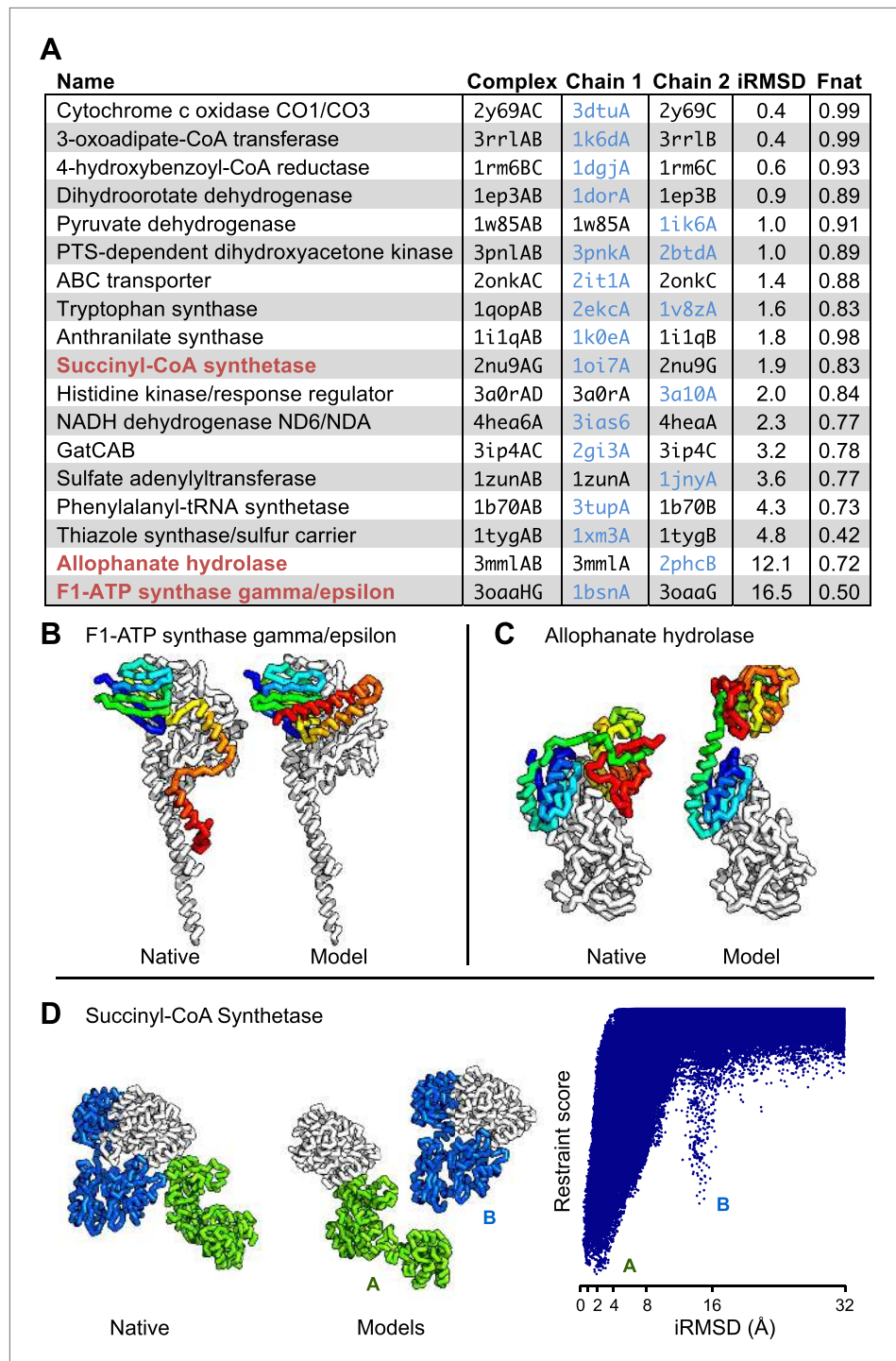


Figure 4. Contact guided protein–protein docking on a benchmark set of 18 protein complexes. **(A)** Structure models for each complex were generated by docking structures of its constituents, at least one of which (blue) was not from the structure of the complex guided by coevolution derived distance restraints. The interface C-alpha RMSD (iRMSD) of the structural model with the lowest energy to the experimentally determined structure and the fraction of native contacts are shown. Structure models for cases in red are shown in **B** and **C** and **D**. **(B and C)** Comparison between native and docked structure for the two largest failures in the benchmark: the large iRMSD is due to large conformational changes in the monomers upon docking but the interface is still modeled correctly in Figure 4. *Continued on next page*

Figure 4. Continued

the region not involved in conformational change. (D) Multiple minima in the docking landscape (right) correspond to distinct interfaces in the complex (left).

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The following source data and figure supplements are available for figure 4:

Source data 1. Bound set.

DOI: [10.7554/eLife.02030.009](https://doi.org/10.7554/eLife.02030.009)

Figure supplement 1. Docking landscapes showing iRMSD (x-axis) vs GREMLIN restraint score (y-axis).

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Figure supplement 2. Bound set.

DOI: [10.7554/eLife.02030.011](https://doi.org/10.7554/eLife.02030.011)

D-methionine transport system

D-methionine transporter is an ATP-driven transport system that transports methionine. We docked the *E. coli* structure of METI (3tui, chain A and B, [Johnson et al., 2012](#)) with a RosettaCM model of METQ based on 3k2d ([Yu et al., 2011](#)). The resulting docked model is consistent with the top ranked GREMLIN predictions ([Figure 5](#)).

Discussion

Our results demonstrate unequivocally that there is strong selective pressure at protein–protein interfaces beyond simple residue conservation, and that co-evolving residue pairs are nearly always in contact in the protein complex. Not all contacting residues across protein interfaces likely co-evolve nor all protein–protein interfaces. Nevertheless, as illustrated in [Figures 1 and 2](#), there is clearly sufficient coevolutionary signal to significantly constrain models of a large number of protein complexes.

There is a notable contrast in the utility of intra-monomer and intersubunit predicted contacts for structure modeling. We found previously ([Kamisetty et al., 2013](#)) that contacts could be predicted with high accuracy for monomeric proteins, provided there were sufficient aligned sequences, but in such cases there was almost always already a structure of a family member from which comparative models could be built, limiting the utility of the predicted contacts in structure prediction (Though predicted contacts can be useful in modeling allosteric changes in protein structures [[Hopf et al., 2012](#); [Morcos et al., 2013](#)]). In contrast, here we find that more than half of the complexes for which the protein families of the constituent subunits are sufficiently large for accurate contact prediction do not currently have three-dimensional structures. Hence, while predicted contacts can be very accurate for both monomeric globular proteins and for protein–protein complexes, they are more useful for structure modeling for the latter due to the much poorer representation of protein complexes in the PDB.

While our approach of constructing a global statistical model from paired sequence alignments is generally applicable to any taxa, the current study focuses on prokaryotes and mitochondria. Doing so allows us to largely avoid the problem of distinguishing between paralogs by exploiting the operon architecture of bacterial genomes ([Jacob et al., 2005](#)). Constructing paired-sequence alignments for more complex genomic architectures is more involved and requires the ability to distinguish orthologs from paralogs, the subject of active research ([Remm et al., 2001](#); [Datta et al., 2009](#)). Protocols for generating paired sequence alignments more generally are an important area for development in this area.

Materials and methods

Individual alignment generation

Multiple sequence alignments were generated for each of the 4303 *E. coli* protein genes as identified by EcoGene 3.0 ([Zhou et al., 2013](#)) using HHblits (-n 8 -e 1E-20 -maxfilt ∞ -neffmax 20 -nodiff -realign_max ∞), and HHfilter (-id 100 -cov 75) in the HHSuite (version: 2.0.15, [Remmert et al., 2011](#)). To reduce redundancy, we constructed HMMs from each MSA and clustered genes based on the HHD ([Kamisetty et al., 2013](#)), a measure of HMM–HMM similarity: a pair of genes was assigned to the same cluster if the HHD is less than 0.5. This procedure resulted in 2340 non-redundant gene clusters.

For the benchmark set, a new alignment was generated using the sequence associated with each PDB. For the 50S ribosome and NADH dehydrogenase, we used *Thermus thermophilus* HB8 sequences

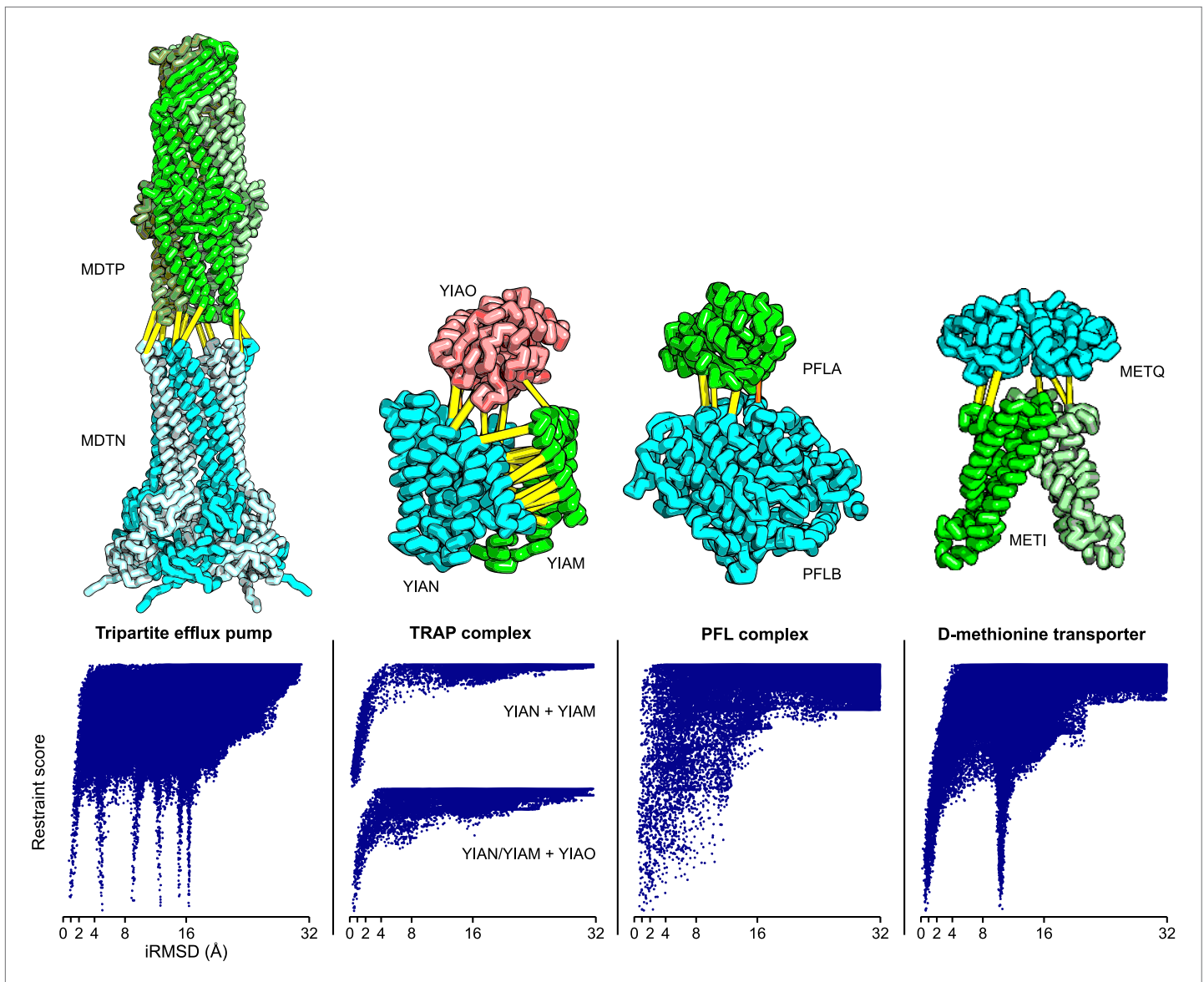


Figure 5. Structure models for complexes with unknown structures. Residue pairs with GREMLIN scores ≥ 0.60 are connected by yellow bars; the structures are pulled apart for clarity. For METQ-METI and PFLA-PFLB GREMLIN scores ≥ 0.3 are shown. For each docking calculation the docking energy landscape is shown, with iRMSD to the selected model on the x-axis. The multiple minima correspond to permutations of the labels on the subunits of the homo-oligomer complex. Predicted structures of each complex are provided with the deposited data.

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from PDB structures 3uxr ([Bulkley et al., 2012](#)) and 4hea ([Baradaran et al., 2013](#)) respectively. For paralogous NADH dehydrogenase chains L, M, and N, we used an e-value of $1E-60$ in the alignment generation protocol. In addition to complexes from the *E. coli* analysis, we also include the GatCAB amidotransferase complex in our benchmark set, using sequences from the PDB structure 3ip4 ([Nakamura et al., 2010](#)). For cases where the PDB sequence length was much longer than average coverage, we modified the coverage filter to 50% of query. The sequences were then realigned using clustal omega v1.2 (--iterations 2 --full-iter) ([Sievers et al., 2011](#)). Residues not present in the query sequence were dropped from subsequent analysis.

Paired alignment generation

We construct alignments of paired protein sequences $[x_1, x_2, \dots, x_p; x_{p+1}, \dots, x_{p+q}]$ from the same genome with positions 1:p and p+1:p+q corresponding to the first and second proteins respectively. We refer to such a multiple sequence alignment of paired sequences as a paired alignment.

For gene families with a single copy in each genome such as the ribosomal proteins, constructing paired alignments is straightforward as sequence pairs from the same genome can simply be concatenated. While the process of generating paired alignments in general is complicated in the presence of multiple paralogs of a gene in a single genome, in prokaryotes, co-regulated genes are often co-located on the genome into operons. We exploit this property to avoid paralogous genes when creating paired sequences by restricting to gene pairs that have small, conserved intergenic distances. A similar approach was used to construct a database of fusion proteins in prokaryotic genomes (*Suhre and Claverie, 2004*). Defining Δ_{gene} as the number of annotated genes between a gene pair, we only consider pairs with Δ_{gene} conserved in 60% of genomes and less than 20. To allow for ambiguity in annotation, if the second or third most common intergenic distance is within 1 of the mode, these gene-pairs are included in the conservation calculation. Given that most UniProt accession IDs are serially assigned in a genome (UniProt Accession), Δ_{gene} can be rapidly evaluated by looking at the difference in accession ids. The paired alignment is then filtered to reduce redundancy to 90% sequence identity and to remove positions that have more than 75% gaps.

Identification of protein complex structures

To identify protein pairs in the same complex structure, a HMM was constructed for each *E. coli* protein using hmmbuild from the already generated HHblits alignments. We then used hmmsearch to scan PDB sequences in the S2C database (Wang et al.; Both hmmbuild and hmmsearch are part of the HMMER v3.1b package [*Eddy, 2009*]). Only hits with e-value less than $1E-10$ were considered. Protein pairs found in the same complex structure (PDB file) were considered to be in contact if a C α atom in one structure was within 12 Angstroms of a C α atom in the other.

Gremlin model construction from paired alignments

GREMLIN constructs a global statistical model of the paired alignment, assigning a probability to every amino-acid sequence in the paired alignment:

$$p(X_1, X_2, \dots, X_p; X_{p+1}, \dots, X_{p+q}) = \frac{1}{Z} \exp(\sum_i^{p+q} [v_i(X_i) + \sum_{j=1}^{p+q} w_{i,j}(X_i, X_j)])$$

where, the v_i are vectors encoding position-specific amino-acid propensities and the w_{ij} are matrices encoding amino-acid coupling between positions i and j . These parameters are obtained from the aligned sequences by maximizing the regularized pseudo-likelihood (*Balakrishnan et al., 2011*) of the alignment as described in (*Kamisetty et al., 2013*):

$$v, w = \arg \max \sum_1^N \sum_1^{p+q} \log P(X_i | X_1 \dots X_{i-1} X_{i+1} \dots X_{p+q}) + R(v, w)$$

where, each term in the summation is a conditional distribution capturing the probability of a particular amino-acid at a position in the context of the entire protein sequence and $R(v, w)$ is a regularization term to prevent over-fitting.

Previous approaches (*Morcos et al., 2011; Jones et al., 2012*) estimated v, w using an approximate moment matching approach (*Kamisetty et al., 2013*) by inverting a generalized covariance matrix. These rely on a Gaussian-like approximation to the global partition function. Unlike these approaches, estimation via the pseudo-likelihood avoids this approximation relying instead on local partition functions (*Balakrishnan et al., 2011; Ekeberg et al., 2013; Kamisetty et al., 2013*). The resulting global optimization problem can be efficiently solved using standard convex optimization techniques and provides estimates for each vector v_i and matrix w_{ij} (*Kamisetty et al., 2013*).

Ranking residue pairs with gremlin scores

To reduce the w_{ij} matrices to single values reflecting the strength of the coupling between positions i and j , we first compute s_{ij} , their vector 2-norm (the square root of the averages of the squares of the individual matrix elements). We correct for differences in s_{ij} due to sequence variability at different positions using the row and column averages of these values:

$$s_{ij}^{\text{corr}} = s_{ij} - \frac{\langle s_{ik} \rangle_k \langle s_{kj} \rangle_k}{\langle s_{kl} \rangle_{kl}}$$

where brackets indicate averages taken over the indices outside the brackets in a manner similar to that of Average Product Correction (APC, *Dunn et al., 2008*). Unlike the APC, we account for

differences in the rates of evolution in the two protein families by computing the averages only over the positions of the proteins corresponding to positions i and j : if i and j are both in the first (second) protein, the averages are computed over the positions in the first (second) protein; if i is in the first protein and j in the second, the column average is computed only over the positions of the first protein and the row average, only over the positions of the second protein. We then compute a normalized coupling strength, ncs_{ij} , by dividing the s_{ij}^{corr} by the average of the top $3L/2$ s_{ij}^{corr} values across the two proteins (since there are roughly $3L/2$ contacts for a protein of length L [Kamisetty et al., 2013; SI]).

As illustrated in **Figure 1D**, the relation between normalized coupling strength and contact frequency varies with the ratio of the number of aligned sequences to the length of the protein complex. We also observed that residues were more frequently in contact for a given coupling strength when the top score for that complex was high. To account for these dependencies, we constructed a model that estimates the probability of being in contact based on the bacterial 50S ribosomal complex:

$$\text{GremlinScore}(x, N/L) = 1/(1 + \exp(-\sigma(x - \mu)))$$

where

$$\mu = m^{N/L+1} + c$$

and x is $\sqrt{ncs_{ij}}$ for the top scoring contact in each complex and $\sqrt{ncs_{ij}}$ scaled by the Gremlin score of the top contact in all other cases. The values of m , c , and σ (0.47, 0.96, and 9.77 respectively) were determined by a non-linear fit to the observed frequencies in the 50S ribosomal data from **Figure 1D**. This function accurately accounts for the observed contact frequencies (**Figure 1—figure supplement 1**).

Conversion of gremlin scores to distance restraints

We converted coupling strengths into residue-pair specific distance restraints and included them in the Rosetta structure prediction program. We use sigmoidal distance restraints of the form:

$$\text{restraint}(d) = \frac{\text{weight}}{1 + \exp(-\text{slope}(d - \text{cutoff}))} + \text{intercept} \quad (1)$$

where, d is the distance between the constrained atoms and the weight is proportional to ncs_{ij} . The restraints were introduced between $C\beta$ atoms ($C\alpha$ in the case of glycine) in the reduced-atom representation of Rosetta (centroid mode) and as ambiguous distance restraints (Lange et al., 2012) between side-chain heavy atoms (cutoff of 5.5 and slope of 4) in the full-atom stage of Rosetta. For the centroid mode, restraints used the amino acid pair specific $C\beta$ - $C\beta$ cutoff and slopes, as described in Kamisetty et al., 2013 SI Table III. These distance restraints supplement the Rosetta all atom energy; the combination ensures the sampling of physically realistic structures consistent with the contact predictions.

Comparative modeling

Comparative models were built using RosettaCM (Song et al., 2013) based on alignments to homologous structures generated using HHsearch (Remmert et al., 2011). For proteins that had missing density in regions predicted to be in contact, we used RosettaCM with co-evolution derived restraints to build the missing region before docking.

De Novo modeling

The Rosetta ab initio protocol consists of two stages: in the initial stage ('centroid') side-chains are represented by fixed center-of-mass atoms allowing for rapid generation and evaluation of various protein-like topologies; the second stage ('full-atom') builds in explicit side-chains and carries out all atom energy minimization (Simons et al., 1999; Raman et al., 2009). YIAM, a membrane protein, was modeled with the Rosetta membrane energy function (Yarov-Yarovoy et al., 2012, Barth et al., 2007). Strong repulsive interactions (Equation 1, weight: -100, cutoff: 35, slope: 2 and intercept: 100) were added between the center of the extracellular regions and the center of predicted intracellular regions, and strong attractive restraints (weight:100, cutoff:35, slope:2 and intercept: 0) within predicted intracellular regions and extracellular regions, effectively constructing a membrane-like sampling space. We used the consensus output of MESSA (Cong and Grishin, 2012) to predict trans-membrane regions. 100,000 models were generated and 20 models that best fit the restraints converged to a single cluster.

Docking test set

Jackhammer (part of HMMER v3.1b package; [Eddy, 2009](#)) was used to identify a subset of 18 complexes in the benchmark set where at least one of the proteins or a close homolog had a solved structure of its *apo* form. In cases where the structure was of a homologous protein (e-value < 1E-20) and where most of the interface residues were present, we generated a structural model of the target protein using comparative modeling. We only considered cases where at least one of the structures was unbound as the bound-bound docking problem is not representative of real world docking challenges ([Betts and Sternberg, 1999](#)). The positive control shown in [Figure 4—source data 1](#) was run on all protein-pairs from the benchmark set, where at least two predicted inter contacts had a high GREMLIN score (>0.6).

Complex assembly by protein–protein docking

For each inter restraint pair that is in the top 3/2L predictions, we used PatchDock v1.0, with clustering parameters (rmsd 0.5; discardClustersSmaller 0) ([Duhovny et al., 2002](#)) to generate an ensemble of conformations that were then scored using all the restraints. For tripartite efflux pump, the surface segmentation parameters were further modified (low_patch_thr 0; prune_thr 0.1; flat 1), to allow for more diverse interfaces. The top 5 models by restraint score were energy-minimized in cartesian space using both inter and intra restraints with cycles of minimization and side chain repacking using Rosetta as described in [Conway et al. \(2014\)](#). The best scoring model by restraint score was then selected.

For fraction of native contact (Fnat) and interface root-mean-squared deviation (iRMSD) calculation, the interface residue–residue contacts are those where the minimal distance between any heavy side-chain atom is less than 5 Å. The Fnat calculation is performed as described in [Kamisetty et al. \(2013\)](#) SI Table III.

All structural figures were drawn with PyMOL (The PyMOL Molecular Graphics System, Version 1.5.0.4 Schrödinger, LLC.).

Data Availability

The multiple sequence alignments used in the analysis and the full GREMLIN results for all the calculations described in the paper are provided at <http://gremlin.bakerlab.org/complexes/> along with a web-server for paired-alignment generation, coevolution analysis and contact prediction/Rosetta restraint generation. The paired-alignments along with the PDB coordinates of the predicted structures are also available at Dryad: [Ovchinnikov et al., 2014](#).

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Note added in proof

Two other studies of protein-coevolution using global statistical models have recently appeared: [Tamir et al., 2014](#), and [Hopf et al., 2014](#). These studies provide independent validation of the robustness of global statistical methods for prediction of protein–protein contacts.

Additional information

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Author contributions

SO, Conception and design, Acquisition of data, Analysis and interpretation of data, Drafting or revising the article, Contributed unpublished essential data or reagents; HK, Conception and design, Analysis and interpretation of data, Drafting or revising the article, Contributed unpublished essential data or reagents; DB, Conception and design, Analysis and interpretation of data, Drafting or revising the article

Additional files**Major datasets**

The following dataset was generated:

Author(s)	Year	Dataset title	Dataset ID and/or URL	Database, license, and accessibility information
Ovchinnikov S, Kamisetty H, Baker D	2014	Data from: Robust and accurate prediction of residue-residue interactions across protein interfaces using evolutionary information	http://dx.doi.org/10.5061/dryad.s00vr	Available at Dryad Digital Repository under a CC0 Public Domain Dedication.

The following previously published datasets were used:

Author(s)	Year	Dataset title	Dataset ID and/or URL	Database, license, and accessibility information
Reshetnikova L, Moor N, Lavrik O, Vassilyev DG	2000	PHENYLALANYL TRNA SYNTHETASE COMPLEXED WITH PHENYLALANINE	1B70; http://dx.doi.org/10.2210/pdb1b70/pdb	Publicly available at the RCSB Protein Data Bank.
Wilkens S, Capaldi RA	1998	SOLUTION STRUCTURE OF THE EPSILON SUBUNIT OF THE F1-ATPSYNTASE FROM ESCHERICHIA COLI AND ORIENTATION OF THE SUBUNIT RELATIVE TO THE BETA SUBUNITS OF THE COMPLEX	1BSN; http://dx.doi.org/10.2210/pdb1bsn/pdb	Publicly available at the RCSB Protein Data Bank
Thoden JB, Wesenberg G, Raushel FM, Holden HM	1999	STRUCTURE OF CARBA-MOYL PHOSPHATE SYNTHETASE COMPLEXED WITH THE ATP ANALOG AMPNP	1BXR; http://dx.doi.org/10.2210/pdb1bxr/pdb	Publicly available at the RCSB Protein Data Bank
Rebelo JM, Macieira S, Dias JM, Huber R, Romao MJ	2000	CRYSTAL STRUCTURE OF THE ALDEHYDE OXIDOREDUCTASE FROM DESULFOVIBRIO DESULFURICANS ATCC 27774	1DGJ; http://dx.doi.org/10.2210/pdb1dgj/pdb	Publicly available at the RCSB Protein Data Bank.
Rowland P, Larsen S	1997	DIHYDROOROTATE DEHYDROGENASE A FROM LACTOCOCCUS LACTIS	1DOR; http://dx.doi.org/10.2210/pdb1dor/pdb	Publicly available at the RCSB Protein Data Bank.
Roberts DL, Salazar D, Fulmer JP, Frerman FE, Kim JJ-P	1999	ELECTRON TRANSFER FLAVOPROTEIN (ETF) FROM PARACOCCUS DENITRIFICANS	1EFP; http://dx.doi.org/10.2210/pdb1efp/pdb	Publicly available at the RCSB Protein Data Bank.
Rowland P, Norager S, Jensen KF, Larsen S	2001	CRYSTAL STRUCTURE OF LACTOCOCCUS LACTIS DIHYDROOROTATE DEHYDROGENASE B. DATA COLLECTED UNDER CRYOGENIC CONDITIONS	1EP3; http://dx.doi.org/10.2210/pdb1ep3/pdb	Publicly available at the RCSB Protein Data Bank.

Becker A, Kabsch W	2002	PYRUVATE FORMATE-LYASE (E.COLI) IN COMPLEX WITH PYRUVATE AND COA	1H16; http://dx.doi.org/10.2210/pdb1h16/pdb	Publicly available at the RCSB Protein Data Bank.
Morollo AA, Eck MJ	2001	STRUCTURE OF THE COOPERATIVE ALLOSTERIC ANTHRANILATE SYNTHASE FROM SALMONELLA TYPHIMURIUM	1I1Q; http://dx.doi.org/10.2210/pdb1i1q/pdb	Publicly available at the RCSB Protein Data Bank.
Kleiger G, Perry J, Eisenberg D	2001	3D structure of the E1beta subunit of pyruvate dehydrogenase from the archeon <i>Pyrobaculum aerophilum</i>	1IK6; http://dx.doi.org/10.2210/pdb1ik6/pdb	Publicly available at the RCSB Protein Data Bank.
Yamada K, Miyata T, Tsuchiya D, Oyama T, Fujiwara Y, Ohnishi T, Iwasaki H, Shinagawa H, Ariyoshi M, Mayanagi K, Morikawa K	2002	RuvA-RuvB complex	1IXR; http://dx.doi.org/10.2210/pdb1ixr/pdb	Publicly available at the RCSB Protein Data Bank.
Vitagliano L, Masullo M, Sica F, Zagari A, Bocchini V	2002	Crystal structure of <i>Sulfolobus solfataricus</i> elongation factor 1 alpha in complex with GDP	1JNY; http://dx.doi.org/10.2210/pdb1jny/pdb	Publicly available at the RCSB Protein Data Bank.
Parsons JF, Jensen PY, Pachikara AS, Howard AJ, Eisenstein E, Ladner JE	2002	THE CRYSTAL STRUCTURE OF AMINODEOXYCHORISMATE SYNTHASE FROM FORMATE GROWN CRYSTALS	1K0E; http://dx.doi.org/10.2210/pdb1k0e/pdb	Publicly available at the RCSB Protein Data Bank.
Korolev S, Koroleva O, Petterson K, Collart F, Dementieva I, Joachimiak A, Midwest Center for Structural Genomics (MCSG)	2002	CRYSTAL STRUCTURE OF ACETATE COA-TRANSFERASE ALPHA SUBUNIT	1K6D; http://dx.doi.org/10.2210/pdb1k6d/pdb	Publicly available at the RCSB Protein Data Bank.
Takahashi H, Tokunaga Y, Kuroishi C, Babayeva N, Kuramitsu S, Yokoyama S, Miyano M, Tahirov TH	2003	THE CRYSTAL STRUCTURE OF SUCCINYL-COA SYNTHETASE ALPHA SUBUNIT FROM THERMUS THERMOPHILUS	1OI7; http://dx.doi.org/10.2210/pdb1oi7/pdb	Publicly available at the RCSB Protein Data Bank.
Weyand M, Schlichting I	2000	CRYSTAL STRUCTURE OF WILD-TYPE TRYPTOPHAN SYNTHASE COMPLEXED WITH INDOLE PROPANOL PHOSPHATE	1QOP; http://dx.doi.org/10.2210/pdb1qop/pdb	Publicly available at the RCSB Protein Data Bank.
Unciuleac M, Warkentin E, Page CC, Dutton PL, Boll M, Ermler U	2004	Structure of 4-hydroxybenzoyl-CoA reductase from <i>Thaueria aromatica</i>	1RM6; http://dx.doi.org/10.2210/pdb1rm6/pdb	Publicly available at the RCSB Protein Data Bank.
Settembre EC, Dorrestein PC, Zhai H, Chatterjee A, McLafferty FW, Begley TP, Ealick SE	2004	Structure of the thiazole synthase/ThiS complex	1TYG; http://dx.doi.org/10.2210/pdb1tyg/pdb	Publicly available at the RCSB Protein Data Bank.
Hioki Y, Ogasahara K, Lee SJ, Ma J, Ishida M, Yamagata Y, Matsuura Y, Ota M, Kuramitsu S, Yutani K, RIKEN Structural Genomics/Proteomics Initiative (RSGI)	2005	X-ray crystal structure of the Tryptophan Synthase b2 Subunit from Hyperthermophile, <i>Pyrococcus furiosus</i>	1V8Z; http://dx.doi.org/10.2210/pdb1v8z/pdb	Publicly available at the RCSB Protein Data Bank.

Frank RAW, Pratap JV, Pei XY, Perham RN, Luisi BF	2004	THE CRYSTAL STRUCTURE OF PYRUVATE DEHYDROGENASE E1 BOUND TO THE PERIPHERAL SUBUNIT BINDING DOMAIN OF E2	1W85; http://dx.doi.org/10.2210/pdb1w85/pdb	Publicly available at the RCSB Protein Data Bank.
Kuzin A, Abashidze M, Vorobiev S, Forouhar F, Acton T, Ma L, Xiao R, Montelione G, Tong L, Hunt J, Northeast Structural Genomics Consortium (NESG)	2004	Crystal structure of Northeast Structural Genomics Target SR156	1XM3; http://dx.doi.org/10.2210/pdb1xm3/pdb	Publicly available at the RCSB Protein Data Bank.
Federici L, Du D, Walas F, Matsumura H, Fernandez-Recio J, McKeegan KS, Borges-Walmsley MI, Luisi BF, Walmsley AR	2005	The crystal structure of the outer membrane protein VceC from the bacterial pathogen <i>Vibrio cholerae</i> at 1.8 resolution	1YC9; http://dx.doi.org/10.2210/pdb1yc9/pdb	Publicly available at the RCSB Protein Data Bank.
Mougous JD, Lee DH, Hubbard SC, Schelle MW, Vocadlo DJ, Berger JM, Bertozzi CR	2006	Crystal Structure of a GTP-Regulated ATP Sulfurylase Heterodimer from <i>Pseudomonas syringae</i>	1ZUN; http://dx.doi.org/10.2210/pdb1zun/pdb	Publicly available at the RCSB Protein Data Bank.
Oberholzer AE, Schneider P, Bachler C, Baumann U, Erni B	2006	CRYSTAL STRUCTURE OF DHAL FROM <i>E. COLI</i>	2BTD; http://dx.doi.org/10.2210/pdb2btd/pdb	Publicly available at the RCSB Protein Data Bank.
Numata T, Fukai S, Ikeuchi Y, Suzuki T, Nureki O	2006	crystal structure of heterohexameric TusBCD proteins, which are crucial for the tRNA modification	2D1P; http://dx.doi.org/10.2210/pdb2d1p/pdb	Publicly available at the RCSB Protein Data Bank.
Asada Y, Kunishima N, RIKEN Structural Genomics/Proteomics Initiative (RSGI)	2007	Structural study of Project ID aq_1548 from <i>Aquifex aeolicus</i> VF5	2EKC; http://dx.doi.org/10.2210/pdb2ekc/pdb	Publicly available at the RCSB Protein Data Bank.
Joint Center for Structural Genomics (JCSG)	2006	Crystal structure of Glutamyl-tRNA(Gln) amidotransferase subunit A (tm1272) from <i>THERMOTOGA MARITIMA</i> at 1.80 Å resolution	2GI3; http://dx.doi.org/10.2210/pdb2gi3/pdb	Publicly available at the RCSB Protein Data Bank.
Lokanath NK, RIKEN Structural Genomics/Proteomics Initiative (RSGI)	2007	Structure of PH0203 protein from <i>Pyrococcus horikoshii</i>	2IT1; http://dx.doi.org/10.2210/pdb2it1/pdb	Publicly available at the RCSB Protein Data Bank.
Fraser ME	2007	C123aT Mutant of <i>E. coli</i> Succinyl-CoA Synthetase Orthorhombic Crystal Form	2NU9; http://dx.doi.org/10.2210/pdb2nu9/pdb	Publicly available at the RCSB Protein Data Bank.
Hollenstein K, Frei DC, Locher KP	2007	ABC transporter ModBC in complex with its binding protein ModA	2ONK; http://dx.doi.org/10.2210/pdb2onk/pdb	Publicly available at the RCSB Protein Data Bank.
Swindell JT II, Chen L, Zhu J, Ebihara A, Shinkai A, Kuramitsu S, Yokoyama S, Fu Z-Q, Chrzas J, Rose JP, Wang B-C, Southeast Collaboratory for Structural Genomics (SECSG), RIKEN Structural Genomics/Proteomics Initiative (RSGI)	2007	Crystal structure of conserved uncharacterized protein PH0987 from <i>Pyrococcus horikoshii</i>	2PHC; http://dx.doi.org/10.2210/pdb2phc/pdb	Publicly available at the RCSB Protein Data Bank.
Jormakka M, Yokoyama K, Yano T, Tamakoshi M, Akimoto S, Shimamura T, Curmi P, Iwata S	2008	POLYSULFIDE REDUCTASE NATIVE STRUCTURE	2VPZ; http://dx.doi.org/10.2210/pdb2vpz/pdb	Publicly available at the RCSB Protein Data Bank.

Ruprecht J, Yankovskaya V, Maklashina E, Iwata S, Cecchini G	2009	E. COLI SUCCINATE: QUINONE OXIDOREDUCTASE (SQR) WITH CARBOXYN BOUND	2WDQ; http://dx.doi.org/10.2210/pdb2wdq/pdb	Publicly available at the RCSB Protein Data Bank.
Kaila VRI, Oksanen E, Goldman A, Verkhovsky MI, Sundholm D, Wikstrom M	2011	Bovine heart cytochrome c oxidase re-refined with molecular oxygen	2Y69; http://dx.doi.org/10.2210/pdb2y69/pdb	Publicly available at the RCSB Protein Data Bank.
Yamada S, Sugimoto H, Kobayashi M, Ohno A, Nakamura H, Shiro Y	2009	Crystal structure of histidine kinase ThkA (TM1359) in complex with response regulator protein TrrA (TM1360)	3A0R; http://dx.doi.org/10.2210/pdb3a0r/pdb	Publicly available at the RCSB Protein Data Bank.
Yamada S, Sugimoto H, Kobayashi M, Ohno A, Nakamura H, Shiro Y	2009	Crystal structure of response regulator protein TrrA (TM1360) from <i>Thermotoga maritima</i> in complex with Mg(2+)-BeF (SeMet, L89M)	3A10; http://dx.doi.org/10.2210/pdb3a10/pdb	Publicly available at the RCSB Protein Data Bank.
Vey JL, Drennan CL	2008	4Fe-4S-Pyruvate formate-lyase Activating Enzyme with partially disordered AdoMet	3C8F; http://dx.doi.org/10.2210/pdb3c8f/pdb	Publicly available at the RCSB Protein Data Bank.
Qin L, Mills DA, Buhrow L, Hiser C, Ferguson-Miller S	2008	Catalytic core subunits (I and II) of cytochrome c oxidase from <i>Rhodobacter sphaeroides</i> complexed with deoxycholic acid	3DTU; http://dx.doi.org/10.2210/pdb3dtu/pdb	Publicly available at the RCSB Protein Data Bank.
Yum S, Xu Y, Piao S, Ha N-C	2009	Crystal structure of E.coli MacA	3FPP; http://dx.doi.org/10.2210/pdb3fpp/pdb	Publicly available at the RCSB Protein Data Bank.
Miallau L, Cascio D, Eisenberg D, TB Structural Genomics Consortium (TBSGC), Integrated Center for Structure and Function Innovation (ISFI)	2009	The crystal structure of the toxin-antitoxin complex RelBE2 (Rv2865-2866) from <i>Mycobacterium tuberculosis</i>	3G5O; http://dx.doi.org/10.2210/pdb3g5o/pdb	Publicly available at the RCSB Protein Data Bank.
Sazanov LA, Berrisford JM	2009	Crystal structure of the hydrophilic domain of respiratory complex I from <i>Thermus thermophilus</i> , oxidized, 4 mol/ASU, re-refined to 3.15 angstrom resolution	3IAS; http://dx.doi.org/10.2210/pdb3ias/pdb	Publicly available at the RCSB Protein Data Bank.
Nakamura A, Yao M, Tanaka I	2009	The high resolution structure of GatCAB	3IP4; http://dx.doi.org/10.2210/pdb3ip4/pdb	Publicly available at the RCSB Protein Data Bank.
Yu S, Rhee S	2010	Crystal structure of Immunogenic lipoprotein A from <i>Vibrio vulnificus</i>	3K2D; http://dx.doi.org/10.2210/pdb3k2d/pdb	Publicly available at the RCSB Protein Data Bank.
Kaufmann M, Chernishof I, Shin A, Germano D, Sawaya MR, Waldo GS, Arbing MA, Perry J, Eisenberg D, Integrated Center for Structure and Function Innovation (ISFI), TB Structural Genomics Consortium (TBSGC)	2010	Allophanate Hydrolase Complex from <i>Mycobacterium smegmatis</i> , Msmeg0435-Msmeg0436	3MML; http://dx.doi.org/10.2210/pdb3mml/pdb	Publicly available at the RCSB Protein Data Bank.
Cingolani G, Duncan TM	2011	Structure of the E.coli F1-ATP synthase inhibited by subunit Epsilon	3OAA; http://dx.doi.org/10.2210/pdb3oaa/pdb	Publicly available at the RCSB Protein Data Bank.

Shi R, McDonald L, Matte A, Cygler M, Ekiel I, Montreal-Kingston Bacterial Structural Genomics Initiative (BSGI)	2011	Crystal Structure of E.coli Dha kinase DhaK	3PNK; http://dx.doi.org/10.2210/pdb3pnk/pdb	Publicly available at the RCSB Protein Data Bank.
Shi R, McDonald L, Matte A, Cygler M, Ekiel I, Montreal-Kingston Bacterial Structural Genomics Initiative (BSGI)	2011	Crystal Structure of E.coli Dha kinase DhaK-DhaL complex	3PNL; http://dx.doi.org/10.2210/pdb3pnl/pdb	Publicly available at the RCSB Protein Data Bank.
Nocek B, Stein A, Marshall N, Jedrzejczak R, Babnigg G, Joachimiak A, Midwest Center for Structural Genomics (MCSG)	2011	Protein-protein complex of subunit 1 and 2 of Molybdopterin-converting factor from <i>Helicobacter pylori</i> 26695	3RPF; http://dx.doi.org/10.2210/pdb3rpf/pdb	Publicly available at the RCSB Protein Data Bank.
Nocek B, Stein A, Marshall N, Jedrzejczak R, Babnigg G, Joachimiak A, Midwest Center for Structural Genomics (MCSG)	2011	Complex structure of 3-oxoadipate coA-transferase subunit A and B from <i>Helicobacter pylori</i> 26695	3RRL; http://dx.doi.org/10.2210/pdb3rrl/pdb	Publicly available at the RCSB Protein Data Bank.
Johnson E, Nguyen PT, Rees DC	2011	Inward facing conformations of the MetNI methionine ABC transporter: CY5 native crystal form	3TUI; http://dx.doi.org/10.2210/pdb3tui/pdb	Publicly available at the RCSB Protein Data Bank.
Safro M, Klipcan L, Moor N, Finarov I, Kessler N, Sukhanova M	2011	Crystal structure of human mitochondrial PheRS complexed with tRNAPhe in the active open state	3TUP; http://dx.doi.org/10.2210/pdb3tup/pdb	Publicly available at the RCSB Protein Data Bank.
Bulkley D, Johnson FA, Steitz TA	2012	The structure of thermorubin in complex with the 70S ribosome from <i>Thermus thermophilus</i> . This file contains the 50S subunit of one 70S ribosome. The entire crystal structure contains two 70S ribosomes	3UXR; http://dx.doi.org/10.2210/pdb3uxr/pdb	Publicly available at the RCSB Protein Data Bank.
Mancusso RL, Gregorio GG, Liu Q, Wang DN	2012	Crystal Structure of a bacterial dicarboxylate/sodium symporter	4F35; http://dx.doi.org/10.2210/pdb4f35/pdb	Publicly available at the RCSB Protein Data Bank.
Baradaran R, Berrisford JM, Minhas GS, Sazanov LA	2013	Crystal structure of the entire respiratory complex I from <i>Thermus thermophilus</i>	4HEA; http://dx.doi.org/10.2210/pdb4hea/pdb	Publicly available at the RCSB Protein Data Bank.
Broussard TC, Kobe MJ, Pakhomova S, Neau DB, Price AE, Champion TS, Waldrop GL	2013	Crystal Structure of Biotin Carboxyl Carrier Protein-Biotin Carboxylase Complex from E.coli	4HR7; http://dx.doi.org/10.2210/pdb4hr7/pdb	Publicly available at the RCSB Protein Data Bank.

References

- Balakrishnan S**, Kamisetty H, Carbonell JG, Lee Su-I, Langmead CJ. 2011. Learning generative models for protein fold families. *Proteins: structure, Function, and Bioinformatics* **79**:1061–1078. doi: [10.1002/prot.22934](https://doi.org/10.1002/prot.22934).
- Baradaran R**, Berrisford JM, Minhas GS, Sazanov LA. 2013. Crystal structure of the entire respiratory complex I. *Nature* **494**:443–448. doi: [10.1038/nature11871](https://doi.org/10.1038/nature11871).
- Barth P**, Schonbrun J, Baker D. 2007. Toward high-resolution prediction and design of transmembrane helical protein structures. *Proceedings of the National Academy of Sciences of the United States of America* **104**:15682–15687. doi: [10.1073/pnas.0702515104](https://doi.org/10.1073/pnas.0702515104).
- Becker A**, Kabsch W. 2002. X-ray structure of pyruvate formate-lyase in complex with pyruvate and CoA. How the enzyme uses the Cys-418 thiol radical for pyruvate cleavage. *Journal of Biological Chemistry* **277**:40036–40042. doi: [10.1074/jbc.M205821200](https://doi.org/10.1074/jbc.M205821200).
- Betts MJ**, Sternberg MJE. 1999. An analysis of conformational changes on protein–protein association: implications for predictive docking. *Protein Engineering* **12**:271–283. doi: [10.1093/protein/12.4.271](https://doi.org/10.1093/protein/12.4.271).

- Bulkley D**, Johnson F, Steitz TA. 2012. The antibiotic thermorubin inhibits protein synthesis by binding to inter-subunit bridge B2a of the ribosome. *Journal of Molecular Biology* **416**:571–578. doi: [10.1016/j.jmb.2011.12.055](https://doi.org/10.1016/j.jmb.2011.12.055).
- Burger L**, van Nimwegen E. 2008. Accurate prediction of protein–protein interactions from sequence alignments using a Bayesian method. *Molecular Systems Biology* **4**:165. doi: [10.1038/msb4100203](https://doi.org/10.1038/msb4100203).
- Cong Q**, Grishin NV. 2012. MESSA: MEta-server for protein sequence analysis. *BMC Biology* **10**:82. doi: [10.1186/1741-7007-10-82](https://doi.org/10.1186/1741-7007-10-82).
- Conway P**, Tyka MD, DiMaio F, Konerding DE, Baker D. 2014. Relaxation of backbone bond geometry improves protein energy landscape modeling. *Protein Science* **23**:47–55. doi: [10.1002/pro.2389](https://doi.org/10.1002/pro.2389).
- Dago AE**, Schug A, Procaccini A, Hoch JA, Weigt M, Szurmant H. 2012. Structural basis of histidine kinase autophosphorylation deduced by integrating genomics, molecular dynamics, and mutagenesis. *Proceedings of the National Academy of Sciences of the United States of America* **109**:E1733–E1742. doi: [10.1073/pnas.1201301109](https://doi.org/10.1073/pnas.1201301109).
- Datta RS**, Meacham C, Samad B, Neyer C, Sjölander K. 2009. Berkeley PHOG: PhyloFacts orthology group prediction web server. *Nucleic Acids Research* **37**:W84–W89. doi: [10.1093/nar/gkp373](https://doi.org/10.1093/nar/gkp373).
- de Juan D**, Pazos F, Valencia A. 2013. Emerging methods in protein co-evolution. *Nature Reviews Genetics* **14**:249–261. doi: [10.1038/nrg3414](https://doi.org/10.1038/nrg3414).
- Duhovny D**, Nussinov R, Wolfson HJ. 2002. Efficient unbound docking of rigid molecules. In: Berlin, editor. *Algorithms in bioinformatics*. Springer: Heidelberg. p. 185–200. doi: [10.1007/3-540-45784-4_14](https://doi.org/10.1007/3-540-45784-4_14).
- Dunn SD**, Wahl LM, Gloor GB. 2008. Mutual information without the influence of phylogeny or entropy dramatically improves residue contact prediction. *Bioinformatics* **24**:333–340. doi: [10.1093/bioinformatics/btm604](https://doi.org/10.1093/bioinformatics/btm604).
- Eddy SR**. 2009. A new generation of homology search tools based on probabilistic inference. *Genome Informatics* **23**:205–211.
- Ekeberg M**, Lökvist C, Lan Y, Weigt M, Aurell E. 2013. Improved contact prediction in proteins: using pseudo-likelihoods to infer Potts models. *Physical Review E* **87**:012707. doi: [10.1103/PhysRevE.87.012707](https://doi.org/10.1103/PhysRevE.87.012707).
- Federici L**, Du D, Walas F, Matsumura H, Fernandez-Recio J, McKeegan KS, Borges-Walmsley MI, Luisi BF, Walmsley AR. 2005. The crystal structure of the outer membrane protein VceC from the bacterial pathogen *Vibrio cholerae* at 1.8 Å resolution. *Journal of Biological Chemistry* **280**:15307–15314. doi: [10.1074/jbc.M500401200](https://doi.org/10.1074/jbc.M500401200).
- Gray JJ**, Moughon S, Wang C, Schueler-Furman O, Kuhlman B, Rohl CA, Baker D. 2003. Protein–protein docking with simultaneous optimization of rigid-body displacement and side-chain conformations. *Journal of Molecular Biology* **331**:281–299. doi: [10.1016/S0022-2836\(03\)00670-3](https://doi.org/10.1016/S0022-2836(03)00670-3).
- Halperin I**, Wolfson H, Nussinov R. 2006. Correlated mutations: Advances and limitations. A study on fusion proteins and on the Cohesin-Dockerin families. *Proteins: structure, Function, and Bioinformatics* **63**:832–845. doi: [10.1002/prot.20933](https://doi.org/10.1002/prot.20933).
- Hopf TA**, Colwell LJ, Sheridan R, Rost B, Sander C, Marks DS. 2012. Three-dimensional structures of membrane proteins from genomic sequencing. *Cell* **149**:1607–1621. doi: [10.1016/j.cell.2012.04.012](https://doi.org/10.1016/j.cell.2012.04.012).
- Hopf TA**, Schärfe CPI, Rodrigues JPGLM, Green AG, Sander C, Bonvin AMJJ, Marks DS. 2014. Sequence co-evolution gives 3D contacts and structures of protein complexes. *bioRxiv*. doi: [10.1101/004762](https://doi.org/10.1101/004762).
- Hosur R**, Peng J, Vinayagam A, Stelzl U, Xu J, Perrimon N, Bienkowska J, Berger B. 2012. A computational framework for boosting confidence in high-throughput protein-protein interaction datasets. *Genome Biology* **13**:R76. doi: [10.1186/gb-2012-13-8-r76](https://doi.org/10.1186/gb-2012-13-8-r76).
- Jacob F**, Perrin D, Sánchez C, Monod J. 2005. L'opéron: groupe de gènes à expression coordonnée par un opérateur [CR Acad. Sci. Paris 250 (1960) 1727–1729]. *Comptes Rendus Biologies* **328**:514–520. doi: [10.1016/j.crv.2005.04.005](https://doi.org/10.1016/j.crv.2005.04.005).
- Johnson E**, Nguyen PT, Yeates TO, Rees DC. 2012. Inward facing conformations of the MetNI methionine ABC transporter: Implications for the mechanism of transinhibition. *Protein Science* **21**:84–96. doi: [10.1002/pro.765](https://doi.org/10.1002/pro.765).
- Jones DT**, Buchan DWA, Cozzetto D, Pontil M. 2012. PSICOV: precise structural contact prediction using sparse inverse covariance estimation on large multiple sequence alignments. *Bioinformatics* **28**:184–190. doi: [10.1093/bioinformatics/btr638](https://doi.org/10.1093/bioinformatics/btr638).
- Kamisetty H**, Ovchinnikov S, Baker D. 2013. Assessing the utility of coevolution-based residue–residue contact predictions in a sequence-and structure-rich era. *Proceedings of the National Academy of Sciences of the United States of America* **110**:15674–15679. doi: [10.1073/pnas.1319550110](https://doi.org/10.1073/pnas.1319550110).
- Lange OF**, Rossi P, Sgourakis NG, Song Y, Hsiau-Wei L, Aramini JM, Ertekin A, Xiao R, Acton TB, Montelione GT, Baker D. 2012. Determination of solution structures of proteins up to 40 kDa using CS-Rosetta with sparse NMR data from deuterated samples. *Proceedings of the National Academy of Sciences of the United States of America* **109**:10873–10878. doi: [10.1073/pnas.1203013109](https://doi.org/10.1073/pnas.1203013109).
- Lapedes A**, Giraud B, Jarzynski C. 2012. Using sequence alignments to predict protein structure and stability with high accuracy. arXiv 1207.2484. <http://arxiv.org/abs/1207.2484>
- Lapedes AS**, Giraud BG, Liu LC, Stormo GD. 1999. Correlated mutations in models of protein sequences: phylogenetic and structural effects. *Lecture Notes-Monograph Series* **33**:236–256. doi: [10.1214/lnms/1215455556](https://doi.org/10.1214/lnms/1215455556).
- Long F**, Su CC, Lei HT, Bolla JR, Do SV, Yu EW. 2012. Structure and mechanism of the tripartite CusCBA heavy-metal efflux complex. *Philosophical Transactions of the Royal Society B: Biological Sciences* **367**:1047–1058. doi: [10.1098/rstb.2011.0203](https://doi.org/10.1098/rstb.2011.0203).
- Mancusso R**, Gregorio GG, Liu Q, Wang Da-N. 2012. Structure and mechanism of a bacterial sodium-dependent dicarboxylate transporter. *Nature* **491**:622–626. doi: [10.1038/nature11542](https://doi.org/10.1038/nature11542).
- Marks DS**, Colwell LJ, Sheridan R, Hopf TA, Pagnani A, Zecchina R, Sander C. 2011. Protein 3D structure computed from evolutionary sequence variation. *PLOS ONE* **6**:e28766. doi: [10.1371/journal.pone.0028766](https://doi.org/10.1371/journal.pone.0028766).

- Marks DS**, Hopf TA, Sander C. 2012. Protein structure prediction from sequence variation. *Nature Biotechnology* **30**:1072–1080. doi: [10.1038/nbt.2419](https://doi.org/10.1038/nbt.2419).
- Morcos F**, Jana B, Hwa T, Onuchic JN. 2013. Coevolutionary signals across protein lineages help capture multiple protein conformations. *Proceedings of the National Academy of Sciences of the United States of America* **110**:20533–20538. doi: [10.1073/pnas.1315625110](https://doi.org/10.1073/pnas.1315625110).
- Morcos F**, Pagnani A, Lunt B, Bertolino A, Marks DS, Sander C, Zecchina R, Onuchic JN, Hwa T, Weigt M. 2011. Direct-coupling analysis of residue coevolution captures native contacts across many protein families. *Proceedings of the National Academy of Sciences of the United States of America* **108**:E1293–E1301. doi: [10.1073/pnas.1111471108](https://doi.org/10.1073/pnas.1111471108).
- Mulligan C**, Fischer M, Thomas GH. 2011. Tripartite ATP-independent periplasmic (TRAP) transporters in bacteria and archaea. *FEMS Microbiology Reviews* **35**:68–86. doi: [10.1111/j.1574-6976.2010.00236.x](https://doi.org/10.1111/j.1574-6976.2010.00236.x).
- Nakamura A**, Sheppard K, Yamane J, Yao M, Söll D, Tanaka I. 2010. Two distinct regions in *Staphylococcus aureus* GatCAB guarantee accurate tRNA recognition. *Nucleic Acids Research* **38**:672–682. doi: [10.1093/nar/gkp955](https://doi.org/10.1093/nar/gkp955).
- Nugent T**, Jones DT. 2012. Accurate de novo structure prediction of large transmembrane protein domains using fragment-assembly and correlated mutation analysis. *Proceedings of the National Academy of Sciences of the United States of America* **109**:E1540–E1547. doi: [10.1073/pnas.1120036109](https://doi.org/10.1073/pnas.1120036109).
- Ochoa D**, Pazos F. 2010. Studying the co-evolution of protein families with the Mirrortree web server. *Bioinformatics* **26**:1370–1371. doi: [10.1093/bioinformatics/btq137](https://doi.org/10.1093/bioinformatics/btq137).
- Ovchinnikov S**, Kamisetty H, Baker D. 2014. Data from: Robust and accurate prediction of residue-residue interactions across protein interfaces using evolutionary information. *Dryad Digital Repository*. doi: [10.5061/dryad.s00vr](https://doi.org/10.5061/dryad.s00vr).
- Pazos F**, Helmer-Citterich M, Ausiello G, Valencia A. 1997. Correlated mutations contain information about protein-protein interaction. *Journal of Molecular Biology* **271**:511–523. doi: [10.1006/jmbi.1997.1198](https://doi.org/10.1006/jmbi.1997.1198).
- Raman S**, Vernon R, Thompson J, Tyka M, Sadreyev R, Pei J, Kim D, Kellogg E, DiMaio F, Lange O, Kinch L, Sheffler W, Kim BH, Das R, Grishin NV, Baker D. 2009. Structure prediction for CASP8 with all-atom refinement using Rosetta. *Proteins: structure, Function, and Bioinformatics* **77**:89–99. doi: [10.1002/prot.22540](https://doi.org/10.1002/prot.22540).
- Remm M**, Storm CEV, Sonnhammer ELL. 2001. Automatic clustering of orthologs and in-paralogs from pairwise species comparisons. *Journal of Molecular Biology* **314**:1041–1052. doi: [10.1006/jmbi.2000.5197](https://doi.org/10.1006/jmbi.2000.5197).
- Remmert M**, Biegert A, Hauser A, Söding J. 2011. HHblits: lightning-fast iterative protein sequence searching by HMM-HMM alignment. *Nature Methods* **9**:173–175. doi: [10.1038/nmeth.1818](https://doi.org/10.1038/nmeth.1818).
- Schug A**, Weigt M, Onuchic JN, Hwa T, Szurmant H. 2009. High-resolution protein complexes from integrating genomic information with molecular simulation. *Proceedings of the National Academy of Sciences of the United States of America* **106**:22124–22129. doi: [10.1073/pnas.0912100106](https://doi.org/10.1073/pnas.0912100106).
- Shoemaker BA**, Panchenko AR. 2007. Deciphering protein–protein interactions. Part II. Computational methods to predict protein and domain interaction partners. *PLOS Computational Biology* **3**:e43. doi: [10.1371/journal.pcbi.0030043](https://doi.org/10.1371/journal.pcbi.0030043).
- Sievers F**, Wilm A, Dineen D, Gibson TJ, Karplus K, Li W, Lopez R, McWilliam H, Remmert M, Söding J, Thompson JD, Higgins DG. 2011. Fast, scalable generation of high-quality protein multiple sequence alignments using Clustal Omega. *Molecular Systems Biology* **7**:539. doi: [10.1038/msb.2011.75](https://doi.org/10.1038/msb.2011.75).
- Simons KT**, Ruczinski I, Kooperberg C, Fox BA, Bystroff C, Baker D. 1999. Improved recognition of native-like protein structures using a combination of sequence-dependent and sequence-independent features of proteins. *Proteins: structure, Function, and Bioinformatics* **34**:82–95. doi: [10.1002/\(SICI\)1097-0134\(19990101\)34:1<82::AID-PROT7>3.0.CO;2-A](https://doi.org/10.1002/(SICI)1097-0134(19990101)34:1<82::AID-PROT7>3.0.CO;2-A).
- Song Y**, DiMaio F, Wang RYR, Kim D, Miles C, Brunette TJ, Thompson J, Baker D. 2013. High-Resolution comparative modeling with RosettaCM. *Structure* **21**:1735–1742. doi: [10.1016/j.str.2013.08.005](https://doi.org/10.1016/j.str.2013.08.005).
- Suhre K**, Claverie JM. 2004. FusionDB: a database for in-depth analysis of prokaryotic gene fusion events. *Nucleic Acids Research* **32**:D273–D276. doi: [10.1093/nar/gkh053](https://doi.org/10.1093/nar/gkh053).
- Sułkowska JI**, Morcos F, Weigt M, Hwa T, Onuchic JN. 2012. Genomics-aided structure prediction. *Proceedings of the National Academy of Sciences of the United States of America* **109**:10340–10345. doi: [10.1073/pnas.1207864109](https://doi.org/10.1073/pnas.1207864109).
- Tamir S**, Rotem-Bamberger S, Katz C, Morcos F, Hailey KL, Zuris JA, Wang C, Conlan AR, Lipper CH, Paddock ML, Mittler R, Onuchic JN, Jennings PA, Friedler A, Nechushtai R. 2014. Integrated strategy reveals the protein interface between cancer targets Bcl-2 and NAF-1. *Proceedings of the National Academy of Sciences of the United States of America* **111**:5177–5182. doi: [10.1073/pnas.1403770111](https://doi.org/10.1073/pnas.1403770111).
- Thomas J**, Ramakrishnan N, Bailey-Kellogg C. 2008. Graphical models of residue coupling in protein families. *IEEE/ACM Transactions on Computational Biology and Bioinformatics (TCBB)* **5**:183–197. doi: [10.1109/TCBB.2007.70225](https://doi.org/10.1109/TCBB.2007.70225).
- UniProt Accession**. UniProt User manual. http://www.uniprot.org/manual/accession_numbers. accessed September 9, 2013.
- Valencia A**, Pazos F. 2002. Computational methods for the prediction of protein interactions. *Current Opinion in Structural Biology* **12**:368–373. doi: [10.1016/S0959-440X\(02\)00333-0](https://doi.org/10.1016/S0959-440X(02)00333-0).
- Vey JL**, Yang J, Li M, Broderick WE, Broderick JB, Drennan CL. 2008. Structural basis for glycol radical formation by pyruvate formate-lyase activating enzyme. *Proceedings of the National Academy of Sciences of the United States of America* **105**:16137–16141. doi: [10.1073/pnas.0806640105](https://doi.org/10.1073/pnas.0806640105).
- Wang G**, Dunbrack RL Jr. S2C: a database correlating sequence and atomic coordinate residue numbering in the Protein Data Bank. Dunbrack Lab. <http://dunbrack.fccc.edu/s2c/>. Accessed October 12, 2013.

- Weigt M**, White RA, Szurmant H, Hoch JA, Hwa T. 2009. Identification of direct residue contacts in protein–protein interaction by message passing. *Proceedings of the National Academy of Sciences of the United States of America* **106**:67–72. doi: [10.1073/pnas.0805923106](https://doi.org/10.1073/pnas.0805923106).
- Wu CH**, Apweiler R, Bairoch A, Natale DA, Barker WC, Boeckmann B, Ferro S, Gasteiger E, Huang H, Lopez R, Magrane M, Martin MJ, Mazumder R, O'Donovan C, Redaschi N, Suzek B. 2006. The Universal Protein Resource (UniProt): an expanding universe of protein information. *Nucleic Acids Research* **34**:D187–D191. doi: [10.1093/nar/gkj161](https://doi.org/10.1093/nar/gkj161).
- Yarov-Yarovoy V**, DeCaen PG, Westenbroek RE, Pan CY, Scheuer T, Baker D, Catterall WA. 2012. Structural basis for gating charge movement in the voltage sensor of a sodium channel. *Proceedings of the National Academy of Sciences of the United States of America* **109**:E93–E102. doi: [10.1073/pnas.1118434109](https://doi.org/10.1073/pnas.1118434109).
- Yu S**, Yeon Lee N, Park SJ, Rhee S. 2011. Crystal structure of toll-like receptor 2-activating lipoprotein IIpA from *Vibrio vulnificus*. *Proteins: structure, Function, and Bioinformatics* **79**:1020–1025. doi: [10.1002/prot.22929](https://doi.org/10.1002/prot.22929).
- Yum S**, Xu Y, Piao S, Sim SH, Kim HM, Jo WS, Kim KJ, Kweon HS, Jeong MH, Jeon H, Lee K, Ha NC. 2009. Crystal structure of the periplasmic component of a tripartite macrolide-specific efflux pump. *Journal of Molecular Biology* **387**:1286–1297. doi: [10.1016/j.jmb.2009.02.048](https://doi.org/10.1016/j.jmb.2009.02.048).
- Zhang QC**, Petrey D, Deng L, Qiang L, Shi Y, Thu CA, Bisikirska B, Lefebvre C, Accili D, Hunter T, Maniatis T, Califano A, Honig B. 2012. Structure-based prediction of protein-protein interactions on a genome-wide scale. *Nature* **490**:556–560. doi: [10.1038/nature11503](https://doi.org/10.1038/nature11503).
- Zhou J**, Rudd KE. 2013. EcoGene 3.0. *Nucleic Acids Research* **41**:D613–D624. doi: [10.1093/nar/gks1235](https://doi.org/10.1093/nar/gks1235).