Conformational selection and induced fit mechanism underlie specificity in noncovalent interactions with ubiquitin

Tomasz Włodarski^{a,b} and Bojan Zagrovic^{b,c,1}

^aLaboratory of Bioinformatics and Bioengineering, Interdisciplinary Centre for Mathematical and Computational Modeling, University of Warsaw, Zwirki i Wigury 93, 02-089, Warsaw, Poland; ^bLaboratory of Computational Biophysics, Mediterranean Institute for Life Sciences, Mestrovicevo setaliste b.b., 21000 Split, Croatia; and ^cDepartment of Physics, University of Split, 2100 Split, Croatia

Edited by Michael Levitt, Stanford University School of Medicine, Stanford, CA, and approved October 5, 2009 (received for review June 28, 2009)

Noncovalent binding interactions between proteins are the central physicochemical phenomenon underlying biological signaling and functional control on the molecular level. Here, we perform an extensive structural analysis of a large set of bound and unbound ubiquitin conformers and study the level of residual induced fit after conformational selection in the binding process. We show that the region surrounding the binding site in ubiquitin undergoes conformational changes that are significantly more pronounced compared with the whole molecule on average. We demonstrate that these induced-fit structural adjustments are comparable in magnitude to conformational selection. Our final model of ubiquitin binding blends conformational selection with the subsequent induced fit and provides a quantitative measure of their respective contributions.

ubiquitin binding | protein recognition | Kolmogorov–Smirnov test

he picture of protein–protein interactions has, over the decades, evolved from the early lock-and-key hypothesis (1) to the generally accepted and widely applied induced-fit model (2, 3). However, several different systems have recently been shown to follow an alternative paradigm whose central element is the idea of conformational selection (4). Within this paradigm, the conformational change in binding is thought to originate primarily from the conformational diversity of the unbound state (5–15). Simply put, the unbound protein explores the energy landscape, spending most of the time in the lowest energy conformations, but also occupying higher-energy ones, some of which are structurally similar to the bound conformations. In the course of binding, because of favorable interactions with the ligand, these conformers get preferentially selected and the population of protein microstates shifts in the direction of bound conformations (4-15). In a way, induced fit and conformational selection are two extremes of possible mechanisms underlying protein interactions (16): in the former, optimal binding is achieved by specific structural change, whereas in the latter it is brought about through selection from the already present unbound ensemble. The two mechanisms have recently been compared from the perspective of kinetics (17) and the energy landscape theory (18).

Some of the earliest-described examples of the conformational selection paradigm are the antibody-antigen interactions where an antibody can be found in different unbound conformations, exhibiting different specificity for different antigens (19–22). Binding then occurs by a simple selection of those antigens whose epitopes are already in a matching conformation for the paratope. In general, growing support for conformational selection in specific protein–protein interactions is based mainly on finding bound-like conformations of proteins in the respective unbound ensembles of structures (12, 14, 23–30). For example, Gsponer et al. (30) proposed that Ca²⁺-bound, ligand-free calmodulin samples the conformational space of calmodulin bound to myosin light chain kinase. Apart from such cases with bound-like conformations in the unbound state on the level of the whole molecule, there are several

examples where specifically those residues that participate in binding are found in a proper conformation already before binding (31–34). Despite such successes, there are many examples of systems that cannot be explained by conformational selection. For example, Sullivan and Holyoak (35) showed that in phosphoenol-pyruvate carboxykinase (PEPCK) formation of the catalytic active complex is combined with a closure of the active site. This means that, even if in the unbound state PEPCK samples bound conformations, they would simply not be available for the substrate. To overcome pitfalls in each of the above models, Grünberg et al. (11) proposed a three-step model where diffusional encounter is followed by the recognition of complementary structures within the conformational ensemble of unbound proteins and subsequent refolding, i.e., induced fit.

Advances in NMR have resulted in a high-resolution, dynamic picture of protein ensembles, allowing us to study protein-protein interactions with unprecedented resolution (36-41). In a recent tour de force study using residual dipolar couplings (RDCs), Lange and coworkers (39, 41) have demonstrated that free ubiquitin samples conformations globally similar to those in the bound state. Ubiquitin is a highly conserved, 76-residue protein that has been well studied from both structural and functional standpoints, because of its extreme importance in different key biological processes such as protein degradation, cell-cycle regulation, or transcription control (42-45). In their study, Lange and coworkers (39, 41) compared an ensemble of X-ray structures of ubiquitin, bound to different ubiquitin-binding proteins, with NMR structures of ubiquitin free in solution. The latter ensemble captured the full dynamic behavior of ubiquitin on the picosecond to microsecond time scale, extending and completing the previous picture of its dynamics (46). Surprisingly, Lange and coworkers demonstrated that for each bound ubiquitin structure there is a member of the unbound ensemble that is structurally similar to it in the rmsd sense, thus giving strong support to the conformational selection mechanism (16). Furthermore, using principal component analysis, they showed that the dominant motion in ubiquitin binding entails a pincer-like motion around the central hydrophobic patch in ubiquitin.

Although the effort by Lange and coworkers (39, 41) provided good evidence for conformational selection in ubiquitin binding, many open questions remain. Here, we analyze the residual induced fit after the conformational selection step in ubiquitin binding and assess the possibility that induced fit and conformational selection actually coexist in the course of ubiquitin binding. Following Lange

 $\label{lem:contributions: T.W. and B.Z. designed research, performed research, analyzed data, and wrote the paper.$

The authors declare no conflict of interest.

This article is a PNAS Direct Submission.

¹To whom correspondence should be addressed. E-mail: zagrovic@medils.hr.

This article contains supporting information online at www.pnas.org/cgi/content/full/ 0906966106/DCSupplemental.

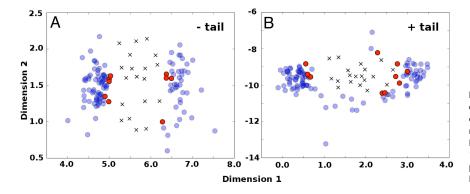


Fig. 1. MDS analysis of ubiquitin binding. Global conformational analysis by MDS, using backbone atom-positional rmsd as a metric, performed on a merged ensemble consisting of X-ray (crosses) and NMR (blue and red circles) structures. (A) Residues 1–70. (B) All residues. NMR structures, which are depicted in red, represent the most similar (in rmsd sense) NMR structures to at least one X-ray structure.

and coworkers we performed an exhaustive structural comparison between a set of ubiquitin X-ray structures bound to different partners with an RDC-refined ensemble of ubiquitin free in solution. We focus on the NMR ensemble by Lange and coworkers, but to confirm our findings, we also analyzed two other structural ensembles (Backrub ensembles from ref. 47) that were refined based on the same RDC data.

Results

Ubiquitin, when free in solution, samples a subset of structures that, in terms of their global conformation, are similar to ubiquitin in the bound state in the rmsd sense (39, 41). Global multidimensional scaling (MDS) analysis using Lange and coworkers' EROS ensemble (39, 41) supports this finding, but also adds information (Fig. 1). Calculated MDS maps inform about relative distances (here, backbone atom-positional rmsd values between conformations) of multidimensional data in a reduced dimensional space. We present our results for: (i) residues 1–70 of ubiquitin (excluding the flexible C-terminal tail) and (ii) residues 1–76 (including the tail). Although the inclusion of the C-terminal tail makes structural comparisons more difficult because of its intrinsic flexibility, our analysis indicates that this region is directly involved in binding in \approx 50% of all ubiquitin complexes studied, and its contribution cannot be ignored (Fig. S1). 2D MDS maps (Fig. 1) demonstrate that, in general, there

appears to be a small number of NMR structures exhibiting globally very similar conformations to bound X-ray structures, whereas the rest are significantly more different. For example, >60% of the whole X-ray ensemble is captured by only three dominant NMR unbound structures (see also Fig. S2), which is also evident from the 2D MDS maps (Fig. 1), despite the inaccuracies associated with projecting multidimensional data into two dimensions (Fig. S3). Additionally, if one focuses on all unbound structures with an rmsd in the range of 0.1 Å from the lowest rmsd value, a similar picture is obtained (Fig. S2). Altogether, this analysis supports the thesis that the unbound protein during its dynamics does not frequently adopt bound conformations. Moreover, the ensemble of unbound NMR structures is clearly separated from the bound X-ray ensemble and appears to be significantly more diffuse in a structural sense. Finally, although quantitatively different, the MDS maps with the C-terminal tail excluded or included are qualitatively fairly similar.

More information about the high-resolution details can be obtained by looking at local conformational differences, especially close to the binding site. In Fig. 2, we see what are the average deviations of atoms between the globally most similar unbound ubiquitin structure and the corresponding bound structure, as a function of distance from the binding site. In other words, our analysis probes the residual induced fit after conformational selection. In each case (Fig. 2), the average local atom–atom rmsd (Fig.

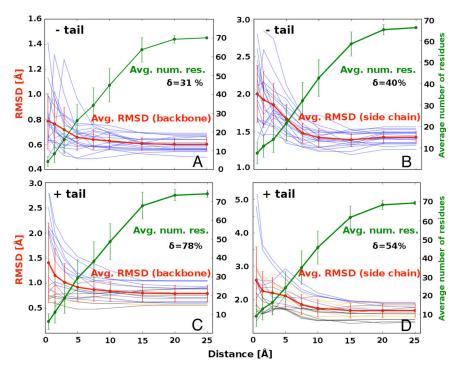


Fig. 2. Induced fit in ubiquitin binding as a function of distance from the binding site. Local structural differences between the conformationally selected unbound structures and the corresponding bound structures of ubiquitin, captured as the average atomic rmsd values, and given as a function of distance from the binding site without the C-terminal tail [backbone (A) and side chains (B)] or with the C-terminal tail [backbone (C) and side chains (D) included]. The red curve represents the mean rmsd values calculated from all 19 pairs of structures, the blue curves represent the individual structural pairs, and the green curve represents the average number of residues in each distance range. The δ parameter, defined as $\delta = (rmsd_{0.5\text{Å}})$ $rmsd_{25\mathring{a}})/rmsd_{25\mathring{a}}) \times 100\%$, captures the extent to which local conformational deviations close to the binding site (0.5 Å range) are greater than the global deviations (25-Å range). We use colors to illustrate the extent to which the residues in the C-terminal tail take part in forming the binding site: blue, more than 1/4 of all of the tail residues are in the binding site (from 25% to 100%); yellow, <25% of tail residues are in the binding site; black, tail does not take part in the binding interaction. All error bars denote standard deviation of a given variable.

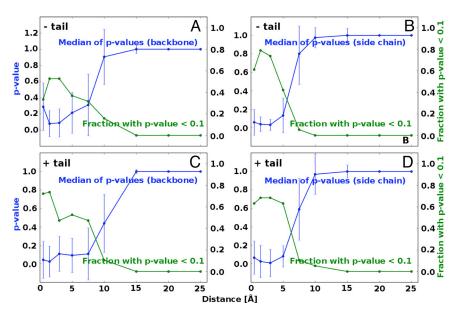


Fig. 3. Statistical significance of induced fit motions in ubiquitin binding. The KS *P* value analysis of statistical significance of local conformational deviations from those of the molecule as a whole as a function of distance from the binding site. The blue curve depicts the median of *P* values from 19 pairs of structures, and the green curve captures the fraction of structures with *P* < 0.1 in each distance range, without the C-terminal tail [backbone (*A*) and side chains (*B*)] or with the C-terminal tail [backbone (*C*) and side chains (*D*)] included.

2, red curve) is highest in the vicinity of the binding site. For example, the backbone atoms of the residues immediately surrounding the binding site deviate 31% more compared with the structure on average (Fig. 2A; 0.5-Å range, residues 1–70, δ = 31%). Even more importantly, there are structures of bound ubiquitin for which this deviation is significantly greater than the average, reaching >100% in some cases. This difference is even more pronounced if one focuses only on side-chain atoms. First, the global structural deviation between the conformationally selected structure and the corresponding bound structure is significantly greater than just for the backbone (1.4 Å versus 0.65 Å on average). Second, the local deviations around the binding site are there even more pronounced with $\delta = 40\%$. In general, the effect of the binding site proximity extends in all cases typically >5–10 Å away from it. In Fig. 2 C and D we illustrate the influence of the flexible region (residues 71–76) in ubiquitin binding. The presence of these tail residues in the binding site (Fig. 2 C and D, blue curves) increases the discrepancy of rmsd values between the binding site and the whole protein. In other words, the tail residues that participate in binding exhibit significant conformational change after conformational selection. Although the δ value for just the backbone is greater than for the side chains (78% vs. 54%), for individual cases the average local structural deviation for side chains is often higher than for the backbone (e.g., 5 Å versus 2–3 Å; Fig. 2 *C* and *D*).

The structural deviations between a given conformationally selected structure and the corresponding bound structure appear to be more pronounced around the binding site compared with the molecule as a whole (Fig. 2). However, are these differences statistically significant? To quantitatively address this question, we have used the two-sample Kolmogorov–Smirnov (KS) P value analysis and asked whether the magnitudes of the local atomic deviations are drawn from the same distribution as the magnitudes of the global atomic deviations. In other words, we probe the statistical significance and the associated P values of the null-hypothesis that the distributions of the magnitudes of local atomic deviations between a given bound structure and the corresponding conformationally selected conformer are drawn from the same distribution as the deviations on the level of the whole molecule. We calculate this as a function of distance from the binding site (see

Materials and Methods). With median values and their average deviation, we present the calculated P values in each distance range for the 19 pairs of structures (Fig. 3). Additionally, we present a fraction of structures with P < 0.1 in each distance range.

For backbone atoms, conformational differences between the local and the global, all-structure levels are statistically significant up to \approx 5 Å away from the binding site, with the median P values typically <0.1. For example, if one focuses only on the backbone, >50% of structures with the C-terminal tail included exhibit P <0.1 up to 5 Å and \approx 40% without the tail included (Fig. 3A and C). A similar situation is seen if one looks at side-chain atoms, although it appears that their fluctuations dissipate over somewhat shorter distances than those of the backbone atoms (Fig. 3 B and D). Note that the incorporation of "tail" residues typically does not change the level of statistical significance, possibly because it influences local and global rmsd values equally. A merged set, in which we pool together the magnitudes of structural deviations for all 19 structures, exhibits an even better level of significance of conformational changes, i.e., up to 10 Å (Fig. 44). Overall, our results suggest that conformational differences close to the binding site are not representative of the global conformational changes after conformational selection, i.e., they are drawn from different distributions. Finally, to provide a structural context to our P value analysis, in Fig. 4B we map P values onto a surface of an X-ray structure of ubiquitin. Here, we use a structure whose P value curve is closest to the median curve in the rmsd sense. This approach for conformational analysis clearly shows the statistical significance of the deviation in structural changes as a function of distance from the binding site. After conformational selection, the residues close to the binding site change in ways that cannot be explained by global structural changes, hinting at induced fit optimization.

What is the relative magnitude of induced fit in ubiquitin interactions, when compared with conformational selection? The histograms presented in Fig. 5 compare the magnitude of conformational selection with induced fit in the first and the last distance range, representing just the binding site and the whole molecule, respectively. In all four cases, from the point of view of the whole molecule, induced fit is quantitatively of a lower magnitude compared with conformational selection ($\Delta_{\text{global}} > 0$), but it is still significant, especially for side chains. When we focus on just the

Wlodarski and Zagrovic PNAS Early Edition | 3 of 6

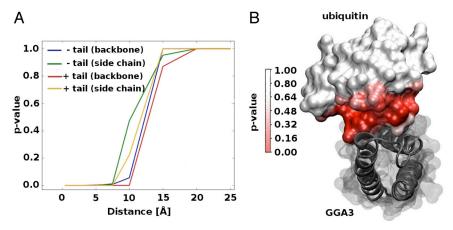


Fig. 4. Pooled statistical significance of induced fit motions in ubiquitin binding. (A) The KS P value analysis conducted on a dataset from all 19 structural pairs pooled together into one distribution for each distance range. (B) Mapping of the calculated KS P values onto a representative ubiquitin surface [structure of ubiquitin bound to ADP-ribosylation-factor-binding protein GGA3 (gray); PDB ID code 1YD8], capturing the P value dependence on the distance from the binding site. The structure shown is closest (in the rmsd sense) in terms of its P value distribution to the median over all 19 structures examined.

binding site, the magnitude of induced fit is still somewhat lower for the backbone compared with conformational selection ($\Delta_{binding-site}$ > 0), but for side chains it is higher ($\Delta_{\text{binding-site}} < 0$). This practically means that side chains are more prone to undergo induced fit as opposed to backbone, especially close to the binding site, where their structural changes can even be greater than the changes caused by conformational selection. Interestingly, if one includes the tail residues (Fig. 5 C and D), the conformational selection appears to be somewhat more dominant than the global induced fit (i.e., distributions are more separated in terms of Δ_{global}) on average, although the local induced fit in the vicinity of the binding site can be sizable, especially in cases where the tail residues participate in binding. Further analysis (local rmsd and P values) carried out on two additional ensembles derived from the same NMR data (47) support our findings presented here. These results are further presented and discussed below and in SI Text and Table S1). Finally, we have also analyzed the correlated motions in the unbound EROS ensemble of ubiquitin. Our findings suggest that there are no major long-range correlations in the unbound state, except for those in the binding site itself (Fig. S4).

Discussion

We have shown that apart from global conformational similarity between unbound and bound structures differences in their local conformation strongly suggest that the residual induced fit after conformational selection is a significant component underlying specificity in noncovalent interactions of ubiquitin. Local structural differences and adjustments are especially pronounced close to the binding site and are described by significantly different distributions of deviations compared with the whole molecule. Moreover, we have shown that the C-terminal tail region of ubiquitin is important in the binding interaction, and that it does undergo extensive structural rearrangements upon binding in many cases. Our results support and extend the picture of protein-protein interactions proposed by Grünberg et al. (11). In the course of a binding event, the structurally closest protein scaffold to the bound conformation is chosen by conformational selection, while subsequently the binding interface is optimized for specific interactions via induced fit (Fig. S5). In this model, the relative magnitude of conformational selection as compared with the magnitude of induced fit could vary for more dynamic proteins and could also depend on protein function. Our results suggest that the magnitude of the residual induced fit in ubiquitin binding is, on the level of the whole molecule, only marginally lower than that of conformational selection, but can locally and per atom be even greater, especially for side

chains. Altogether, our study furthers our understanding of ubiquitin binding and provides a common framework for analyzing all ubiquitin interactions.

In the present study, we have chosen the rmsd measure because it is probably the most widely used measure for structural comparison of biomolecules, and it is the principal tool used in previous analyses of conformational selection (39, 41). By choosing only atom-to-atom rmsd we were able to avoid size dependence present in typical global rmsd calculations. However, we have also tested other approaches, especially for local structural comparison. We have used a nine-residue window sliding down the ubiquitin sequence and have superimposed and calculated the atom-positional rmsd inside this window. In a similar analysis, all atoms within a 6-Å radius sphere around each backbone atom were analyzed and atoms inside this sphere were superimposed and their atom-positional rmsd was calculated. Results from such analyses were in general agreement with the findings presented herein without adding new insight.

In all of our local rmsd analyses, we have chosen an unbound NMR structure with the lowest backbone global rmsd to the corresponding bound X-ray structure. Following Lange and coworkers (39, 41), this, by definition, is the conformationally selected unbound conformer associated with a given X-ray structure. The choice of this unbound structure to compare with the bound one, critically depends not only on the conformational selection model being correct, but also on its operational definition. In this sense, our analysis can be thought of as the study of the residual induced fit after conformational selection, as defined through the global backbone rmsd criteria. This fact touches on a fundamental issue concerning the very definition of conformational selection. Namely, here we have defined conformational selection in the global rmsd sense. However, one could envision a local definition of conformational selection, in which just the binding site would need to be structurally similar in the unbound and the bound state, whereas the rest of the structure could be fairly different. In this sense, the balance between conformational selection and induced fit in binding is in part a matter of definition of the two components.

We have carried out a detailed analysis based on an NMR ensemble (39, 41) that is, to the best of our knowledge, the most detailed available structural description of ubiquitin dynamics, covering a dynamical range from ps to μ s. However, it is known that different structural ensembles can reproduce the NMR data to a similar extent, including ensembles that were generated through refinement (39, 41, 47) or even represent different structural models obtained by other methods such as X-ray crystallography

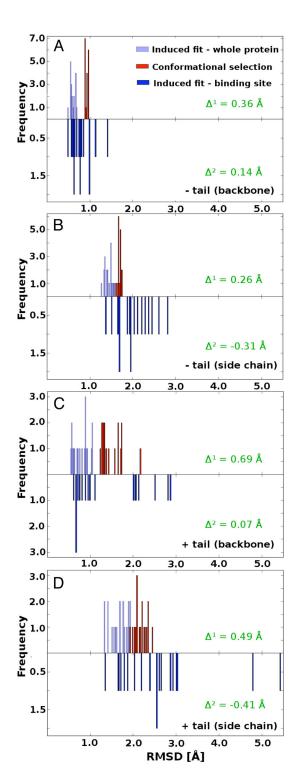


Fig. 5. Relative magnitude of conformational selection and induced fit motions. A histogram presenting the magnitude of conformational changes and induced fit of just the binding site (dark blue) or the whole molecule (light blue), and the conformational selection part of interaction between ubiquitin and its binding partners (red), shown separately for analysis without [backbone (A) and side-chain (B)] and with the C-terminal tail [backbone (C) and side chain (D)]. The calculated Δ parameter describes the probability-weighted distance between two histograms: Δ_{global} — conformational selection versus induced fit of the whole molecule; $\Delta_{\text{binding-site}}$ — conformational selection versus induced fit of the binding site.

(48). For this reason, we have repeated our complete analysis on two additional NMR ensembles generated from the same RDC data as the EROS ensemble [Backrub ensembles generated via Monte Carlo simulations (47)]. All of the principal results obtained with the EROS ensemble are reproduced with these ensembles (see *SI Text*). For example, 47% of the structures in the Backrub 1.2 ensemble (Table S1) without the tail, and 58% with the tail included, exhibit significantly greater fluctuations (P < 0.1) in the region <1.5 Å away from the binding site compared with the molecule as a whole. However, here one should emphasize that the original EROS ensemble still slightly better reproduces the original RDC data (47) and may potentially be somewhat more realistic as it was generated by using time-dependent molecular dynamics simulations.

Finally, our findings could be of use for further development of computational docking methods, which have recently started to incorporate the conformational selection model for ligand–receptor interactions (49). One particularly exciting future development would entail using molecular dynamics simulations to sample bound and unbound ensembles of different interacting partners. The principal advantage of such an approach is that one would obtain information about populations of each conformational state. Altogether, the methods developed herein provide a quantitative basis for comparing induced fit and conformational selection in general and open up ways for similar efforts in different systems.

Materials and Methods

Structural dataset used in our analysis is based on the one used by Lange and coworkers (39, 41). It consists of two ensembles: (i) an ensemble of X-ray structures containing a number of ubiquitin conformers bound to different partners and (ii) an NMR ensemble of ubiquitin structures free in solution, the EROS ensemble $refined \ from\ RDCs\ (39,41).\ The\ first\ ensemble\ consists\ of\ 19\ high-resolution\ X-ray$ structures of ubiquitin bound to different binding partners (Table S2), chosen from the Protein Data Bank (PDB). The selection criteria together with the exact PDB codes are given in SI Text. The NMR EROS ensemble includes 116 conformations of unbound ubiquitin. For additional analysis, we have used two Backrub ensembles described in ref. 47 (two ensembles of 50 structures with maximum segment length 3 and 12 and kT = 2.4 and 1.2, respectively) that also reproduce the RDC data well. However, if not further specified, our analysis refers to the EROS ensemble. In addition to the analysis of ubiquitin without the flexible C-terminal region (residues 71–76 comprising ubiquitin's tail), as done by Lange and coworkers (39, 41), we have performed analysis on the complete structure of ubiquitin (including all 76 residues, when available, or the maximum number of residues, when the structure of the complete molecule was not available, such as in a number of X-ray structures). The reason for including the tail residues is that, in fact, they often directly participate in ubiquitin binding and are in many cases an integral part of ubiquitin's binding site (Fig. S1).

Structural rmsd Analysis. Our structural analysis focused on: (i) ubiquitin's global structural features, encompassing the whole molecule and (ii) local structural features and their dependence on the distance from the binding site. For the global analysis we have used nonmetric MDS [the Sammon mapping algorithm as implemented in R (50) and modified for our purposes] and applied it to a distance matrix produced from backbone atom-positional rmsd values calculated for all pairs from the merged set (X-ray with NMR ensemble). The exact procedure for generating MDS maps is given in SI Text.

Our local structural analysis is based on first finding, for each X-ray structure, one unbound NMR structure with the most similar conformation in the backbone atom-positional rmsd sense. As suggested before, this structure is the one that is conformationally selected for a given binding partner (39, 41). Subsequently, we analyze individual atom-to-atom rmsd (separately for backbone and side-chain atoms) between the atoms in the ubiquitin X-ray structures and the matching atoms from the corresponding conformationally selected NMR structures, as a function of distance from the binding site (Fig. 2). Importantly, these local deviations are analyzed upon global superposition of the two structures by using all backbone atoms for superposition. To acquire the dependence of the deviations on the distance from the binding site, we have grouped all ubiquitin atoms in each pair into nine distance ranges: from 0 to 0.5, 1.5, 3, 5, 7.5, 10, 15, 20, and 25 Å, depending on the distance from the binding site. A given atom is assigned to one of the above ranges if its distance to the closest nonhydrogen atom from the binding partner is less than the sum of the upper bound of a given range and the van der Waals radii of the two atoms. The binding site itself is defined as all of the atoms in the 0.5-Å distance range (51). Note that distance ranges are

Wlodarski and Zagrovic PNAS Early Edition | 5 of 6

cumulative, meaning that if an atom belongs to one group it also belongs to all other groups with larger upper bounds.

Statistical Significance of Structural Differences. We have used the standard two-sample KS test to compare, for each pair of structures, the distribution of atom-to-atom rmsd values in each distance range with the distribution of atomto-atom rmsd values in the last distance range (25 Å) representing the whole molecule. We have chosen this nonparametric test, because the distributions of magnitudes of atomic deviations did not conform well to any distribution for which a parametric test might be used. In addition to applying the above test to individual structural pairs, we have applied our P value analysis to a merged dataset where data from all 19 structural pairs was pooled into one distribution for each distance range. The details about the KS test are given in SI Text.

Quantitative Comparison of Conformational Selection and Induced Fit. One can quantify the magnitude of conformational selection as the average backbone atom-positional rmsd value between all unbound structures and the one chosen to be the most similar to the bound state. Similarly, the calculated local atompositional rmsd values (for pairs of X-ray and NMR structures) can be taken as a measure of the magnitude of induced fit. The differences in magnitude between conformational selection and induced fit in ubiquitin binding is analyzed here via: (i) histograms of average backbone atom-positional rmsd values between all unbound structures and the one chosen to be the most similar to the bound state,

- 1. Fisher E (1894) Einfluss der Konfiguration auf die Wirkung der Enzyme. Ber Dt Chem Ges 27:2985-2993
- 2. Koshland DE (1958) Application of a theory of enzyme specificity to protein synthesis. Proc Natl Acad Sci USA 44:98–104.

 3. Bosshard HR (2001) Molecular recognition by induced fit: How fit is the concept? News
- 4. Ma B, Kumar S, Tsai CJ, Nussinov R (1999) Folding funnels and binding mechanism. Protein Eng 12:713–720.
- 5. Tsai CJ, Kumar S, Ma B, Nussinov R (1999) Folding funnels, binding funnels, and protein function. Protein Sci 8:1181–1190
- 6. Tsai CJ, Ma B, Nussinov R (1999) Folding and binding cascades: Shifts in energy landscapes. Proc Natl Acad Sci USA 96:9970-9972.
- 7. Kumar S, Ma B, Tsai CJ, Sinha N, Nussinov R (2000) Folding and binding cascades: Dynamic landscapes and population shifts. Protein Sci 9:10-19.
- 8. Tsai CJ, Ma B, Sham YY, Kumar S, Nussinov R (2001) Structured disorder and confor-
- mational selection. *Proteins* 44:418–427.

 9. Ma B, Shatsky M, Wolfson HJ, Nussinov R (2002) Multiple diverse ligands binding at a single protein site: A matter of pre-existing populations. Protein Sci 11:184-197.
- James LC, Tawfik DS (2003) Conformational diversity and protein evolution: A 60-year-old hypothesis revisited. Trends Biochem Sci 28:361–368.
- 11. Grünberg R, Leckner J, Nilges M (2004) Complementarity of structure ensembles in protein-protein binding. Structure (London) 12:2125-2136.
- 12. Eisenmesser EZ, et al. (2005) Intrinsic dynamics of an enzyme underlies catalysis. Nature 438:117-121.
- Tobi D, Bahar I (2005) Structural changes involved in protein binding correlate with intrinsic motions of proteins in the unbound state. Proc Natl Acad Sci USA 102:18908-
- Boehr DD, McElheny D, Dyson HJ, Wright PE (2006) The dynamic energy landscape of dihydrofolate reductase catalysis. *Science* 313:1638–1642. Kapur S, Khosla C (2008) Fit for an enzyme. *Nature* 454:832–833.
- Boehr DD, Wright PE (2008) How do proteins interact? Science 320:1429-1430.
- Weikl TR, von Deuster C (2009) Selected-fit versus induced-fit protein binding: Kinetic differences and mutational analysis. Proteins 75:104-110.
- Okazaki K, Takada S (2008) Dynamic energy landscape view of coupled binding and protein conformational change: Induced-fit versus population-shift mechanisms. Proc Natl Acad Sci USA 105:11182-11187.
- Foote J, Milstein C (1994) Conformational isomerism and the diversity of antibodies. Proc Natl Acad Sci USA 91:10370-10374.
- Berger C, et al. (1999) Antigen recognition by conformational selection. FEBS Lett 20. 450:149-153.
- 21. James LC, Roversi P, Tawfik DS (2003) Antibody multispecificity mediated by conformational diversity. Science 299:1362-1367.
- 22. Jimenez R, Salazar G, Baldridge KK, Romesberg FE (2003) Flexibility and molecular ecognition in the immune system. Proc Natl Acad Sci USA 100:92–97
- Kovrigin EL, Loria JP (2006) Enzyme dynamics along the reaction coordinate: Critical role of a conserved residue. Biochemistry 45:2636–2647.
- Koglin A, et al. (2008) Structural basis for the selectivity of the external thioesterase of the surfactin synthetase. *Nature* 454:907–911.
 Osborne MJ, Schnell J, Benkovic SJ, Dyson HJ, Wright PE (2001) Backbone dynamics in
- dihydrofolate reductase complexes: Role of loop flexibility in the catalytic mechanism.
- Biochemistry 40:9846–9859.

 26. Pontiggia F, Zen A, Micheletti C (2008) Small- and large-scale conformational changes of adenylate kinase: A molecular dynamics study of the subdomain motion and
- mechanics. *Biophys J* 95:5901–5912.

 27. Tang C, Schwieters CD, Clore GM (2007) Open-to-closed transition in apo maltosebinding protein observed by paramagnetic nmr. Nature 449:1078-1082.
- 28. Xu Y, et al. (2008) Induced-fit or preexisting equilibrium dynamics? Lessons from protein crystallography and MD simulations on acetylcholinesterase and implications for structure-based drug design. Protein Sci 17:601-605.

capturing the magnitude of conformational selection for all 19 structures, and (ii) histograms of structural atom-to-atom deviations from the first and the last distance range (see above), capturing the induced fit of the binding site and the whole molecule, respectively. Additionally, to quantitatively describe relative differences between induced fit and conformational selection, we have calculated probability-weighted "distance" between distributions:

$$\Delta = \sum_{x_i, f_i \in D_1} \sum_{x_i, f_i \in D_2} \left(\frac{x_i f_i}{n_i} - \frac{x_j f_j}{n_j} \right),$$
[1]

where D_1 and D_2 are distributions of magnitudes of conformational selection and induced fit, respectively, x are rmsd values, f are frequencies, and n is a number of points in each distribution.

ACKNOWLEDGMENTS. We thank I. Dikic, V. S. Pande, S. Bacallado, and the members of the Laboratory of Computational Biophysics at the Mediterranean Institute for Life Sciences for useful comments on the manuscript and T. Kortemme (University of California, San Francisco, CA) for providing the Backrubrefined ensembles of ubiquitin. This work was supported in part by the National Foundation for Science, Higher Education, and Technological Development of Croatia, a European Molecular Biology Organization installation grant (to B.Z.), Unity Through Knowledge Fund Grant UKF 1A (to B.Z.), and a Federation of European Biochemical Societies Summer Research Fellowship (to T.W.).

- 29. Yang L, et al. (2007) Insights into equilibrium dynamics of proteins from comparison of NMR and X-ray data with computational predictions. Structure (London) 15:741-749.
- 30. Gsponer J, et al. (2008) A coupled equilibrium shift mechanism in calmodulin-mediated signal transduction. Structure (London) 16:736-746.
- 31. Najmanovich R, Kuttner J, Sobolev V, Edelman M (2000) Side-chain flexibility in proteins upon ligand binding. Proteins 39:261-268.
- 32. Li X, Keskin O, Ma B, Nussinov R, Liang J (2004) Protein-protein interactions: Hot spots and structurally conserved residues often locate in complemented pockets that preorganized in the unbound states: Implications for docking. J Mol Biol 344:781-795.
- 33. Rajamani D, Thiel S, Vajda S, Camacho CJ (2004) Anchor residues in protein-protein interactions. Proc Natl Acad Sci USA 101:11287-11292.
- 34. Yogurtcu ON, Erdemli SB, Nussinov R, Turkay M, Keskin O (2008) Restricted mobility of conserved residues in protein-protein interfaces in molecular simulations. Biophys J 94:3475-3485.
- 35. Sullivan SM, Holyoak T (2008) Enzymes with lid-gated active sites must operate by an induced fit mechanism instead of conformational selection. Proc Natl Acad Sci USA 105:13829-13834.
- 36. Lipari G, Szabo A (1982) Model-free approach to the interpretation of nuclear magnetic resonance relaxation in macromolecules. J Am Chem Soc 104:4546-4570.
- 37. Akke M, Palmer AG (1996) Monitoring macromolecular motions on microsecond to millisecond time scales by R_{1p}-R₁ constant relaxation time NMR spectroscopy. J Am Chem Soc 118:911-912.
- 38. Meiler J, Prompers JJ, Peti W, Griesinger C, Brüschweiler R (2001) Model-free approach to the dynamic interpretation of residual dipolar couplings in globular proteins. JAmChem Soc 123:6098-6107.
- 39. Lange OF, et al. (2008) Recognition dynamics up to microseconds revealed from an RDC-derived ubiquitin ensemble in solution. Science 320:1471-1475
- 40. Markwick PRL, Malliavin T, Nilges M (2008) Structural biology by NMR: Structure, dynamics, and interactions. PLoS Comput Biol 4:e1000168.
- 41. Lakomek N, et al. (2008) Residual dipolar couplings as a tool to study molecular recognition of ubiquitin. Biochem Soc Trans 36:1433-1437. 42. Hicke L, Schubert HL, Hill CP (2005) Ubiquitin-binding domains. Nat Rev Mol Cell Biol
- 6:610-621 43. Harper JW, Schulman BA (2006) Structural complexity in ubiquitin recognition. Cell
- 124:1133-1136.
- 44. Mukhopadhyay D, Riezman H (2007) Proteasome-independent functions of ubiquitin in endocytosis and signaling. Science 315:201-205. 45. Ikeda F, Dikic I (2008) Atypical ubiquitin chains: New molecular signals. EMBO Rep
- 9:536-542. 46. Lindorff-Larsen K, Best RB, Depristo MA, Dobson CM, Vendruscolo M (2005) Simulta-
- neous determination of protein structure and dynamics. Nature 433:128-132.
- 47. Friedland GD, Lakomek N, Griesinger C, Meiler J, Kortemme T (2009) A correspondence between solution-state dynamics of an individual protein and the sequence and conformational diversity of its family. PLoS Comput Biol 5:e1000393.
- 48. Best RB, Lindorff-Larsen K, DePristo MA, Vendruscolo M (2006) Relation between native ensembles and experimental structures of proteins. Proc Natl Acad Sci USA 103:10901-10906
- 49. Chaudhury S, Gray JJ (2008) Conformer selection and induced fit in flexible backbone protein-protein docking using computational and NMR ensembles. J Mol Biol 381:1068-1087.
- 50. R Development Core Team (2008) A Language and Environment for Statistical Computing (R Foundation for Statistical Computing, Vienna).
- 51. Tsai CJ, Lin SL, Wolfson HJ, Nussinov R (1996) A dataset of protein-protein interfaces generated with a sequence-order-independent comparison technique. J Mol Biol 260:604-620

Supporting Information

Wlodarski and Zagrovic 10.1073/pnas.0906966106

SI Text

Selection Criteria for the X-Ray Structures of Bound Ubiquitin. The structural dataset used here (Table S2) was created by filtering the set used by Lange and coworkers (1, 2) and applying more stringent selection criteria to it to make it, albeit smaller, more representative of the full diversity of ubiquitin interactions. Our main objective was to create a nonbiased dataset containing all known ubiquitin binding interfaces, but exhibiting as little as possible sampling bias for any particular interaction or ligand. Briefly, we chose only those ubiquitin X-ray structures that: (i) were cocrystallized with a covalently or noncovalently bound partner (including polyubiquitin chains), and (ii) are bound to different ligands except when the same ligand is bound to two different binding sites, or when the backbone rmsd between the two ubiquitin conformers bound to the same ligand in the same binding site exceeds 0.8 Å. Although smaller, the present dataset does fully describe all binding interfaces found in ubiquitin via crystallographic studies (3).

MDS. The procedure for generating MDS maps was as follows: (i) we used MDS to map all 19 X-ray structures into 2D space, (ii) the distances between points in this map were then kept fixed during the next part of analysis, and (iii) we used MDS to map each NMR structure separately to the fixed X-ray MDS map. In this way, we obtained a set of points in 2D space (Fig. 1), each representing one X-ray or NMR structure, and distributed in such a way that the distances between points are proportional to the corresponding rmsd values. However, because of the above modifications in the MDS algorithm, the position of points representing a given NMR structure is not influenced by other NMR structures, but only by the whole X-ray ensemble. The algorithm performs the above dimensionality reduction by minimizing the Sammon stress function:

$$S_{sam} = \frac{1}{\sum_{i < j} d_{ij}} \sum_{i < j} \frac{(d_{ij} - d_{ij}^*)^2}{d_{ij}}.$$
 [1]

where d_{ij}^* and d_{ij}^* are distances (rmsd value) between structures in the original and reduced space, respectively.

KS Test. The KS test is a statistical significance test with a null hypothesis that the two distributions that are being compared are drawn from the same continuous distribution. Its statistic for a two sample test is defined by:

$$D_n = \sup_{x} |F_n(x) - F_{n'}(x)|,$$

where $F_n(x)$ and $F_{n'}(x)$ are empirical distribution functions for the first and second of the compared distributions, respectively:

$$F_n(x) = \frac{1}{n} \sum_{i=1}^n I(X_i \le x),$$

where X_i are elements of distribution and $I(X_i \leq [lt]x)$ is the indicator function, which is equal to 1 when $X_i \leq [lt]x$, and 0 otherwise. The null hypothesis is rejected on an α -level of significance when

$$\sqrt{\frac{nn'}{n+n'}}D_{n,n'}>K_{\alpha},$$

where K_{α} is obtained from:

$$Pr(K \le K_{\alpha}) = 1 - \alpha,$$

where K is drawn from the Kolmogorov distribution. Here, it should be noted that the KS test assumes that the samples are drawn independently from the nonparametric distribution, and by using it we neglect, to a first approximation, the possibility of local correlations.

Analysis of Additional Structural Ensembles Refined from the Lange and Coworkers (1, 2) RDC Data. The additional Backrub ensembles used in our analysis were generated by T. Kortemme and coworkers (4) using Monte Carlo Backrub moves with 3- and 12-residue segments at kT = 2.4 and 1.2, respectively. We have redone the complete local rmsd and P value analysis for these two ensembles to verify our findings obtained with the EROS ensemble. The results of this analysis are summarized and compared with our principal findings in Table S1. The δ -values show that local conformational deviations close to the binding site (0.5-Å range) are greater than the global deviations (25-Å range), to a degree that is in some cases greater and in some lower compared with the EROS ensemble. In general, the δ-values for the cases with the C-terminal tail included (relevant for induced fit) are significantly greater for both Backrub ensembles than for the EROS ensemble (e.g., 133% versus 78% for backbone for Backrub 1.2 and EROS, respectively), whereas without the C-terminal tail included, they are on average lower (e.g., 13% versus 30% for backbone, for Backrub 1.2 and EROS, respectively). Moreover, the P value analysis for the Backrub ensembles suggests that conformational changes close to the binding site are more pronounced than the global conformational changes, which is in agreement with our main findings. For example, for distances up to 5 Å from the binding site, for \approx 50% of structures for both Backrub ensembles with and without the tail included, P < 0.1. Altogether, these results obtained on other structural ensembles, which still well reproduce the RDC data, support all of our main conclusions obtained for the EROS ensemble. Here, it should be mentioned that although the Backrub ensembles agree well with the RDC data from Lange and coworkers (1, 2), the original data were not used directly in the course of their refinement, but rather in a selection procedure (4). The structures generated in a Backrub Monte Carlo procedure were filtered for those that agree well with experimental data. This is the fundamental difference between these ensembles and the EROS ensemble, and it may be that because of it, the latter ensemble may be somewhat more physical as it was generated using time-dependent molecular dynamics.

Analysis of Correlations in the Fluctuations of the Unbound State of Ubiquitin. We have used the g_covar routine included in the Gromacs package (5) to obtain the mass-weighted variance-covariance matrix of structural fluctuations for the EROS ensemble of unbound ubiquitin structures after backbone-atom least-squares superposition. The calculated diagonal elements of the variance-covariance matrix, $\sigma_{i,i}$, are variances of positions of each of the atoms, whereas the off-diagonal elements, $\sigma_{i,j}$, are covariances between the ith and the jth atom, defined by: $\sigma_{i,j} = \langle (\mathbf{r}_i - \langle \mathbf{r}_i \rangle (\mathbf{r}_j - \langle \mathbf{r}_j \rangle) \rangle$, where \mathbf{r}_i is the instantaneous position of the ith atom (a three-element vector containing the x, y, and z positions) and $\langle \mathbf{r}_i \rangle$ is its average position over the entire ensemble. The variance-covariance matrix was converted into a

correlation matrix by normalizing each covariance element in the matrix by the associated variances (6):

$$c_{i,j} = \frac{\sigma_{i,j}}{\sqrt{\sigma_{i,i}\sigma_{i,j}}}.$$

In Fig. S4, we show a correlation map for the unbound EROS ensemble of ubiquitin. To relate this to binding, we have marked with crosses on the correlation map all of the atoms belonging to the residues in the binding site of ubiquitin for structure 1YD8 (see also Fig. 4). This is the structure whose induced-fit motions in the course of binding are, in terms of their magnitude, closest to the average obtained for all bound structures analyzed in this

- Lange OF, et al. (2008) Recognition dynamics up to microseconds revealed from an RDC-derived ubiquitin ensemble in solution. *Science* 320:1471–1475.
- Lakomek N, et al. (2008) Residual dipolar couplings as a tool to study molecular recognition of ubiquitin. Biochem Soc Trans 36:1433–1437.
- 3. Hicke L, Schubert HL, Hill CP (2005) Ubiquitin-binding domains. *Nat Rev* Mol Cell Biol 6:610–621.
- Friedland GD, Lakomek N, Griesinger C, Meiler J, Kortemme T (2009) A correspondence between solution-state dynamics of an individual protein and the sequence and conformational diversity of its family. PLoS Comput Biol 5:e1000393.

study. The variances of the binding-site atoms themselves are marked with crosses on the diagonal, together with their covariances off of the diagonal. Clearly, the atoms in the binding site exhibit significant correlated motions already in the unbound state. However, except between each other, none of these atoms exhibit significant correlated motions with any other atoms: all of the major correlated motions occur only within the binding site. One may speculate that precisely these correlated motions of the binding-site atoms contribute to conformational selection. Interestingly, there are no major long-range correlated motions between the binding site and other parts of the ubiquitin's structure, a feature that is sometimes seen in allosterically controlled proteins.

- 5. Van Der Spoel D, et al. (2005) GROMACS: Fast, flexible, and free. J Comput Chem 26:1701–1718.
- Theobald DL, Wuttke DS (2008) Accurate structural correlations from maximum likelihood superpositions. PLoS Comput Biol 4:e43.

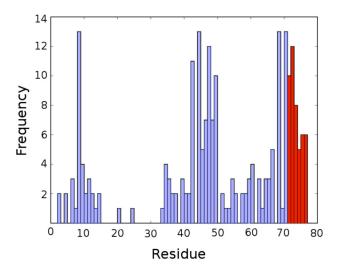
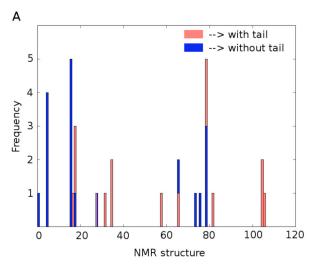


Fig. S1. The role of ubiquitin's tail in binding. Frequency of occurrence of ubiquitin's residues as a part of the binding interface, defined as all ubiquitin atoms in the 0.5-Å distance range (see *Materials and Methods*) and calculated from the X-ray ensemble; the tail residues (residues 71–76) are colored red.



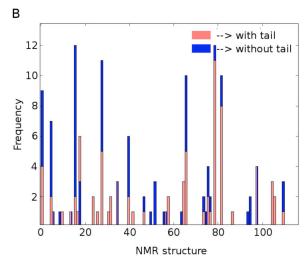


Fig. S2. Characterization of the conformationally selected structures. (A) The frequency of how often each unbound NMR structure of ubiquitin from the EROS ensemble is the one that is the most similar to one of the 19 bound X-ray structures in the rmsd sense (red, with tail; blue, without tail). Overall, 63% of X-ray structures are captured by the top three most similar NMR structures without the tail, whereas this number drops to 52% with the tail included. (B) The frequency of how often each unbound NMR structure of ubiquitin from the EROS ensemble has rmsd value within a range of 0.1 Å from the rmsd value of the most similar NMR structure to each X-ray.

Distortion plot

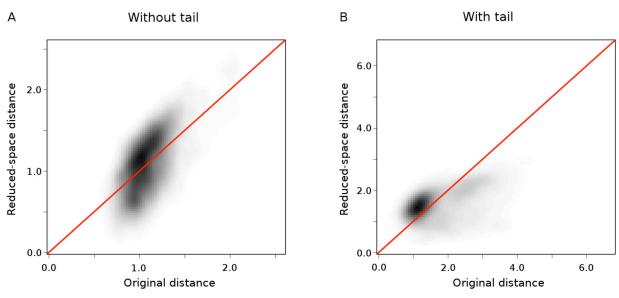


Fig. S3. MDS distortion plots. Density distortion plots describe how well the MDS method (Sammon mapping) reduces the dimensionality of the data by plotting original distances against mapped (reduced-space distance). We have used density plots to better present the most common relation between this two distances. For ideal mapping points should be spread close to red curve. An additional measure of the quality of mapping in MDS is the value of Sammon stress function, which is minimized during MDS calculations. For our dataset we obtain mean S_{sam} equal 0.056 with standard deviation: 0.016 and 0.129 with 0.055, without and with the C-terminal tail, respectively.



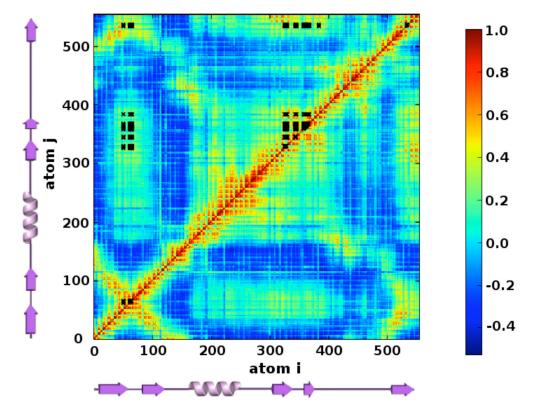
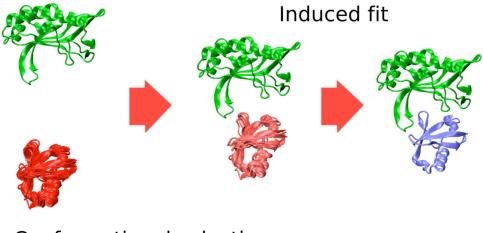


Fig. S4. Analysis of the correlations in the fluctuations of the unbound state of ubiquitin. Correlation map (normalized variance-covariance matrix) for the EROS ensemble of ubiquitin. The variances and covariances for atoms belonging to the residues in the binding site of ubiquitin for a bound complex structure 1YD8 are depicted with crosses. The secondary structure elements in ubiquitin, in the N to C direction, are depicted next to the axes.



Conformational selection

Fig. S5. Model of protein-protein interactions for ubiquitin. The first step in binding entails conformationally selecting an unbound conformation that is structurally similar to the bound state conformation from the spectrum of available conformers. After conformational selection, one observes a population shift in the direction of those conformers that are structurally optimized for binding. The last step is induced fit optimization of the interactions in the binding site, which, as shown in the present manuscript for ubiquitin, could be significant. For illustration purposes, here we use the TSG101 protein in complex with ubiquitin (PDB ID code 1S1Q).

Other Supporting Information Files

Table S1 (PDF)
Table S2 (PDF)

Table S1. Comparative analysis of the EROS (1, 2) and two Backrub ensembles (4)

								%	% of structures with p-value < 0.1	s with p-valu	e < 0.1
			RMSD 0.5 Å	RMSD 25Å	δ value	p-value median	p-value average	0.5 Å	1.5 Å	3.0 Å	5.0 Å
EROS	without tail	qq	0.78 +/- 0.22	90.0 -/+ 09.0	30 %	0.29	3.5667e-09	33,00%	53,00%	53,00%	37,00%
		os	2.00 +/- 0.38	1.42 +/- 0.08	41 %	0.06	< 2.2e-16	61,00%	%00'62	74,00%	42,00%
	with tail	qq	1.41 +/- 0.79	0.79 +/- 0.17	% 82	0.04	< 2.2e-16	72,00%	74,00%	47,00%	53,00%
		os	2.58 +/- 1.00	1.67 +/- 0.20	54 %	0.07	< 2.2e-16	63,00%	%00'89	%00%99	63,00%
BACKRUB 1.2	without tail	qq	0.76 +/- 0.23	20'0 -/+ 29'0	13 %	0.52	0.42	28,00%	47,00%	37,00%	31,00%
		os	2.01 +/- 0.38	1.56 +/- 0.12	29 %	0.18	4.1660e-10	28,00%	21,00%	21,00%	31,00%
	with tail	qq	1.98 +/- 2.12	0.85 +/- 0.26	133 %	0.18	< 2.2e-16	44,00%	28,00%	58,00%	37,00%
		SC	2.28 +/- 0.51	1.73 +/- 0.18	32 %	0.09	< 2.2e-16	53,00%	53,00%	53,00%	47,00%
BACKRUB 2.4	without tail	qq	0.65 +/- 0.29	60.0 -/+ 72.0	14 %	0.29	0.03	27,00%	42,00%	37,00%	31,00%
		SC	2.02 +/- 0.45	1.51 +/- 0.12	34 %	0.10	8.2635e-10	50,00%	42,00%	37,00%	26,00%
	with tail	qq	1.88 +/- 2.14	0.75 +/- 0.27	151 %	0.17	< 2.2e-16	50,00%	58,00%	58,00%	47,00%
		sc	2.40 +/- 0.43	1.63 +/- 0.20	47 %	0.16	< 2.2e-16	47,00%	47,00%	47,00%	47,00%

Analysis was carried out for structures with and without the C-terminal tail and for backbone (bb) and side-chain atoms (sc), separately. The rmsd calculations are presented for the binding-site atoms (0.5 Å range) and the whole molecule (25 Å range). δ value is defined as $(\text{rmsd}_{0.5\text{Å}} - \text{rmsd}_{25\text{Å}})/\text{rmsd}_{25\text{Å}})$ x 100%. The median of the p-value distribution is presented for the binding site only (0.5 Å range). The average p-value (for pooled data) is given for the binding site only. The percentage of structures with p-value < 0.1 is given for four distance ranges from the binding site.

Table S2. PDB codes and chain names of structures from the analyzed ensemble of bound ubiquitin conformers

PDB code	Ubiquitin	Binding partner
1NBF	С	В
1AAR	A	В
IAAK	В	A
1CMX	В	A
1P3Q	V	R,Q
1S1Q	D	C
1TBE	A	В
	В	A
1UZX	В	A
1XD3	В	A
1 YD8	U	Н
2AYO	В	A
2C7M	В	A
2C7N	J	C
	Н	G
DD3G	A	P
2D3G	В	P
2FCQ	A	В
2G45	В	A