A disordered encounter complex is central to the yeast Abp1p SH3 domain binding pathway

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Abstract

Protein-protein interactions are involved in a wide range of cell processes. These interactions often involve intrinsically disordered proteins (IDPs) and protein binding domains. However, the details of IDP binding pathways are hard to characterize using experimental approaches that cannot always capture low populations of intermediate states. SH3 domains are common protein interaction domains that typically bind proline-rich disordered segments and are involved in cell signaling, regulation, and assembly. We hypothesized, given the flexibility of SH3 binding peptides, that their binding pathways include multiple steps important for function. Molecular dynamics simulations were used to characterize the binding between the yeast Abp1p SH3 domain (AbpSH3) and a proline-rich IDP, ArkA, to capture all steps along the pathway. Before binding, the N-terminal segment 1 of ArkA is pre-structured and adopts a polyproline II helix, while segment 2 of ArkA (C-terminal) adopts a 3_{10} helix, but is far less structured than segment 1. As segment 2 interacts with AbpSH3, it becomes more structured, but retains flexibility even in the fully engaged state. Binding simulations revealed that ArkA enters a flexible encounter complex before forming the fully engaged bound complex. In the encounter complex, transient nonspecific hydrophobic and long-range electrostatic contacts form between ArkA and the binding surface of SH3. The encounter complex ensemble includes conformations with segment 1 in both the forward and reverse orientation, suggesting segment 2 may play a role in stabilizing the correct binding orientation. While the encounter complex forms quickly, the slow step of binding is the transition from the disordered encounter ensemble to the fully engaged state. In this transition, ArkA makes specific contacts with AbpSH3 and buries more hydrophobic surface. Simulating the binding between ApbSH3 and ArkA provides insight into the role of encounter complex

intermediates and nonnative hydrophobic interactions for other SH3 domains and IDPs more generally.

Author Summary

Complex cellular processes are mediated by interactions between proteins, and to determine how these physical interactions affect cellular function we need to understand the protein binding pathway. Many protein interaction domains, such as the SH3 domain, bind to intrinsically disordered proteins in a coupled folding and binding process. Using molecular dynamics simulations, we find that the binding of the disordered ArkA peptide to the yeast Abp1p SH3 domain proceeds through a flexible, disordered encounter complex before reaching a more stable fully bound state. This encounter complex is stabilized by nonspecific long-range electrostatic interactions and nonspecific hydrophobic interactions between the peptide and domain. Our simulations highlight the important role of hydrophobic interactions in the entire SH3 binding process, both nonspecific hydrophobic contacts in the encounter complex and specific hydrophobic contacts in the fully bound complex. The encounter complex could be key to understanding the functional behavior of SH3 domain interactions, as it forms very quickly and the transition from the encounter to the fully bound complex is slower. In cells, the SH3 domain may form an encounter complex quickly and nonspecifically with many potential binding partners, allowing it to search for the correct sequence before completing the binding process.

Introduction

Protein-protein interactions are involved in most cellular processes, especially cellular signaling. These interactions often involve binding of small protein domains to intrinsically

disordered proteins (IDP), but the binding pathways for these interactions are not always well understood [1-3]. Regions of disorder are now known to be present in between 25% and 41% of eukaryotic proteins, and can exhibit functional diversity by having multiple interaction partners [4]. IDPs also tend to bind with lower affinity to their partners than folded proteins, with fast association and dissociation [3,5-7]. This fast binding and unbinding, along with the fast turnover rates of IDPs within cells, allows for regulation of processes that require rapid responses [8]. Despite fast on and off rates, IDP binding interactions must still be very specific in order to relay signals accurately [3,5,6,9,10]. To fully understand how IDPs bind their partners, it is essential to characterize the complete binding pathway in addition to analyzing the structure of the bound protein-protein complex.

IDPs often bind to folded proteins through a process that takes place in at least two steps [11-15]. Binding typically begins with the creation of an encounter complex ensemble when the IDP "dances" on top of the domain it is binding to before transitioning to a more structured fully engaged state through a process of induced fit folding [13,15-19]. IDPs are well suited to quickly forming this initial encounter complex because they have a larger capture radius than folded proteins of the same length, since they generally adopt a more extended conformational ensemble [20]. Electrostatic interactions have been shown to often drive the formation of encounter complex ensembles and can even accelerate association beyond the diffusion limited rate, predominantly by electrostatic orientational steering [5,11,13,21-23]. Additionally, if one segment of the IDP possesses more intrinsic pre-folded structure, binding may proceed starting with this segment and then extending to the rest of the sequence which folds upon interaction with the partner protein [11,24]. Thus, pre-formed secondary structure can improve affinity and influence the binding pathway, including the nature of intermediate states [25,26], but too much structure can slow

binding without improving affinity [11,27]. Additionally, the ability of IDPs to form nonnative contacts during binding, and the presence of significant disorder even after binding can also be important for IDP function [20,27,28]. This underscores the importance of understanding the interactions at play during the binding process as well as in the fully engaged complex.

In addition to the nature of the intermediate states in the binding pathway, the relative heights of free-energy barriers in the pathway dictate the binding kinetics and are therefore critical to function. The transition state for binding can either precede or follow the encounter complex intermediate [5]. Fast-binding proteins are canonically thought to bind in a diffusion-limited manner, and therefore experience a rate determining transition state that precedes the encounter complex [5]. However, weaker IDP complexes with short lifetimes may exhibit binding kinetics that are different in nature from higher affinity complexes [29], and a few IDPs have been shown to associate quickly to form an encounter complex followed by a slower transition to the fully engaged complex [14,17,19,30-35]. Because of their fast binding and dissociation rates and short-lived intermediate states, IDP binding often appears as two-state in NMR experiments [36-38] and stopped-flow experiments [39-41]. Due to experimental challenges, for many IDPs the specific binding pathway, including the nature of binding intermediates and the timescale of their formation, is still unknown.

Computer simulations have been a valuable tool for examining the binding pathway in more detail than can be obtained through experiments. Initially, coarse-grained molecular dynamics (MD) simulations based on the topology of the fully engaged complex indicated that the initial step in the binding process for IDPs might often be the formation of a flexible encounter complex [13,35,42-48]. Another strategy to simulate IDP binding and characterize the encounter complex with limited computational power has been to conduct atomistic MD using an advanced

sampling algorithm [49], such as multicanonical MD [50]. More recently, advancements in both hardware and the accuracy of force fields have enabled unbiased atomistic MD simulations of IDP binding on the microsecond timescale [51]. Those unbiased MD simulations that have explicitly examined the IDP binding pathway generally reveal a fast initial associate between the IDP and its partner, followed by a slower evolution into the fully engaged complex [23,52-56]. However, the details of the binding pathway, including the nature of intermediates in the binding process, has yet to be determined for most IDPs and their binding partners.

One common IDP binding domain is the SH3 domain. It is conserved through more than one billion years of evolution from yeast to humans, and frequently occurs in protein-protein interaction modules, often involving cellular signaling, assembly, or regulation [57]. SH3 domains bind disordered proline-rich target peptides that usually contain a PxxP motif, where x can be any residue [58,59]. This PxxP motif forms a polyproline type II (PPII) helix in the bound complex and is flanked by specificity elements, which often include positively charged lysine or arginine residues [57,60]. The PxxP motif, which is pseudo-palindromic, has been observed to bind the SH3 domain surface in two different orientations (class I and class II) depending on the location of positively charged residues either N or C terminal to the PxxP motif (+xxPxxP or PxxPx+) [61,62]. Despite a very similar binding motif and bound structure, SH3 domains perform a wide variety of different functions in different contexts that require specific binding interactions and biophysical properties [63-65]. Understanding the SH3 domain-peptide binding process may help to reveal the mechanism for the functional diversity of SH3 domains and serve as a model for understanding binding properties of extended IDPs.

Previous studies of proline-rich peptides binding to SH3 domains indicated that the fully engaged bound complexes often exhibit conformation exchange between different bound states

[37,38,66]. However, information about the binding pathway is more limited, as NMR experiments on SH3 domains often indicate two-state binding, possibly due to fast exchange of an encounter complex with either the fully engaged bound state or the unbound state [23,36-38]. One study of SH3-peptide binding found that the transition state for binding is stabilized by long-range electrostatic interactions, but there is less electrostatic enhancement to the binding rate than for folded proteins, which form more short-range electrostatic interactions in the transition state [29]. Simulation studies of the C-CRK N-terminal SH3 domain binding to a proline rich peptide have also indicated that electrostatic interactions are important for the formation of the highly dynamic encounter complex, which transitions to the fully engaged complex when the PPII helix locks into the hydrophobic grooves of the binding site [21,23,54]. This is somewhat in contrast to the picture that hydrophobic interactions are most important for stabilizing the encounter complex for IDPs [17], and it is not clear if electrostatic interactions are the dominant type of interaction in stabilizing SH3 encounter complexes.

We have used all-atom molecular dynamics (MD) simulations to characterize the initial binding interaction between an SH3 domain of yeast Actin Binding Protein 1 (Abp1p) and ArkA, a disordered region of the yeast actin patch kinase, Ark1p. Abp1p is involved in assembly of the actin cytoskeleton through localization of cortical actin patches, actin organization, and endocytosis [67,68]. While several other sequences are known to bind the Abp1p SH3 domain (AbpSH3), ArkA is the partner with the highest affinity for the domain [69]. The structures of AbpSH3 alone and bound to ArkA (Fig 1A) have been solved by x-ray crystallography and NMR, respectively [67,69]. We focus on the binding of a 12-residue truncation of ArkA (residues Lys(3) to Lys(-8)) that binds AbpSH3 with a K_d of 1.7 μM and is comprised of an N-terminal segment containing the PxxP motif and an adjacent C-terminal segment containing key specificity elements

(we use a standard numbering system for peptide positions based on [61], as shown in Fig 1B) [69].

B
$$ACE - K_{(3)}\ P_{(2)}\ T_{(1)}\ P_{(0)}\ P_{(-1)}\ P_{(-2)}\ K_{(-3)}\ P_{(-4)}\ S_{(-5)}\ H_{(-6)}\ L_{(-7)}\ K_{(-8)} - NHE$$

Fig 1. Description of system studied. A) Surface view of AbpSH3 bound to ArkA from NMR [69], showing the two binding surfaces (SI (red) and SII (blue)) with bound ArkA in gray (seg1), magenta (Lys(-3)) and green (seg2). The N and C-termini are labeled. B) Sequence of ArkA used in all simulations with seg1 shown in black, the central lysine in magenta, and seg2 in green. The capping groups on the C and N-terminal ends are also shown.

AbpSH3 has the typical SH3 fold with a five-stranded β-sandwich and long RT-loop, which is involved in ArkA binding [69]. The 12-residue ArkA peptide contains three Lys residues (Fig 1B), giving it a net positive charge, while the AbpSH3 domain has a net negative charge of -12. Thus, electrostatic attraction contributes to the affinity between the peptide and domain. ArkA

can be described as two segments (seg1 and seg2) where seg1 is the N-terminal proline rich end and seg2 is the C-terminal segment (Fig 1B). The proline rich seg1 interacts with AbpSH3 in the typical class II orientation, with a PxxPx+ sequence that forms a PPII helix with each Px well packed into a groove [69]. The region which binds to the PxxP motif is referred to as surface I (SI) (Fig 1A). The C-terminal seg2 forms a 3_{10} helix in the NMR structure and makes contacts on a region of AbpSH3 distinct from SI, referred to as surface II (SII) (Fig 1A). The conserved Lys(-3) serves as the dividing residue between seg1 and seg2, and binds between SI and SII in a negatively charged 'specificity pocket', packing against a Trp side chain [69]. Previous NMR experiments have shown that seg1, containing the PxxP motif, can bind to AbpSH3 without seg2, but does not fully engage the binding surface [38]. Seg2 alone, on the other hand, shows no detectable binding by NMR titration [38]. The role of each segment in the full binding pathway has not previously been investigated.

Using MD simulations, we found that ArkA initially forms a heterogeneous encounter ensemble, followed by the tight binding of seg1 and seg2 in the correct orientation with the formation of specific contacts (Fig 2). Significantly, ArkA forms many nonnative contacts in this encounter ensemble, but they are restricted to the canonical highly acidic binding surface of the AbpSH3 domain. Seg1 is largely pre-structured in a PPII helix and only needs to lock into the grooves of SI to bind, while seg2 is more conformationally flexible. The PPII helix in seg1 can bind in a reverse orientation in the encounter complex, and seg2 may be important for stabilizing the correct orientation of ArkA on the binding surface. Nonspecific hydrophobic and long-range electrostatic interactions stabilize the encounter complex, while specific hydrophobic interactions form only on transition from the encounter complex to the fully engaged state. Our simulations show that step 1 of binding is two orders of magnitude faster than the overall association rate

determined by NMR relaxation dispersion experiments. Our binding model also explains the greater influence of hydrophobic interactions on binding compared to long-range electrostatic interactions, which likely only affect step 1 of the binding pathway. Overall, we have gained an understanding of the different interactions of the two ArkA peptide segments with AbpSH3 along the binding pathway. This provides insights for how binding of this common interaction domain in other proteins may be tailored to meet their specific functional needs.

Fig 2. Proposed ArkA-AbpSH3 binding model. Two-step binding pathway with ArkA shown as a line and AbpSH3 as a circle. The heterogeneous encounter complex is proposed as an intermediate state between the unbound state and fully engaged complex. The forward and reverse rate constants for each step are labeled.

Methods

MD Simulations

MD simulations were run on four constructs: ArkA bound to AbpSH3 (bound simulations), ArkA alone (unbound simulations), ArkA binding to AbpSH3 (ArkA binding simulations), and ArkA seg1 binding to AbpSH3 (seg1 binding simulations). The bound simulations were all started from the lowest energy NMR structure of ArkA bound to AbpSH3 (PDB: 2RPN) [69]. Before running the simulations, the ArkA sequence was truncated to the 12-residue (Fig 1B) construct and a capping acetyl group was added to the ArkA N-terminus. Two different starting structures were used to initiate the unbound simulations. One starting structure was from the NMR structure of ArkA bound to AbpSH3 and the other was a fully extended peptide. For the ArkA and seg1 binding simulations, the peptide construct was placed at least 10 Å from AbpSH3 to ensure that the peptide and domain were not interacting at the beginning of the binding simulations (simulations were run with a nonbonded cutoff distance of 9 Å). For the binding simulations, the starting structure of both ArkA and seg1 came from the ArkA unbound simulations, and the AbpSH3 structure came from the unbound crystal structure (PDB: 1JO8) [67]. The effective concentration of the protein in our simulations was between 3 and 4 mM, which is close to the experimental concentration of 1 mM. In all constructs, ArkA or seg1 were edited to have a capping acetyl group on the N-terminus (ACE) and a capping amide group on the C-terminus (NHE); except the bound simulations, which only have the acetyl group.

All simulations were run on Amber 16 using the Amber ff14SB forcefield [70], and the binding simulations were run with dihedral angle modifications that improve accuracy for the energy barrier between *cis* and *trans* states of the peptide bond [71]. The CUDA version of pmemd in Amber 16 was used to run the simulations on GPUs [72]. All simulations were solvated with TIP3P-FB water [73]. The bound simulations were solvated such that the edge of the box was at least 9 Å from any peptide or protein. Binding simulations were solvated with the edge at least 12

Å from any peptide or protein atom and adjusted to have an equal volume. The unbound simulations were solvated with water 15 Å from the edge of the peptide. Salt ions were added to neutralize each system: 10 sodium ions for the seg1 binding and bound simulations, 9 sodium ions for the ArkA binding simulations and 3 chloride ions for the unbound simulations.

All systems were subject to energy minimization (1000 steps using harmonic restraints with a force constant of 10 kcal/mol, 1000 steps without restraints) where the first 500 steps were steepest descent and the second 500 steps conjugate gradient in both cases. The systems were then subject to heating from 100 to 300 K (harmonic restraints with a force constant of 10 kcal/mol), and equilibration (50 ps with harmonic restraints with a force constant of 10 kcal/mol). All constructs, except the unbound simulations, were equilibrated again for 200 ps without restraints. Independent simulations were started with new random velocities. Bonds to hydrogen were constrained using the SHAKE algorithm during all simulations. The nonbonded cutoff was set to 9 Å.

The unbound simulations were run using temperature Replica Exchange MD [74]. 48 replicas were simulated using the NVT ensemble with temperatures from 290.00 - 425.00 K with geometric spacing to achieve similar exchange probabilities for all replicas (S1 Table) [74]. Each replica was equilibrated without restraints for 500 ps. The simulations were run with an integration step every 2 fs and coordinates stored every 25 ps. Three independent simulations from an extended peptide and the conformation in the NMR structure were run for at least 125 ns each, resulting in a total of 1.15 μs of simulation at 300 K. The first 50 ns of each simulation was removed before analysis, resulting in 0.850 μs of simulation data used in the ArkA unbound ensemble.

The bound and binding simulations were run using the NPT ensemble at 300 K with a Monte Carlo barostat with new system volumes attempted every 100 steps an integration step every 2 fs, and coordinates stored every 10 ps. The number and length of all simulations are summarized below (table 1). The replica exchange simulations were run on the XSEDE resource Xstream [75], as well as a local cluster, and all other simulations were run on a local cluster.

Table 1. Summary of simulations run.

Construct	# of simulations	Length analyzed per simulation (ns)	Total length analyzed (μs)
Unbound ArkA	2	100	0.849
	1	75	
	1	200	
	1	249.3	
	1	125	
Bound ArkA	5	1832	9.16
Binding ArkA	50	1000	50
Binding seg1	50	500	25

Simulation Analysis

To analyze the trajectories, the AmberTools 16 package was used to measure dihedral angles, distances, secondary structure, solvent accessible surface area, and hydrogen bonds [72]. In house Python scripts were used for additional analysis. Error bars and standard deviations were calculated by computing independent values from each independent simulation and taking the standard deviation of those values.

Sampling Completeness. The running average of secondary structure per residue was used as a measure of completeness of sampling for the unbound simulations. The autocorrelation time between replicas was also calculated to ensure the replicas were exchanging as expected [76].

Structural analysis of ensemble. Dihedral RMSD of ArkA from the NMR structure was calculated as described by Kreiger *et al*. [77]. The distance between the binding surface of AbpSH3 and ArkA was defined by the average of seven pairwise ArkA-AbpSH3 contact distances between α carbons where the SH3 domain residue changes chemical shift upon ArkA binding (S1 Fig) [38]. ArkA seg1 lacks three of these pairs, so the remaining four were used, in both cases this is called the average pairwise distance. The dihedral angles were used to calculate the polyproline II helix (PPII) content as described by Masiaux *et al*. [78]. Residue distances were calculated based on the center of mass for each residue in ArkA and AbpSH3, and 8 Å was used as the cut off distance to define a contact. Contact maps were created based on the percentage of the simulation during which residue contacts were made. Contact maps that describe a subset of the simulated ensemble (encounter, forward, reverse, seg2 only, encounter other, or unbound) were created based on the percentage of that subset that is making a contact.

The data from the binding simulations were divided into unbound, encounter complex, and fully engaged based on the average pairwise distance (Table 2). The definition of the fully engaged

complex was chosen to be based on a pairwise distance less than 11.5 Å because the bound simulations had a pairwise distance less than 11.5 Å in 98% of the simulated ensemble. We chose the encounter complex distance to be twice the fully engaged distance. Within the encounter complex ensemble, four categories were defined based on contacts they make (Table 2).

Table 2. Categories in binding simulations.

Binding Interaction	Contacts
Unbound	Average pairwise distance > 23 Å
Fully engaged	Average pairwise distance < 11.5 Å
Encounter Complex	11.5 Å < average pairwise distance < 23 Å
Encounter forward	K(3) to AbpSH3 8, 9, or 10 and K(-3) to AbpSH3 33, 35, or 36
Encounter reverse	K(-3) to AbpSH3 8, 9, or 10 and K(3) to AbpSH3 33, 35, or 36
Encounter seg2 only	K(-3) to AbpSH3 33, 35, or 36 and K(-8) to AbpSH3 14, 15,16,17 or 49 or L(-7) to AbpSH3 32, 33, 36, 49 and not Forward or Reverse
Encounter other	All other structures

The ArkA simulated ensembles were divided into natively folded conformations and nonnative conformations based on a histogram of the ArkA dihedral RMSD from the unbound

simulations. The first minimum in the histogram, at 33.7° was used as a cutoff (S10 Fig). All ArkA conformations with a dihedral RMSD less than 33.7° are considered to have a native conformation, and those with a dihedral RMSD greater than 33.7° are considered to have a nonnative conformation. For the seg1 dihedral RMSD, a cutoff of 38.1° was used to define the native and nonnative conformations (S9 Fig).

Long-range electrostatic interactions between ArkA and AbpSH3s were analyzed by calculating the distance between the charged groups of the positively charged residues (N_ζ of Lys) on ArkA and the negatively charged residues (C_γ of Asp and C_δ of Glu) on AbpSH3. The calculation was performed on the ArkA binding simulations and bound simulations. A 10 Å cut off distance was used to define an electrostatic interaction. An in-house python script was used to calculate the percentage of time each electrostatic interaction was present and the average number of electrostatic interactions present simultaneously for each simulation. An interaction had to be present in one of the simulations for at least 10% of the ensemble to be included in the results.

SH3 domain dipole moment. The net dipole of the AbpSH3 domain was calculated using the Protein Dipole Moments Server administered by the Weizmann Institute Department of Structural Biology [79] by uploading the crystal structure of the domain (PDB: 1JO8) [67].

Calculation of k_{on} and k_1. The time constant, τ, for binding of ArkA and seg1 to AbpSH3 is related to the rate constant, k_{on} or k_1, and the concentration of protein and peptide by the equation

$$rate = k[\text{AbpSH3}][\text{ArkA}] \# (1)$$

where k is either k_{on} or k_1. The volume of the boxes varied slightly between the two ArkA constructs, so the concentrations and rates were slightly different (S1).

If the AbpSH3 concentration is held constant, then the transition from unbound to either the encounter complex or the fully engaged complex can be treated as a first-order reaction, so the binding time follows a Poisson distribution [80,81]. The binding time constant, τ, was calculated from a fit of the empirical cumulative distribution function to the theoretical cumulative distribution function (TCDF) [81],

$$TCDF = 1 - e^{\frac{-t}{\tau}}, \#(2)$$

where t is time of the simulation when binding occurs. Since we observed some overlap in the distribution of average pairwise distances for fully engaged and encounter complexes, we used a more stringent definition of binding for identifying transitions between states. To go from the encounter complex to the fully engaged complex we required that the pairwise distance be below 10.5 Å for at least 1 ns, and to go from unbound to encounter we required it to be below 21 Å for 1 ns. τ was then used to calculate the binding rate constant,

$$k = \frac{1}{[\text{AbpSH3}]\tau}. \#(3)$$

The standard deviation in calculation of k_1 was determined using the bootstrap method, but we could not determine a standard deviation for k_{on} because not all simulations reached the fully engaged state.

Our simulations were performed without salt present (aside from neutralizing ions), while the experimental rate constants were measured at 100 mM NaCl, which could affect the binding rate, particularly for the formation of the encounter complex, which is partially driven by electrostatic interactions.

To determine the number of transitions from the encounter complex, we used a similar method and required that the pairwise distance be above 25 Å for at least 1 ns for the transition from encounter complex to unbound to be counted in order to make sure that we only counted true transitions.

Experimental

The AbpSH3 protein and ArkA peptide were produced as previously [38]. A set of [15]N Carr-Purcell-Meiboom-Gill (CPMG) relaxation dispersion experiments were recorded on 1 mM ^{15}N labeled domain with ~9% bound to unlabeled peptide in 50 mM phosphate, 100 mM NaCl, pH 7.0 at 10 °C. The data was collected on up to two different static magnetic field strengths (500 and 800 MHz) generating a series of 2D [1]H-^{15}N correlation maps measured as a function of CPMG frequency. The spectra were processed using standard approaches and the program chemex [82,83] assuming a 2-state reaction using known chemical shift differences between the ground and excited states [38,84], which has been applied to a few other domains [29,36]. Exchange data was processed with previously measured K_d values [38] to generate rate constants.

Results & Discussion

ArkA disorder when unbound & bound

We first sought to characterize the structural ensemble of the unbound ArkA peptide to help in determining how the intrinsic structural propensities of ArkA contribute to binding. In order to determine the structural ensemble of the unbound ArkA peptide, we simulated the 12-residue ArkA alone using REMD [74] (unbound simulations). The completeness of sampling was examined using the running averages of 3_{10} helix, bend, and turn structure, as well as end-to-end

distance (S2 Fig). The efficiency of exchange between the replicas was confirmed by determining that the autocorrelation time for the replica temperature (time constant 7-10 ns) was well below that of our simulation time (S3 Fig). The structural ensemble of ArkA simulated alone shows that it is behaving as a disordered peptide with both the end-to-end distance and dihedral angle RMSD sampling multiple states that are different from the NMR reference structure (Fig 3A).

Fig 3. Characterization of unbound, bound, and encounter ensembles. A) Conformational ensemble of unbound ArkA from REMD simulations. End-to-end distance is the distance between the C and N-terminal ends of ArkA and dihedral angle RMSD is calculated only for ArkA with the lowest energy NMR structure (2RPN) as the reference [69]. Darker shading indicates a larger

fraction of the total ensemble, as indicated by the color bar. B) Overlay of 13 randomly selected ArkA conformations from unbound simulations with seg1 residues Pro(2) to Pro(-2) backbone aligned. C) Overlay of 15 randomly selected ArkA conformations from bound simulations with the SH3 domain aligned. D) Overlay of 38 randomly selected ArkA conformations in the encounter complex from ArkA binding simulations with the SH3 domain aligned. AbpSH3 SI is shown in red and SII in blue. ArkA is shown in in gray (seg1), magenta (Lys(-3)) and green (seg2).

We also ran simulations starting from the NMR bound structure of ArkA to compare its structure when fully engaged with AbpSH3 (bound simulations) (Fig 3C). NMR experiments have shown that in the fully engaged state ArkA seg1 adopts a PPII helix and seg2 is often in a 3_{10} helix (S5 Fig) [69]. In both the alone and bound simulations, seg1 of ArkA is largely structured with the majority of time spent in a PPII helix, while seg2 is less structured (Fig 4). Even in the bound simulations seg2 is only in a 3_{10} helix 35% of time, indicating that the majority of the diversity in ArkA secondary structure occurs in the seg2 region. Contact maps between ArkA and AbpSH3 also show that the entire peptide is more flexible in the bound simulations than the NMR structures indicates. Both segments of ArkA have fewer contacts with AbpSH3 in the simulated ensemble than the NMR ensemble, especially seg2 (S7 Fig).

Fig 4. Quantifying bound and unbound secondary structure. Fraction of time each ArkA residue is spending in PPII Helix (A) or 3_{10} Helix (B) during the bound and unbound simulations. The one letter codes for ArkA residues are included on the x-axis. The shaded region represents the standard deviation between independent simulations.

ArkA binds via an encounter complex intermediate

After determining the conformational ensemble of the unbound ArkA peptide, we wanted to characterize the binding pathway by running ArkA binding simulations. To start the ArkA binding simulations, we chose five structures with differing values of dihedral RMSD and end-to-end distance (S8 Fig) and used these to initiate 50 independent binding simulations (10 from each structure) each 1 μs long (Table 1). This ensured that the binding simulations would not be biased by a single starting ArkA conformation. We also ran binding simulations with the shorter seg1 peptide (seg1 binding simulations), which contains the PxxP motif and is relatively structured, to test examine the different roles of seg1 and seg2 in binding (S1 Text). We divided the unbound ArkA ensemble into two states based on the seg1 dihedral RMSD with a cutoff of 38.1˚ (S9 Fig).

We ran 25 500-ns independent seg1 binding simulations starting from each of the two states, for a total of 50 seg1 simulations (Table 1).

In the ArkA binding simulations, we found that an initial encounter complex forms quickly before ArkA transitions to a fully engaged state (S1 Movie and S2 Movie). As described in the methods, the fully engaged state was defined as a structure where the average of pairwise distances is below 11.5 Å, while we defined the encounter complex as having an average pairwise distance between 11.5 Å and 23 Å. In the binding simulations, ArkA passes through the encounter complex (Fig 5A-B) before reaching the fully engaged state. Interestingly, some of the independent simulations in the encounter complex unbind before rebinding (Fig 5A), while others quickly reach the same fully engaged state observed in the bound simulations (Fig 5B). In the seg1 binding simulations we similarly observed the formation of an initial encounter complex, followed by either unbinding and rebinding, or transition to a stable fully engaged bound state (S11 Fig), while in the bound simulations, the complex remains in the fully engaged state 98% of the time (Fig 5C).

Fig 5. Time traces of binding and bound simulations. Distance between ArkA and the binding surface of AbpSH3 over time for the first half of two example ArkA binding simulations (A, B)

and one of the bound simulations (C). The black lines correspond to our definition of the encounter complex (23 Å) and the fully engage complex (11.5 Å).

The ArkA encounter complex is a heterogeneous ensemble that includes nonnative interactions

To further examine the nature of the encounter complex, we projected the data onto coordinates corresponding to native folding of ArkA peptide secondary structure (dihedral RMSD) and binding (pair-wise binding surface distance) (Fig 6). In the binding simulations, ArkA samples many states with different degrees of native folding and binding before reaching the fully engaged and native folded state found in the lower left of the plot (Fig 6A, blue rectangle). In particular, the encounter complex ensemble is a highly heterogeneous state, as shown in Fig 3D, and 57% of the ArkA binding ensemble occupies the encounter complex without forming the native ArkA fold (Fig. 6A), indicating that ArkA does not need to be already preformed in the native conformation before interacting with AbpSH3, consistent with an induced fit binding mechanism. However, in 20% of the binding ensemble ArkA has a native fold but is still in the encounter complex. This indicates that at times ArkA may first adopt a native fold and then reorient and dock into the fully engaged state in a conformational selection mechanism. Additionally, Fig 6B confirms that the bound simulations stay fully engaged but do rarely (4% of the ensemble) sample nonnative conformations that are different from the NMR structure (unfolded and fully engaged in Fig 6B). We only observe two brief events, totaling to less than 2% of the ensemble, where the complex transitions to the encounter complex and back to the fully engaged state in one of the five bound simulations.

Fig 6. ArkA states sampled during binding and bound simulations. Distance between ArkA

and the binding surface of AbpSH3 graphed against the ArkA dihedral angle RMSD for ArkA

binding simulations (A) and bound simulations (B). Darker shading indicates a larger fraction of

the total ensemble, as indicated by the color bar. Colored boxes partition the ensemble into four

states: folded and fully engaged (blue), unfolded and fully engaged (green), folded and encounter

(red), unfolded and encounter (cyan). In this figure, folded refers to the native NMR structure

fold (dihedral RMSD less than 33.7°), and unfolded refers to a nonnative conformation different

from the NMR fold (dihedral RMSD greater than 33.7°). Percentages indicate the occupancy of

each state in the overall simulated ensemble. In the ArkA binding simulations, 8% of the ensemble is in the unbound state, which is not shown on the plot.

Contact maps also illustrate the heterogeneous nature of the encounter complex ensemble (Fig 7). In the encounter ensemble, ArkA makes nonnative contacts which are not seen in the bound simulations. These contacts are mainly on the two binding surfaces (SI and SII), indicating that electrostatic orientational steering guides the positively charged ArkA to interact with the correct surface of the domain, though not necessarily with native contacts. This is explained by the presence of a net dipole moment on the SH3 domain of 242 Debyes, with the negative end of the dipole located at the binding surface and the positive end on the opposite side of the domain, and has been observed for other SH3 domains [85]. In the encounter complex, seg1 forms more contacts (9 on average) with the binding surface than seg2 (6 on average).

Fig 7. Contact maps between AbpSH3 and ArkA. Contact maps of the fully engaged state from the bound simulations (A) and the encounter complex ensemble from the binding simulations (B). The darker squares indicate more of the ensemble with that contact. The red and blue squares along

the AbpSH3 residue index show which amino acids are in SI and SII, respectfully. The black line indicates the separation of seg1 and seg2, and the single letter amino acid codes are included for ArkA and the residues in SI and SII. Squares that show some occupancy in the encounter complex but are not occupied at all in the bound simulations indicate that nonnative contacts are part of the encounter complex ensemble. On average in the encounter complex, seg1 is in contact with 9 SH3 domain residues, while seg2 is in contact with 6 residues. ArkA K(-3) is in contact with 2 SH3 domain residues on average in the encounter complex, consistent with its role as an important central residue for binding.

The nonnative contacts on the binding surface that are formed in the encounter complex are consistent with ArkA binding in reverse in part of the encounter ensemble (Fig 8). In general, SH3 domains depend on the PxxP motif to bind, and as this is a pseudo-palindromic motif, reverse binding for seg1 on SI is not surprising. To further examine the conformational states in the encounter complex, we broke the encounter complex into four categories (defined in the methods, Table 2): forward, reverse, seg2 only, and other (Table 3). The reverse structures correspond to some of the states in the encounter ensemble that have a binding surface distance in Fig 6A higher than 16 Å. The encounter complex contact map (Fig 7) shows that seg1 interacts more with the binding surface overall than seg2, and Table 3 shows that then encounter complex is twice as likely to sample a state with seg1 engaged in the correct orientation on SI than a state with only seg2 engaged in the correct place on SII. Together, this indicates that seg1 likely binds before seg2. Seg2 may be needed to ensure specific forward binding since seg2 does not interact with the domain when ArkA binds in reverse (Fig 8). Both the ArkA and seg1 binding simulations show forward and reverse binding, showing that in the encounter complex the two segments behave somewhat independently. However, the ArkA encounter complex ensemble is much more complex

and heterogeneous than that of the short seg1 peptide (S12 Fig), and the seg1 peptide samples the

forward state less often than the seg1 encounter complex (Table 3), indicating that this short

peptide may not give an accurate representation of how that segment behaves as part of the longer

sequence.

Fig 8. Forward and reverse representations of ArkA during binding. Snapshots from MD

simulation showing both the forward and reverse orientations of ArkA that are possible during

binding. ArkA is shown in the stick representation with seg1 in grey, seg2 in green, and K(-3) in

magenta. AbpSH3 is shown in a surface representation with the red being SI and blue SII. The

double headed arrow signifies that the ArkA orientation can flip during a simulation.

Table 3. Percentage of the encounter complex ensemble in each category for ArkA and seg1 simulations.

	Forward	Reverse	Seg2 only	Other
Encounter ArkA	9.6	8.7	3.9	77.7
Encounter seg1	24.5	18.7	NA	56.9

We found that generally when individual simulations begin interacting in either the forward or reverse encounter state they do not change orientation; however, sometimes ArkA spins around and shifts over to go from forward to reverse (2 times out of 31) or reverse to forward (4 times out of 15) (S3 Movie). This shows that the encounter complex is in dynamic exchange between different predominantly nonnative conformations and contacts, including the forward and reverse orientation of the peptide (Fig 3D).

The nonspecific, disordered encounter complex that we characterize from our simulations is similar to encounter complexes seen previously in MD simulation studies of the proline-rich Sos peptide binding to the c-Crk N-SH3 domain [21,23,54]. Ahmad *et al.* found it their simulations that the initial Sos-SH3 domain complex formed very rapidly, and they identified three binding modes, including forward and reverse orientations of the PPII helix on the binding surface [21]. Because these binding modes formed very quickly, they likely represent different conformations within the encounter complex ensemble, rather than fully engaged states of the complex. Ahmad *et al.* also identified a binding mode where the Sos peptide interacts with the c-Crk N-SH3 domain

at a new binding surface [21]; however, our simulations do not show any evidence that the ArkA peptide interacts significantly with a surface of the AbpSH3 domain other than the canonical binding surface. The nonnative interactions of ArkA with the AbpSH3 binding surface we observe in the encounter complex ensemble are similar to the alternate states observed by Yuwen *et al.* in simulations characterizing a mutant c-Crk N-SH3 domain interacting with the Sos peptide (designed to imitate the encounter complex) [54]. A diverse encounter complex ensemble that includes nonnative interactions may be characteristic of the binding between proline-rich peptides and SH3 domains.

Electrostatic interactions stabilize the encounter complex

Because of the complimentary charges of ArkA and AbpSH3 and previous studies that focused on electrostatic interactions, we chose to particularly examine the role that long-range electrostatic interactions play in the encounter complex. We measured the intermolecular electrostatic contacts present in the encounter complex ensemble and in the fully engaged ArkA-AbpSH3 complex. While the long-range electrostatic contacts present in the encounter complex ensemble are more diverse (nonspecific) than those in the bound simulations (Fig 9), the average number of electrostatic contacts in the encounter complex ensemble at any given time is very similar to the average number in the bound simulations (Table 4). Thus, the main favorable contribution of the positively charged ArkA peptide interacting with the negatively charged AbpSH3 binding surface is gained upon formation of the encounter complex rather than upon transitioning from the encounter to the fully engaged state.

Fig 9. Long-range electrostatic interactions are non-specific in the encounter complex and specific in the bound simulations. Frequency of particular long-range electrostatic interactions in the ArkA encounter complex (blue bars on the left) and ArkA bound simulations (orange bars on the right). The large labels indicate the ArkA residue involved in the long-range electrostatic interaction and the small labels indicate the AbpSH3 residue. Error bars represent the standard deviation between independent simulations.

Table 4. Average number of long-range electrostatic contacts for different states.

	Bound simulations	ArkA binding simulations		
		Fully engaged	Encounter	Unbound
Average number of ArkA-AbpSH3 long-range electrostatic contacts	5.7 ± 0.3	6 ± 1	5 ± 1	2.0 ± 0.8

Previous studies of SH3 binding have also found that electrostatic interactions are important for the formation of the complex [21,23,85]. MD simulations of the Sos peptide binding to c-Crk N-SH3 were able to specifically identify electrostatic interactions that stabilize that encounter complex, including nonnative contacts [21,23]. Experimental studies of the viral NS1

peptide binding to the CrkII N-SH3 domain indicate that electrostatic contacts are important for specific binding, and that flexibility in the fully engaged state allows increased electrostatic stabilization as multiple interactions form as part of the ensemble of bound states [66,85]. Our simulations indicate that a diversity of different electrostatic contacts, each present in only part of the ensemble, is even more characteristic of the ArkA-AbpSH3 encounter complex than the fully engaged complex. The heterogeneity, or 'fuzziness', of the encounter complex ensemble is important, as there can be multiple pathways from this fuzzy encounter state to the fully engaged complex [35]. Electrostatic interactions can enhance this effect, not only stabilizing the encounter complex, but also lowering the free energy barrier to transition between basins and transition to the fully engaged state [13,35]. As many of the ArkA-AbpSH3 encounter complex electrostatic contacts do not form at all in the fully engaged ensemble, it is also important that none are strong enough interactions to trap the complex in a conformation incompatible with transitioning to the fully engaged state.

Hydrophobic interactions are nonspecific in the encounter complex and specific in the fully engaged complex

Since simulations indicate that long-range electrostatic interactions are already formed in the ArkA-AbpSH3 encounter complex, we sought to identify what energetically favorable changes occur upon transitioning from the encounter complex to the fully engaged state. Measurements of solvent accessible surface area of the complex indicate that in the encounter complex part of the SH3 domain binding surface is buried (Fig 10A) because ArkA forms transient nonspecific hydrophobic interactions with the binding surface. However, in transitioning to the fully engaged complex, the ArkA PPII helix packs into the grooves in SI, and native contacts that are largely absent in the encounter ensemble form between hydrophobic sidechains at the interface (Fig 10C).

This buries more of the binding surface, which transitions from ~50 to ~45 to ~40 nm² solvent exposed surface area as the complex transitions from unbound to the encounter complex to the fully engaged complex (Fig 10A).

Fig 10. Solvent accessibility and specific intermolecular interactions for ArkA-AbpSH3 complex . A) Average solvent accessible surface area (SASA) of the ArkA-AbpSH3 system in the bound simulations (first bar) and binding simulations by state of the complex. B) Occupancy of the P(2) to Y54 hydrogen bond in the bound simulations and ArkA binding simulations by state

of the complex. C) Specific hydrophobic contacts between ArkA and the AbpSH3 binding surface in the fully engaged and encounter complexes. Hydrophobic contacts were selected from those hydrocarbon groups that are closest together in the NMR structural ensemble (2RPN) [69]. Contacts were defined based on a 6 Å cutoff distance between hydrocarbon groups. Error bars represent the standard deviation between independent simulations.

Additionally, in transitioning from the encounter complex to the fully engaged complex, one to two additional short-range hydrogen bond interactions appear (S14 Fig). In particular, in the bound simulations, there is one hydrogen bond, from the ArkA P(2) carbonyl oxygen to the AbpSH3 Y54 side chain hydroxyl group, that is present more than any others, in 88% of the simulated ensemble (Fig 10B). This hydrogen bond is rarely formed in the encounter ensemble, indicating that it may also to help to stabilize the fully engaged state and prevent unbinding. Previous mutation studies have found that mutating P(2) causes a large reduction in binding affinity of ArkA [69], possibly in part due to disruption of the hydrogen bond to Y54. The other mutations that caused the biggest reduction in binding affinity are K(-3), and L(-7) [69], which both contain hydrophobic groups that are buried as the fully engaged complex forms in our simulations (Fig 10C), supporting the central role of these specific hydrophobic interactions . While the encounter complex is characterized by nonspecific electrostatic and hydrophobic interactions, the fully engaged complex requires more specific and complete hydrophobic contacts between ArkA and the AbpSH3 binding surface and is additionally geometrically constrained by the formation of a specific hydrogen bond.

ArkA-AbpSH3 two-step binding model

Putting together all of our data from the binding simulations, we can form a picture of how ArkA binding to AbpSH3 proceeds (Fig 2). Initially the ArkA peptide is (orientationally) steered by long-range electrostatic attraction to the AbpSH3 binding surface and forms a metastable encounter complex (step 1). This encounter complex is stabilized by transient and nonnative interactions, including long-range electrostatic interactions and partially engaged hydrophobic contacts, but the binding surface is still partially solvated, especially SII. Even before formation of the encounter complex, seg1 of ArkA is pre-folded into a PPII helix, and in the encounter complex it often forms nonspecific hydrophobic interactions with SI and nonnative hydrogen bonds, although part of the peptide is still solvated in any given conformation, and the P(2) to Y54 hydrogen bond is essentially absent. The seg1 PPII helix can interact with SI of AbpSH3 in either the forward or reverse orientation in the encounter complex, but the reverse orientation requires that seg2 interact with solvent rather than SII of AbpSH3. From the forward state of the encounter complex, ArkA-AbpSH3 can transition to the fully engaged state through a zippering process [5], burying hydrophobic sidechains and displacing more solvent, particularly on seg2 and SII, and forming the P(2) to Y54 hydrogen bond (step 2). This transition also coincides with seg2 of ArkA becoming a bit more structured, although it is clear that the fully engaged state is still in dynamic exchange, consistent with previous co-liner chemical shift perturbation measurements [38]. This binding pathway (Fig 2) is similar to that observed in MD simulations of binding to the c-Crk N-SH3 domain [21,23], indicating that this may be a common binding progression for proline-rich peptides binding to SH3 domains.

ArkA intrinsic structure affects the binding pathway

Even in the fuzzy encounter complex ensemble, seg1 of ArkA is largely folded into a PPII helix, exhibiting a binding strategy also employed by other IDPs of one segment with pre-formed

structure that can dock into place first, followed by the coupled folding and binding of more flexible segments [5,7,11,25,26,46]. Polyproline sequences are especially well adapted to this strategy as PPII helices are rigid, allowing them to project from folded parts of the full length protein and hydrophobic yet also highly soluble in water [5]. This involvement of pre-formed structure in the binding pathway is useful for modulating the entropy change on binding by tuning the degree of structure present in the unbound state [5,24]. Sequence changes that change the PPII propensity but maintain the same fully-engaged SH3 complex could be a mechanism to tune the association rate, affinity, and specificity of the interaction for different cellular functions. Seg2 of ArkA is more flexible and therefore less likely to form the first tight interactions with the AbpSH3 binding surface. Modulating the amount of intrinsic 3_{10} helix structure in seg2 would be unlikely to affect the peptide binding affinity [25], since this segment also remains quite flexible in the fully engaged complex.

Binding rates probed by NMR and MD simulations

Using NMR CPMG experiments, we determined that the ArkA peptide binds quickly, on the μs timescale at our experimental concentrations (Table 5) [38]. In the binding simulations, ArkA generally reached a stable state in the 1 μs of simulation time, but often this state was part of the metastable encounter complex rather than the fully engaged state. Based on the 9 simulations out of 50 that did reach the fully engaged state, we calculated a binding rate constant, k_{on}, which we compare to the experimental binding rate (Table 5). Our simulations show that binding happens on a similar timescale to the binding rates measured by NMR. However, the k_{on} value from our simulations is imprecise because most simulations remained in the encounter complex (panel A in S16 Fig), and we are only able to definitively state that our simulations are not inconsistent with

the rate constants determined by NMR. We can more precisely calculate a rate constant, k_1, for step 1 of binding (Fig 2), since all simulations reached the encounter complex (panel B in S16 Fig). Step 1 occurs about two orders of magnitude more rapidly than the complete binding process. This extremely rapid k_1 indicates that k_2 could be quite slow and still result in the fast k_{on} observed experimentally.

Table 5. Binding rate constants for ArkA and seg1 determined from binding simulations and NMR experiments.

Construct	Simulations used	k_{on}		k_1
		Simulations (s^{-1} M^{-1})	NMR (s^{-1} M^{-1})	Simulations (s^{-1} M^{-1})
ArkA	50×1 μs	8×10^7 *	3.00×10^7	$5.9 \pm 0.8 \times 10^9$
seg1	50×500 ns	5×10^7 *	Not performed	$1.1 \pm 0.2 \times 10^{10}$

* The precision of the values of k_{on} based on simulations cannot be estimated from the small number of times binding in the fully engaged state was observed.

Although our simulations clearly show a two-step binding process for ArkA-AbpSH3, previous kinetics experiments on SH3 domains have shown rapid binding consistent with a diffusion-limited process that occurs in a single step, enhanced by electrostatic steering [41]. The Fyn SH3 domain binds its peptide partner with a k_{on} of 1.03×10^8 s^{-1} M^{-1} [36], c-Crk N-SH3 binds Sos with a k_{on} of 2×10^9 s^{-1} M^{-1} [23], and the CrkII N-SH3 binds JNK1 with a k_{on} of 1.06×10^8 s^{-1} M^{-1} and NS1 with a k_{on} of 1.1×10^{10} s^{-1} M^{-1} [85] (all with salt concentrations similar to or slightly

lower than our experiments). However, SH3 binding peptides are IDPs, which makes the theoretical diffusion limited rate more difficult to calculate than for ordered proteins, and some IDPs seem to exceed the upper limit for binding [41,86]. One study of a disordered region of PUMA binding to Mcl-1 found that an association rate that at first seemed to be diffusion limited in fact showed a temperature dependence for k_{on}, indicating an energy barrier in the association process, and therefore two-step binding [41]. In our simulations, the diffusion limited association rate with an electrostatic enhancement is captured by k_1; however, the overall association rate, k_{on}, also depends on step 2 in our binding model (Fig 2).

In our simulations, step 1 (formation of the encounter complex) happens about two orders of magnitude more rapidly than the overall binding process, indicating that there is a larger free energy barrier going from the encounter complex to the fully engaged state than from the unbound ensemble to the encounter complex. Although the step 1 rate depends on the concentration of peptide and SH3 domain, the number of transitions out of the encounter complex to either the unbound (reverse of step 1) or fully engaged state (step 2) is independent of concentration, and therefore easier to compare directly, so we focused on comparing these transitions. In our 50 independent simulations of ArkA binding, we observe 51 transitions from the encounter complex to the unbound state and only 9 from the encounter complex to the fully engaged state. This indicates that the encounter complex is approximately 5 times more likely to transition to the unbound state, and the barrier between the encounter complex and the fully engaged state is about 4 kJ/mol higher than the barrier between the encounter complex and the unbound state. This is a relatively small difference in the free energies of the two transition states, indicating that step 2 of binding still proceeds fairly rapidly after encounter complex formation, which is consistent with the high k_{on} value measured by NMR.

Our simulation result showing that the rate limiting step for ArkA binding to AbpSH3 occurs after the encounter complex formation contrasts with previous experimental data that indicate a two-state binding process [38]. Typically, when a single association rate is observed, it indicates that step 2 is very fast compared to step 1 [7], but this does not appear to be the case for SH3 domain binding based on this and other simulation studies where the encounter complex forms more quickly than the fully engaged state [23,87]. Other MD simulation studies of IDP binding have also revealed encounter complexes that form quickly, followed by a slower transition to the fully engaged complex [48,52,56]. One alternative explanation for the apparent two-state binding is that the encounter complex is only present at a very low population (< 0.5%), and therefore not detectable by NMR [36].

The role of the encounter complex and hydrophobic interactions in binding kinetics and function

Our two-step binding model (Fig 2) that includes a fuzzy encounter complex stabilized by nonspecific hydrophobic and electrostatic interactions followed by formation of native contacts in the fully-engaged complex is observed in simulations of other IDP binding proteins, including other SH3 domains [21,23], the PDZ domain [52], self-binding proteins [55], and the TAZ1 domain [48]. IDP complexes that lack strong charge complementarity, such as pKID and KIX, are similar, but rely mainly on nonspecific hydrophobic interactions to stabilize the encounter complex [46]. The presence of a metastable encounter complex intermediate in the binding pathway allows nonnative interactions to play an important role in the binding process. The nonspecific encounter complex can form quickly, and nonnative, transient interactions (both electrostatic and hydrophobic) allow the encounter complex to remain flexible and avoid being trapped in a state that is off-pathway for complete binding. In this binding model, hydrophobic interactions are

critical in both the encounter and fully engaged complexes. While long-range electrostatic interactions only form during step 1 of binding and then remain essentially constant, hydrophobic interactions are critical to steps 1 and 2, forming nonspecifically in the transition to the encounter complex and specifically, to bury more surface area, when transitioning the fully engaged complex. If binding only occurred in one step, mutations that affect hydrophobic interactions would only affect the dissociation rate, and not association rate of the peptide. However, with our two step binding model, we predict that hydrophobic interactions affect the stability of both the encounter complex and fully engaged state, and therefore play a role in determining the overall association rate.

The central role of hydrophobic interactions in SH3 binding was also observed by Meneses and Mittermaier [29]. They find that electrostatic rate enhancement of binding to the Fyn SH3 domain is minimal since long-range electrostatic interactions do not significantly increase the association rate compared to hydrophobic interactions. In our model (Fig 2), hydrophobic interactions form during both reaction steps and could have large effects on the association rate as well as the dissociation rate. This is consistent with the differences in CrkII N-terminal SH3 binding by the virus protein NS1 and the endogenous binding partner JNK1 observed by Shen *et al*. [85]. While the increased binding affinity and higher association rate of NS1 has been attributed to its higher positive charge [85], NS1 also contains more hydrophobic residues than JNK1, particularly within the PxxPx+ motif, which likely also has an effect on association since hydrophobic interactions enhance the formation of the encounter complex and fully engaged complex.

The encounter complex could play an important functional role in SH3 binding in the cellular context. Competition between binding partners may need to be tuned by modulating the

encounter complex to determine which interaction will be dominant, as in the case of CITED2 competing with HIF-1α to bind TAZ1 [88]. CITED2 forms an encounter complex with TAZ1 while HIF-1α is bound, which allows it to completely displace HIF-1α even though both partners have similar affinities to TAZ1. There is evidence that AbpSH3 can form an intramolecular interaction with a proline-rich sequence of the Abp1p protein, which may inhibit binding of other partners [89]. The formation of an encounter complex in this case, may allow the competition between binding partners to be tuned as necessary for function, instead of being dictated only by their relative affinities. The transient interactions of the SH3 encounter complex may also be critical when SH3 domains need to locate a small proline rich binding motif within a longer disordered protein segment. The domain could rapidly associate with (positively charged) disordered sequences, while the interaction remains fuzzy enough to allow rearrangement or even translation of the SH3 domain along the disordered peptide sequence until the specific PPII partner segment can align with the binding surface and lock into the fully engaged state. This search process may resemble a transcription factor searching an elongated DNA strand for its promotor site. Thus, the encounter complex is likely an important functional intermediate in the binding pathway of this ubiquitous protein interaction domain.

Conclusions

Through examination of ArkA-AbpSH3 binding by MD simulations and NMR experiments, we have created a two-step binding model that includes the formation of a heterogeneous encounter complex stabilized by transient, nonspecific hydrophobic and electrostatic interactions. ArkA samples many states with different degrees of folding and binding before reaching the fully engaged state, though contacts with the domain are limited to the

canonical highly acidic binding surface. The fuzziness of the encounter complex ensemble allows multiple paths, driven by specific hydrophobic interactions and a key hydrogen bond rather than electrostatics, to the fully engaged state. The PxxP motif in seg1 is preformed in a PPII helix, which locks into the hydrophobic grooves of SI in a zippering mechanism during step 2 of binding. The more disordered seg2 could prevent the peptide from binding in reverse as a result of the pseudo-palindromic PxxP motif in seg1. While encounter complex formation is diffusion limited and enhanced by electrostatic orientational steering, the transition to the fully engaged state is slower and relies on specific hydrophobic interactions and a key hydrogen bond. In the cellular context, rapid formation of the encounter complex stabilized by transient, nonspecific interactions could allow SH3 domains to search for proline rich motifs within disordered sequences. Our binding model gives insights into the mechanism that SH3 domains use to perform a wide variety of functions. Future studies of AbpSH3 thermodynamics and kinetics would further elucidate the different roles for electrostatic and hydrophobic interactions in the binding pathway.

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References

1.   Dyson HJ, Wright PE. Coupling of folding and binding for unstructured proteins. Current Opinion in Structural Biology. Elsevier Current Trends; 2002;12: 54–60.

2.   Uversky VN. Functional roles of transiently and intrinsically disordered regions within proteins. FEBS Journal. 2015;282: 1182–1189. doi:10.1111/febs.13202

3.   Dyson HJ. Making Sense of Intrinsically Disordered Proteins. Biophys J. Elsevier; 2016;110: 1013–1016. doi:10.1016/j.bpj.2016.01.030

4.   Liu J, Faeder JR, Camacho CJ. Toward a quantitative theory of intrinsically disordered proteins and their function. Proc Natl Acad Sci USA. 2009;106: 19819–19823. doi:10.1073/pnas.0907710106

5.   Williamson M. How Proteins Work. Nature communications. Garland Science; 2012. doi:10.1038/s41467-019-10172-6

6.   Berlow RB, Dyson HJ, Wright PE. Functional advantages of dynamic protein disorder. FEBS Letters. 2015;589: 2433–2440. doi:10.1016/j.febslet.2015.06.003

7.   Shammas SL, Crabtree MD, Dahal L, Wicky BIM, Clarke J. Insights into Coupled Folding and Binding Mechanisms from Kinetic Studies. Journal of Biological Chemistry. American Society for Biochemistry and Molecular Biology; 2016;291: 6689–6695. doi:10.1074/jbc.R115.692715

8.   Wright PE, DYSON H. Intrinsically unstructured proteins: re-assessing the protein structure-function paradigm. J Mol Biol. 1999;293: 321–331. doi:10.1006/jmbi.1999.3110

9.   Brown T, Brown N, Stollar EJ. Most yeast SH3 domains bind peptide targets with high intrinsic specificity. Helmer-Citterich M, editor. PLoS ONE. Public Library of Science; 2018;13: e0193128. doi:10.1371/journal.pone.0193128

10.   Ivarsson Y, Jemth P. Affinity and specificity of motif-based protein–protein

interactions. Current Opinion in Structural Biology. Elsevier Current Trends; 2019;54: 26–33. doi:10.1016/j.sbi.2018.09.009

11. Zhou H-X, Pang X, Lu C. Rate constants and mechanisms of intrinsically disordered proteins binding to structured targets. Phys Chem Chem Phys. The Royal Society of Chemistry; 2012;14: 10466–10476. doi:10.1039/C2CP41196B

12. Kiefhaber T, Bachmann A, Jensen KS. Dynamics and mechanisms of coupled protein folding and binding reactions. Current Opinion in Structural Biology. Elsevier Current Trends; 2012;22: 21–29. doi:10.1016/j.sbi.2011.09.010

13. Ganguly D, Zhang W, Chen J. Electrostatically accelerated encounter and folding for facile recognition of intrinsically disordered proteins. de Groot BL, editor. PLoS Comput Biol. Public Library of Science; 2013;9: e1003363. doi:10.1371/journal.pcbi.1003363

14. Gianni S, Dogan J, Jemth P. Coupled binding and folding of intrinsically disordered proteins: what can we learn from kinetics? Current Opinion in Structural Biology. Elsevier Current Trends; 2016;36: 18–24. doi:10.1016/j.sbi.2015.11.012

15. Bonetti D, Troilo F, Brunori M, Longhi S, Gianni S. How Robust Is the Mechanism of Folding-Upon-Binding for an Intrinsically Disordered Protein? Biophys J. Cell Press; 2018;114: 1889–1894. doi:10.1016/j.bpj.2018.03.017

16. Lacy ER, Filippov I, Lewis WS, Otieno S, Xiao L, Weiss S, et al. p27 binds cyclin–CDK complexes through a sequential mechanism involving binding-induced protein folding. Nat Struct Mol Biol. Nature Publishing Group; 2004;11: 358–364.

doi:10.1038/nsmb746

17.   Sugase K, Dyson HJ, Wright PE. Mechanism of coupled folding and binding of an intrinsically disordered protein. Nature. 2007;447: 1021–1025. doi:10.1038/nature05858

18.   Thayer KM, Lakhani B, Beveridge DL. Molecular Dynamics-Markov State Model of Protein Ligand Binding and Allostery in CRIB-PDZ: Conformational Selection and Induced Fit. J Phys Chem B. American Chemical Society; 2017;121: 5509–5514. doi:10.1021/acs.jpcb.7b02083

19.   Dahal L, Kwan TOC, Shammas SL, Clarke J. pKID Binds to KIX via an Unstructured Transition State with Nonnative Interactions. Biophys J. 2017;113: 2713–2722. doi:10.1016/j.bpj.2017.10.016

20.   Shoemaker BA, Portman JJ, Wolynes PG. Speeding molecular recognition by using the folding funnel: The fly-casting mechanism. Proc Natl Acad Sci USA. National Academy of Sciences; 2000;97: 8868–8873. doi:10.1073/pnas.160259697

21.   Ahmad M, Gu W, Helms V. Mechanism of fast peptide recognition by SH3 domains. Angew Chem Int Ed Engl. Wiley-Blackwell; 2008;47: 7626–7630. doi:10.1002/anie.200801856

22.   Ubbink M. The courtship of proteins: Understanding the encounter complex. FEBS Letters. Wiley-Blackwell; 2009;583: 1060–1066. doi:10.1016/j.febslet.2009.02.046

23.   Xue Y, Yuwen T, Zhu F, Skrynnikov NR. Role of electrostatic interactions in

binding of peptides and intrinsically disordered proteins to their folded targets. 1. NMR and MD characterization of the complex between the c-Crk N-SH3 domain and the peptide Sos. Biochemistry. American Chemical Society; 2014;53: 6473–6495. doi:10.1021/bi500904f

24.   Csermely P, Palotai R, Nussinov R. Induced fit, conformational selection and independent dynamic segments: an extended view of binding events. Trends Biochem Sci. Elsevier Current Trends; 2010;35: 539–546. doi:10.1016/j.tibs.2010.04.009

25.   Poosapati A, Gregory E, Borcherds WM, Chemes LB, Daughdrill GW. Uncoupling the Folding and Binding of an Intrinsically Disordered Protein. J Mol Biol. 2018;430: 2389–2402. doi:10.1016/j.jmb.2018.05.045

26.   Schneider R, Blackledge M, Jensen MR. Elucidating binding mechanisms and dynamics of intrinsically disordered protein complexes using NMR spectroscopy. Current Opinion in Structural Biology. Elsevier Current Trends; 2019;54: 10–18. doi:10.1016/j.sbi.2018.09.007

27.   Wright PE, Dyson HJ. Linking folding and binding. Current Opinion in Structural Biology. 2009;19: 31–38. doi:10.1016/j.sbi.2008.12.003

28.   Uversky VN. Intrinsically Disordered Proteins and Their "Mysterious" (Meta)Physics. Front Phys. Frontiers; 2019;7: 2985. doi:10.3389/fphy.2019.00010

29.   Meneses E, Mittermaier A. Electrostatic Interactions in the Binding Pathway of a Transient Protein Complex Studied by NMR and Isothermal Titration Calorimetry. J

Biol Chem. 2014;289: 27911–27923. doi:10.1074/jbc.M114.553354

30.   Dahal L, Shammas SL, Clarke J. Phosphorylation of the IDP KID Modulates

Affinity for KIX by Increasing the Lifetime of the Complex. Biophys J. 2017;113:

2706–2712. doi:10.1016/j.bpj.2017.10.015

31.   Dogan J, Schmidt T, Mu X, Engström Å, Jemth P. Fast association and slow

transitions in the interaction between two intrinsically disordered protein domains.

Journal of Biological Chemistry. American Society for Biochemistry and Molecular

Biology; 2012;287: 34316–34324. doi:10.1074/jbc.M112.399436

32.   Dogan J, Mu X, Engström Å, Jemth P. The transition state structure for coupled

binding and folding of disordered protein domains. Sci Rep. Nature Publishing

Group; 2013;3: 1–6. doi:10.1038/srep02076

33.   Chemes LB, Sánchez IE, de Prat-Gay G. Kinetic Recognition of the Retinoblastoma

Tumor Suppressor by a Specific Protein Target. J Mol Biol. Academic Press;

2011;412: 267–284. doi:10.1016/j.jmb.2011.07.015

34.   Dosnon M, Bonetti D, Morrone A, Erales J, di Silvio E, Longhi S, et al.

Demonstration of a Folding after Binding Mechanism in the Recognition between

the Measles Virus N TAILand X Domains. ACS Chem Biol. 2015;10: 795–802.

doi:10.1021/cb5008579

35.   Ganguly D, Otieno S, Waddell B, Iconaru L, Kriwacki RW, Chen J. Electrostatically

accelerated coupled binding and folding of intrinsically disordered proteins. J Mol

Biol. 2012;422: 674–684. doi:10.1016/j.jmb.2012.06.019

36.   Demers J-P, Mittermaier A. Binding Mechanism of an SH3 Domain Studied by NMR and ITC. J Am Chem Soc. American Chemical Society; 2009;131: 4355–4367. doi:10.1021/ja808255d

37.   Horness RE, Basom EJ, Mayer JP, Thielges MC. Resolution of Site-Specific Conformational Heterogeneity in Proline-Rich Molecular Recognition by Src Homology 3 Domains. J Am Chem Soc. 2016;138: 1130–1133. doi:10.1021/jacs.5b11999

38.   Stollar EJ, Lin H, Davidson AR, Forman-Kay JD. Differential Dynamic Engagement within 24 SH3 Domain: Peptide Complexes Revealed by Co-Linear Chemical Shift Perturbation Analysis. Campos-Olivas R, editor. PLoS ONE. Public Library of Science; 2012;7: e51282. doi:10.1371/journal.pone.0051282

39.   Crabtree MD, Mendonça CATF, Bubb QR, Clarke J. Folding and binding pathways of BH3-only proteins are encoded within their intrinsically disordered sequence, not templated by partner proteins. Journal of Biological Chemistry. American Society for Biochemistry and Molecular Biology; 2018;293: 9718–9723. doi:10.1074/jbc.RA118.002791

40.   Gianni S, Morrone A, Giri R, Brunori M. A folding-after-binding mechanism describes the recognition between the transactivation domain of c-Myb and the KIX domain of the CREB-binding protein. Biochemical and Biophysical Research Communications. Academic Press; 2012;428: 205–209. doi:10.1016/j.bbrc.2012.09.112

41. Rogers JM, Steward A, Clarke J. Folding and Binding of an Intrinsically Disordered Protein: Fast, but Not "Diffusion-Limited." J Am Chem Soc. American Chemical Society; 2013;135: 1415–1422. doi:10.1021/ja309527h

42. Turjanski AG, Gutkind JS, Best RB, Hummer G. Binding-Induced Folding of a Natively Unstructured Transcription Factor. Shakhnovich EI, editor. PLoS Comput Biol. 2008;4: e1000060. doi:10.1371/journal.pcbi.1000060

43. Huang Y, Liu Z. Nonnative Interactions in Coupled Folding and Binding Processes of Intrinsically Disordered Proteins. Uversky VN, editor. PLoS ONE. Public Library of Science; 2010;5: e15375. doi:10.1371/journal.pone.0015375

44. Zhang W, Ganguly D, Chen J. Residual structures, conformational fluctuations, and electrostatic interactions in the synergistic folding of two intrinsically disordered proteins. PLoS Comput Biol. 2012;8: e1002353. doi:10.1371/journal.pcbi.1002353

45. Kumar S, Showalter SA, Noid WG. Native-Based Simulations of the Binding Interaction Between RAP74 and the Disordered FCP1 Peptide. J Phys Chem B. 2013;117: 3074–3085. doi:10.1021/jp310293b

46. Kurcinski M, Kolinski A, Kmiecik S. Mechanism of Folding and Binding of an Intrinsically Disordered Protein As Revealed by ab Initio Simulations. Journal of chemical theory and computation. American Chemical Society; 2014;10: 2224–2231. doi:10.1021/ct500287c

47. Knott M, Best RB. Discriminating binding mechanisms of an intrinsically disordered protein via a multi-state coarse-grained model. J Chem Phys. AIP

Publishing; 2014;140: 175102. doi:10.1063/1.4873710

48.   Gao M, Yang J, Liu S, Su Z, Huang Y. Intrinsically Disordered Transactivation

        Domains Bind to TAZ1 Domain of CBP via Diverse Mechanisms. Biophys J.

        2019;117: 1301–1310. doi:10.1016/j.bpj.2019.08.026

49.   Baker CM, Best RB. Insights into the binding of intrinsically disordered proteins

        from molecular dynamics simulation. Wiley Interdisciplinary Reviews:

        Computational Molecular Science. 2014;4: 182–198. doi:10.1002/wcms.1167

50.   Higo J, Nishimura Y, Nakamura H. A free-energy landscape for coupled folding and

        binding of an intrinsically disordered protein in explicit solvent from detailed all-

        atom computations. J Am Chem Soc. 2011;133: 10448–10458.

        doi:10.1021/ja110338e

51.   Hollingsworth SA, Dror RO. Molecular Dynamics Simulation for All. Neuron.

        2018;99: 1129–1143. doi:10.1016/j.neuron.2018.08.011

52.   Blöchliger N, Xu M, Caflisch A. Peptide Binding to a PDZ Domain by Electrostatic

        Steering via Nonnative Salt Bridges. Biophys J. 2015;108: 2362–2370.

        doi:10.1016/j.bpj.2015.03.038

53.   Milles S, Mercadante D, Aramburu IV, Jensen MR, Banterle N, Koehler C, et al.

        Plasticity of an Ultrafast Interaction between Nucleoporins and Nuclear Transport

        Receptors. Cell. Cell Press; 2015;163: 734–745. doi:10.1016/j.cell.2015.09.047

54.   Yuwen T, Xue Y, Skrynnikov NR. Role of Electrostatic Interactions in Binding of

Peptides and Intrinsically Disordered Proteins to Their Folded Targets: 2. The Model of Encounter Complex Involving the Double Mutant of the c-Crk N-SH3 Domain and Peptide Sos. Biochemistry. 2016;55: 1784–1800. doi:10.1021/acs.biochem.5b01283

55.   Yang C, Zhang S, Bai Z, Hou S, Wu D, Huang J, et al. A two-step binding mechanism for the self-binding peptide recognition of target domains. Mol Biosyst. The Royal Society of Chemistry; 2016;12: 1201–1213. doi:10.1039/c5mb00800j

56.   Zhou G, Pantelopulos GA, Mukherjee S, Voelz VA. Bridging Microscopic and Macroscopic Mechanisms of p53-MDM2 Binding with Kinetic Network Models. Biophys J. 2017;113: 785–793. doi:10.1016/j.bpj.2017.07.009

57.   Kaneko T, Li L, Li SS-C. The SH3 domain--a family of versatile peptide- and protein-recognition module. Front Biosci. 2008;13: 4938–4952. doi:10.2741/3053

58.   Musacchio A. How SH3 domains recognize proline. Adv Protein Chem. Academic Press; 2002;61: 211–268. doi:10.1016/S0065-3233(02)61006-X

59.   Kurochkina N, Guha U. SH3 domains: modules of protein–protein interactions. Biophys Rev. Springer-Verlag; 2012;5: 29–39. doi:10.1007/s12551-012-0081-z

60.   Zarrinpar A, Bhattacharyya RP, Lim WA. The Structure and Function of Proline Recognition Domains. Sci STKE. American Association for the Advancement of Science; 2003;2003: re8–re8. doi:10.1126/stke.2003.179.re8

61.   Lim WA, Richards FM, Fox RO. Structural determinants of peptide-binding

orientation and of sequence specificity in SH3 domains. Nature. Nature Publishing Group; 1994;372: 375–379. doi:10.1038/372375a0

62.   Feng S, Chen JK, Yu H, Simon JA, Schreiber SL. Two binding orientations for peptides to the Src SH3 domain: development of a general model for SH3-ligand interactions. Science. American Association for the Advancement of Science; 1994;266: 1241–1247. doi:10.1126/science.7526465

63.   Kishan KVR, Agrawal V. SH3-like fold proteins are structurally conserved and functionally divergent. Curr Protein Pept Sci. 2005;6: 143–150.

64.   Haynes J, Garcia B, Stollar EJ, Rath A, Andrews BJ, Davidson AR. The biologically relevant targets and binding affinity requirements for the function of the yeast actin-binding protein 1 Src-homology 3 domain vary with genetic context. Genetics. Genetics; 2007;176: 193–208. doi:10.1534/genetics.106.070300

65.   Teyra J, Huang H, Jain S, Guan X, Dong A, Liu Y, et al. Comprehensive Analysis of the Human SH3 Domain Family Reveals a Wide Variety of Non-canonical Specificities. Structure. Cell Press; 2017;25: 1598–1610.e3. doi:10.1016/j.str.2017.07.017

66.   Shen Q, Shi J, Zeng D, Zhao B, Li P, Hwang W, et al. Molecular Mechanisms of Tight Binding through Fuzzy Interactions. Biophysj. Cell Press; 2018;114: 1313–1320. doi:10.1016/j.bpj.2018.01.031

67.   Fazi B, Cope MJTV, Douangamath A, Ferracuti S, Schirwitz K, Zucconi A, et al. Unusual binding properties of the SH3 domain of the yeast actin-binding protein

Abp1: structural and functional analysis. J Biol Chem. American Society for Biochemistry and Molecular Biology; 2002;277: 5290–5298. doi:10.1074/jbc.M109848200

68.   Garcia B, Stollar EJ, Davidson AR. The importance of conserved features of yeast actin-binding protein 1 (Abp1p): the conditional nature of essentiality. Genetics. 2012;191: 1199–1211. doi:10.1534/genetics.112.141739

69.   Stollar EJ, Garcia B, Chong PA, Rath A, Lin H, Forman-Kay JD, et al. Structural, Functional, and Bioinformatic Studies Demonstrate the Crucial Role of an Extended Peptide Binding Site for the SH3 Domain of Yeast Abp1p. J Biol Chem. American Society for Biochemistry and Molecular Biology; 2009;284: 26918–26927. doi:10.1074/jbc.M109.028431

70.   Maier JA, Martinez C, Kasavajhala K, Wickstrom L, Hauser KE, Simmerling C. ff14SB: Improving the Accuracy of Protein Side Chain and Backbone Parameters from ff99SB. Journal of chemical theory and computation. American Chemical Society; 2015;11: 3696–3713. doi:10.1021/acs.jctc.5b00255

71.   Doshi U, Hamelberg D. Reoptimization of the AMBER Force Field Parameters for Peptide Bond (Omega) Torsions Using Accelerated Molecular Dynamics. J Phys Chem B. American Chemical Society; 2009;113: 16590–16595. doi:10.1021/jp907388m

72.   Case DA, Betz RM, Cerutti DS, Cheatham TE, Darden TA, Duke RE, et al. Amber 16. University of California, San Francisco; 2016.

73. Wang L-P, Martinez TJ, Pande VS. Building Force Fields: An Automatic, Systematic, and Reproducible Approach. J Phys Chem Lett. 2014;5: 1885–1891. doi:10.1021/jz500737m

74. Sugita Y, Okamoto Y. Replica-exchange molecular dynamics method for protein folding. Chem Phys Lett. North-Holland; 1999;314: 141–151. doi:10.1016/S0009-2614(99)01123-9

75. Towns J, Cockerill T, Dahan M, Foster I, Gaither K, Grimshaw A, et al. XSEDE: accelerating scientific discovery. ieeexploreieeeorg. 2014;16: 62–74. doi:10.1109/MCSE.2014.80

76. Nguyen TH, Minh DDL. Intermediate Thermodynamic States Contribute Equally to Free Energy Convergence: A Demonstration with Replica Exchange. Journal of chemical theory and computation. American Chemical Society; 2016;12: 2154–2161. doi:10.1021/acs.jctc.6b00060

77. Krieger JM, Fusco G, Lewitzky M, Simister PC, Marchant J, Camilloni C, et al. Conformational recognition of an intrinsically disordered protein. Biophys J. 2014;106: 1771–1779. doi:10.1016/j.bpj.2014.03.004

78. Mansiaux Y, Joseph AP, Gelly J-C, de Brevern AG. Assignment of PolyProline II conformation and analysis of sequence--structure relationship. Buehler M, editor. PLoS ONE. Public Library of Science; 2011;6: e18401. doi:10.1371/journal.pone.0018401

79. Felder CE, Prilusky J, acids ISN, 2007. A server and database for dipole moments of

proteins. academicoupcom.

80.   Wang Y, Martins JM, Lindorff-Larsen K. Biomolecular conformational changes and ligand binding: from kinetics to thermodynamics. Chemical Science. Royal Society of Chemistry; 2017;8: 6466–6473. doi:10.1039/C7SC01627A

81.   Salvalaglio M, Tiwary P, Parrinello M. Assessing the Reliability of the Dynamics Reconstructed from Metadynamics. Journal of chemical theory and computation. American Chemical Society; 2014;10: 1420–1425. doi:10.1021/ct500040r

82.   Tharatipyakul A, Numnark S, Wichadakul D, Ingsriswang S. ChemEx: information extraction system for chemical data curation. BMC Bioinformatics. BioMed Central; 2012;13: S9–11. doi:10.1186/1471-2105-13-S17-S9

83.   Bouvignies G, Kay LE. Measurement of Proton Chemical Shifts in Invisible States of Slowly Exchanging Protein Systems by Chemical Exchange Saturation Transfer. J Phys Chem B. American Chemical Society; 2012;116: 14311–14317. doi:10.1021/jp311109u

84.   Vallurupalli P, Bouvignies G, Kay LE. Increasing the exchange time-scale that can be probed by CPMG relaxation dispersion NMR. J Phys Chem B. 2011;115: 14891–14900. doi:10.1021/jp209610v

85.   Shen Q, Zeng D, Zhao B, Bhatt VS, Li P, Cho J-H. The Molecular Mechanisms Underlying the Hijack of Host Proteins by the 1918 Spanish Influenza Virus. ACS Chem Biol. 2017;12: 1199–1203. doi:10.1021/acschembio.7b00168

86.   Shammas SL, Travis AJ, Clarke J. Remarkably fast coupled folding and binding of

        the intrinsically disordered transactivation domain of cMyb to CBP KIX. J Phys

        Chem B. 2013;117: 13346–13356. doi:10.1021/jp404267e

87.   Ahmad M, Gu W, Helms V. Mechanism of Fast Peptide Recognition by SH3

        Domains. Angew Chem Int Ed Engl. Wiley-Blackwell; 2008;47: 7626–7630.

        doi:10.1002/anie.200801856

88.   Berlow RB, Dyson HJ, Wright PE. Hypersensitive termination of the hypoxic

        response by a disordered protein switch. Nature. Nature Publishing Group;

        2017;543: 447–451. doi:10.1038/nature21705

89.   Garcia B, Stollar EJ, Davidson AR. The Importance of Conserved Features of Yeast

        Actin-Binding Protein 1 (Abp1p): The Conditional Nature of Essentiality. Genetics.

        Genetics; 2012;191: 1199–1211. doi:10.1534/genetics.112.141739

S1 Text. Seg1 binding results.

S1 Table. Temperatures (in Kelvin) used in the unbound ArkA replica exchange simulations.

S2 Table. Summary of volume, concentration, and binding frequency for the two binding simulations.

S1 Fig. Pairwise distances used in the determination of whether the encounter simulation has reached the bound state. The pairs in red and magenta where used for both the seg1 and ArkA1 simulations and the cyan pairs where added for the ArkA simulations. The distances were

determined based on the SH3 residues whose chemical shifts where used to determine binding in NMR experiments. The pairwise distance in the NMR structures ranges from 7.25 to 7.62 Å.

S2 Fig. Running averages of measures used to determine convergence of REMD simulations. ArkA structural measures plotted vs. simulation time for each of the independent REMD simulations. The first 50 ns of each independent simulation was removed before analysis.

S3 Fig. Representative autocorrelation of replica state index graph, for one simulation, showing that the replica exchange was exchanging as expected and the number of replicas was sufficient.

S4 Fig. Overlay of the 20 ArkA conformations from the NMR (2RPN) ensemble with the SH3 domain aligned [69].

S5 Fig. Percentage of time each ArkA residue is spending in PPII Helix or 3-10 Helix in the NMR ensemble (2RPN) [69].

S6 Fig. Binding surface distance and dihedral angle RMSD for NMR structures (2RPN) [69]. The distance between ArkA and the binding surface of AbpSH3 is graphed against the dihedral angle RMSD for the bound simulations with the 20 NMR structures shown as red x's. The NMR structures all have low binding surface distances, but cover a range of different dihedral angle RMSD values. Darker shading indicates a larger fraction of the total bound simulation ensemble, as indicated by the color bar.

S7 Fig. Contact maps of the fully engaged state from the NMR ensemble (2RPN) and bound simulations. The darker squares indicate more of the ensemble with that contact. The red and blue squares along the AbpSH3 residue index show which amino acids are in SI and SII,

respectfully. The black line indicates the separation of seg1 and seg2, and the single letter amino acid codes are included for ArkA and the residues in SI and SII.

S8 Fig. Starting structures for ArkA binding simulations. The conformational ensemble of unbound ArkA from the REMD simulations is plotted in terms of end-to-end distance and dihedral angle RMSD, with starting structures for ArkA binding simulations indicated by blue circles. End-to-end distance is the distance between the C and N-terminal ends of ArkA and dihedral angle RMSD is calculated only for ArkA with the lowest energy NMR structure (2RPN) as the reference [69].

S9 Fig. Histogram of dihedral RMSD based on only the seg1 dihedral angles from the unbound simulation ensemble. The vertical line at 38.1° indicates the cutoff that was determined between the two states (native and nonnative conformations).

S10 Fig. Histogram of dihedral RMSD for the full ArkA peptide from the unbound simulation ensemble. The vertical line at 33.7° indicates the cutoff that was determined between the native folded and nonnative states.

S11 Fig. Distance between seg1 and the binding surface of AbpSH3 over time for an example seg1 binding simulation. The black lines correspond to our definition of the encounter complex (23 Å) and the fully engage complex (11.5 Å).

S12 Fig. Distance between seg1 and the binding surface of AbpSH3 graphed against the dihedral angle RMSD for seg1 binding simulations. Darker shading indicates a larger fraction of the total ensemble, as indicated by the color bar. Colored boxes partition the ensemble into

four states: folded and fully engaged (blue), unfolded and fully engaged (green), folded and encounter (red), unfolded and encounter (cyan).

S13 Fig. Solvent accessible surface area (A) and occupancy of the P(2) to Y54 hydrogen bond (B) for the seg1-AbpSH3 complex in different states. The first bar on the plot represents the solvent accessible surface area in bound simulations. Error bars represent the standard deviation between independent simulations.

S14 Fig. Average number of hydrogen bonds between AbpSH3 and the ArkA peptide for ArkA (A) and seg1 (B) in the bound simulations (first bar) and binding simulations by state of the complex. Error bars represent the standard deviation between independent simulations.

S15 Fig. Specific hydrophobic contacts between the seg1 peptide and the AbpSH3 binding surface in the fully engaged and encounter complexes. Hydrophobic contacts were selected from those hydrocarbon groups that are closest together in the NMR structural ensemble (2RPN) [69]. Contacts were defined based on a 6 Å cutoff distance between hydrocarbon groups. Error bars represent the standard deviation between independent simulations.

S16 Fig. Fit of TCDF curve (red line) to time to formation data (blue circles) for the fully engaged ArkA complex (A) and for the ArkA encounter complex (B) from ArkA binding simulations. These fits were used to determine k_{on} and k_1 respectively.

S1 Movie. Example simulation of initial ArkA binding to the AbpSH3 domain.

S2 Movie. Example simulation of initial ArkA binding to the AbpSH3 domain.

S3 Movie. Simulation of ArkA changing orientation from forward to reverse in the encounter complex.

A

B

$$ACE - \mathbf{K}_{(3)}\ \mathbf{P}_{(2)}\ \mathbf{T}_{(1)}\ \mathbf{P}_{(0)}\ \mathbf{P}_{(-1)}\ \mathbf{P}_{(-2)}\ \mathbf{K}_{(-3)}\ \mathbf{P}_{(-4)}\ \mathbf{S}_{(-5)}\ \mathbf{H}_{(-6)}\ \mathbf{L}_{(-7)}\ \mathbf{K}_{(-8)} - NHE$$

Figure

Figure

Figure

Figure

Figure

A

ArkA binding simulations

Binding Surface Distance (Å) vs Dihedral Angle RMSD

- folded & encounter
- unfolded & encounter
- folded & fully engaged
- unfolded & fully engaged

B

Bound simulations

Binding Surface Distance (Å) vs Dihedral Angle RMSD

- folded & encounter
- unfolded & encounter
- folded & fully engaged
- unfolded & fully engaged

Figure

Figure

Figure

Figure

Figure