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Structure-Guided Design of a Peptide Lock for Modular Peptide Binders

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unstructured, linear epitopes in whole proteins. Yet, due to the many different sequence possibilities even for short peptides, classical selection of binding proteins from a library, one at a time, is not scalable to proteomes. However, moving away from selection to a rational assembly of preselected modules binding to predefined linear epitopes would split the problem into smaller parts. These modules could then be reassembled in any desired order to bind to, in principle, arbitrary sequences, thereby circumventing any new rounds of selection. Designed Armadillo repeat proteins (dArmRPs) are modular, and they do bind



elongated peptides in a modular way. Their consensus sequence carries pockets that prefer arginine and lysine. In our quest to select pockets for all amino acid side chains, we had discovered that repetitive sequences can lead to register shifts and peptide flipping during selections from libraries, hindering the selection of new binding specificities. To solve this problem, we now created an orthogonal binding specificity by a combination of grafting from β -catenin, computational design and mutual optimization of the pocket and the bound peptide. We have confirmed the design and the desired interactions by X-ray structure determination. Furthermore, we could confirm the absence of sliding in solution by a single-molecule Förster resonance energy transfer. The new pocket could be moved from the N-terminus of the protein to the middle, retaining its properties, further underlining the modularity of the system.

INTRODUCTION

Protein—peptide interactions are the basis for 15–40% of all interactions in the cell, and many biochemical detection assays rely on the specific interaction of binding molecules or protein domains with unfolded proteins, loops, or peptide tags lacking a secondary structure.¹ Furthermore, 35–45% of the eukaryotic proteome is predicted to be unstructured.² Thus, linear, unstructured epitopes are of broad interest in protein detection and identification.

Nowadays, many different binding scaffolds can achieve peptide binding; even picomolar affinities have been reported for the interaction between synthetic antibodies and longer peptides.³ However, the sheer number of potential amino acid sequences, even for short peptides, would make it impossible to stockpile sequence-specific binders for every peptide by classical immunization or classical selection techniques, one binder at a time. For the detection of proteins and posttranslational modifications on proteomic scales, a rapid access to specific sequence recognition would be transformative. Yet, this problem can be overcome with a binding scaffold in which peptide binding relies solely on the primary sequence. Ideally, such a scaffold is made up of binding modules that recognize short parts of the peptide sequence. In that case, one would only need to generate specific binding pockets or modules for the 20 different amino acids and their modified versions for the detection of post-translational modifications. A reassembly of previously generated modules would then allow the generation of binders to, in principle, any given peptide target sequence.

Modular proteins already exist in nature, and the family of natural Armadillo repeat proteins (nArmRPs) binds their target peptides in an elongated conformation and already in a modular manner with, on average, two amino acids per repeat.^{4–6} Over the past years, nArmRPs served as the basis for the design of fully modular and highly stable peptide binders called designed Armadillo repeat proteins (dArmRPs;^{7–16} reviewed in refs 17 and 18). The peptide, which carries alternating arginines and lysines, is bound to specific pockets on the surface of the dArmRP, and the backbone is kept

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Figure 1. Peptide binding on dArmRPs (PDB-ID: SAEI). (A) Close-up view of a dArmRP with five internal repeats binding a $(KR)_5$ peptide, showing the interactions of the arginine and lysine side chains (pale green) to the pockets of the dArmRP (yellow-orange). Side chains are binding to surface pockets, and the backbone is kept elongated and fixed by a ladder of conserved asparagines at position 37 (N37). (B) Potential register shifts that can occur due to the repetitive nature of the peptide (red) and the repetitive binding surface (gray). All units are identical, and the different shades are only to guide the eye.

elongated by a conserved ladder of asparagines at position 37 (N37) of each repeat (Figure 1A). Furthermore, the affinity can be easily adjusted by varying the length of the peptide or the number of binding modules, respectively.¹⁵ Increasing the number of dipeptide units or binding modules leads to a linear increase of binding energy, with dissociation constants ranging from high nanomolar to low picomolar.¹⁵ Alanine scanning has shown a constant constant contribution to the binding energy from each arginine pocket and a smaller constant contribution from each lysine pocket.

However, the repetitive nature of both the peptide and the dArmRP can lead to additional, alternative binding modes. Recent crystal structures confirm the existence of those alternative binding modes, in which the peptide binds to the dArmRP in different registers (Figure 1B).¹⁹ However, multiple binding modes would be compromising the selection process, as they prevent the sequence-specific recognition of individual amino acids if alternative residues from a different binding register occupy a given binding pocket. One solution for this problem is the design of new binding specificities, which could lock one binding mode and would thus break the symmetry of the repetitive peptide.

In this paper, we present a design process in which we grafted a hydrophobic binding pocket of β -catenin onto the interaction surface of dArmRPs and describe how computational design was utilized to optimize its affinity and specificity. Computational protein—protein interface design can be a complementary approach for generating new binding proteins, especially with the protein design suite Rosetta.²⁰ Yet, especially the design of polar interactions can be a hurdle.²¹ Nevertheless, many successful examples have been reported,^{22–25} and predefined natural binding interfaces or hotspot residues may already provide a good starting point for the design.^{26–28} For these reasons, we proceeded from a graft to a computational improvement of the pocket.

This pocket prevents the peptide from binding in multiple binding modes (registers) by locking the peptide in the desired orientation. This is supported by the crystal structures of the complexes and experiments in solutions with single-molecule Förster resonance energy transfer (FRET), which we report here. We also provide the structural basis for how dArmRPs can differentiate between very similar amino acids, namely, isoleucine and leucine, which supports our design goal of creating sequence-specific peptide binders.

Furthermore, we show that this lock is "movable" to another position in the dArmRPs, from the end to the middle, further underlining the modularity concept. The peptide can thus also be successfully locked by having the redesigned pocket in an internal position. Two crystal structures illustrate that both the grafted and the optimized binding pockets can indeed be freely moved to an internal position while still recognizing their target sequence.

With our design strategy, we provide a basis to lock the peptide into a specific binding register on the otherwise very repetitive surface of dArmRPs. Not only is this a prerequisite for selecting sequence-specific recognition sites by common *in vitro* selection techniques but it also opens up new possibilities for the further computational design of new specific binding pockets for dArmRPs.

RESULTS AND DISCUSSION

Binding Lock to Prevent Multiple Binding Modes. Currently, modular binding of designed ArmRPs has been demonstrated for positively charged peptides or alternating arginines and lysines.¹⁵ Recently published crystal structures of regularized, consensus dArmRPs with bound peptides confirmed the modular binding mode, where each Arg-Lys dipeptide unit is recognized by one internal repeat.¹⁵ However, the repetitive nature of both the peptide and the dArmRP enables several binding modes, i.e., register shifts and flips of the peptide in the binding pocket.¹⁹ Multiple binding modes would hamper the development of new specificities by selection, as the randomized pocket on the dArmRP could be occupied by different amino acids in the different registers (Henning et al., unpublished data).

One approach to address this problem is by breaking the symmetry and repetitiveness of the peptide and the binding pocket by creating a pocket with an orthogonal binding preference. Here, we achieved this goal by grafting a hydrophobic binding pocket observed in β -catenin on a dArmRP, thus locking the peptide carrying the complementary sequence in a specific binding mode. The Armadillo repeat



Figure 2. Design and structure of Lock 1 in complex with the LSF moiety at 2.1 Å. (A and B) Design model with the dArmRP is in gray (A), and chain B of the crystal structure is in green (B). Grafted amino acids are in yellow. The LSF peptide moiety is in cyan, and the $2F_o-F_c$ map contoured at 1 σ is in black. (C) Schematic interaction map of complex shown in (B) with the coloring according to B. Only side chain interactions are shown; hydrogen bonds (length in Å) are displayed in green. The map was created with LigPlot+.⁵⁴

domain of β -catenin binds a variety of ligands, among them intrinsically disordered domains of transcription factors, which is a consequence of its diverse roles in the cell.²⁹ The central binding pocket P8 of β -catenin is of special interest, as most of the bound ligands have a phenylalanine at this position, interacting very specifically with the pocket.¹⁷

A sequence alignment of 12 β -catenin structures (including one structure of the homologous protein plakoglobin), complexed with six different ligands, revealed that this phenylalanine is always part of a hydrophobic motif, interacting with pocket P8 (Supporting Information (SI) Figure S1). We hypothesized that this pocket introduces a hydrophobic peptide-binding motif, making it sufficiently different from the polar arginine and lysine pockets, while still following the conserved binding mode. The binding motif for Leu-Ser-Phe (LSF) was chosen on the basis of an alignment of the most common bound ligands. Although isoleucine (52%) is more common than leucine (39.5%) at position P6, leucine was chosen, because the highest resolution structure available bound the motif LVF (PDB ID: 117W). Serine and valine are equally common (both 26%) at position P7, and Ser was introduced to lower the hydrophobicity of the whole motif. Finally, phenylalanine was chosen as the most common amino acid at position P8.

We inspected the residues in β -catenin that interact with this peptide motif and grafted the respective amino acids onto the binding surface of the dArmRPs to generate a binding pocket with a new specificity (Figure 2A). The pocket was moved to the N-terminus of the dArmRP to make it compatible with selections in which the randomized pocket is located in the central repeats of the dArmRP. Hereafter, the grafted pocket will be called Lock 1, comprising the following amino acid changes: R33W, S36R, Q37N, Q71S, E72S, W75G, and E114L.

To test the validity of our approach, we conducted a fluorescence anisotropy binding assay with the peptide sequence KRKRKAKLSF, which binds the dArmRP containing the Lock 1 pocket, and we determined a K_D of 83 ± 10 nM (Table 1). One alanine residue was introduced into the peptide sequence to reduce its overall binding affinity, as K_D 's

Table 1. Equilibrium Dissociation Constants of DifferentPeptides Binding to Lock 1 and Lock 2

peptide	Lock 1 $K_{\rm D} \pm$ SD (nM)	Lock 2 $K_{\rm D} \pm$ SD (nM)
KRKRKAKITE	238 ± 4	28 ± 7
KRKRKAK ITD	220 ± 20	31 ± 2
KRKRKAK LTW	114 ± 2	23 ± 1
KRKRKAKL SW	93 ± 6	21 ± 4
KRKRKAKL SF	83 ± 10	18 ± 2
KRKAKRK ASF	230 ± 54	
KRKAKRKLAF	107 ± 34	
KRKAKRK LSA	182 ± 38	
KRKRKAK ITW	74 ± 2	7 ± 0.3
KRKRKAK ATW		30 ± 1
KRKRKAK IAW		4 ± 1
KRKRKAK ITA		14 ± 4

below 1 nM are more difficult to determine accurately with this assay, and we wished to determine all affinities with the same system. Interestingly, the observed affinity is in the same range as previously reported for $(KR)_4$ peptides binding to a dArmRP of the same length, which suggests an interaction of the KLSF peptide motif with the new pocket with a strength of approximately one KR unit.¹⁵

To analyze the interactions in more detail, we cocrystallized a $(KR)_4KLSF$ peptide with the Lock 1 dArmRP containing five internal repeats and the N-terminal grafted pocket. Indeed, the crystal structure shows a specific interaction between the LSF peptide motif and the Lock 1 pocket (Figure 2 B,C). The asymmetric unit consists of two dArmRPs, both binding the peptide.

One complex shows good agreement with the model of the grafted pocket (Figure 2A). The peptide lodges slightly deeper in the pocket than in the model, with both leucine and phenylalanine binding in the new pocket, stabilized by a cation- π stacking between Arg36 of the dArmRP and phenylalanine. The exact same interaction between Arg36 and the phenylalanine was also aimed for in the grafted model, with a further stabilization of the hydrophobic patch by the leucine. Additionally, Arg36 also forms a salt bridge to the C-

terminus of the peptide. It might be noted that, between the second dArmRP and the corresponding LSF motif, a crystal contact was formed, thereby pulling the LSF motif out of its pocket.

The success of the grafting can be explained by the hydrophobic nature of the pocket. Only the hydrophobic interactions of the design were reproduced in the crystal structure but not the ionic interactions; the newly formed interaction to the C-terminus was not predicted but is a result of the peptide lodging deeper in the whole pocket, thereby allowing the new interaction to be formed. Our design protocol only rewards total interaction energy, and we have not attempted at this point to break it down into individual components during design. The overall success of this design is also in line with the analysis of computational interface designs by Stranges and Kuhlman,²¹ who suggested that hydrophobic designs are still more likely to succeed than highly polar ones.

Computational Design Increases the Affinity of the Grafted Pocket. In the next step, we aimed to improve the specificity and binding affinity of the Lock 1 pocket. An alanine scan of the peptide revealed that the leucine residue contributed most to the binding but the serine and phenylalanine did so only marginally (Table 1). To improve the affinity, we used the crystal structure of the grafted pocket as a starting point for a computational redesign using the Rosetta Scripts interface design protocol.^{30,31}

On the basis of the idea that the whole surface of the Lock 1 pocket can potentially contribute to binding, we allowed mutations in a sphere around 12 Å. Additionally, positions W33 and N37 were changed back to the original amino acids (R33 and Q37) of the dArmRP, as they did not interact significantly with the peptide in the crystal structure. Different short peptide motifs were considered as targets for the design: LSF, KLSF, and AKLSF were used as binding partners for the dArmRP. The sequence AKLSF was finally chosen in order to decrease side effects of the neighboring arginine on the outcome of the design, which would have led to the design of an RKLSF pocket. The output design models were analyzed with the interface analyzer in Rosetta, and the model with the lowest computed Rosetta $\Delta\Delta G$ (ddG) energy value from each setup was chosen for expression and affinity measurements.³⁰ All models showed only few mutations, ranging from six to eight, of which five were conserved in all designs (SI Figure S2). After measuring the affinities to the peptide KRKRKAKLSF, we found that all designs bind better to this peptide than the original pocket (Table 2). The pocket from

 Table 2. Equilibrium Dissociation Constants of the

 Different Designs Binding the Peptide KRKRKAKLSF

peptide	$ \begin{array}{c} \text{Lock 1} \\ K_{\rm D} \pm \text{SD} \\ (\text{nM}) \end{array} $	AKLSF design K _D ± SD (nM)	KLSF design K _D ± SD (nM)	LSF design $K_{\rm D} \pm { m SD}$ (nM)
KRKRKAK LSF	83 ± 10	18 ± 2	51 ± 11	29 ± 2

the redesign with the AKLSF peptide has the best affinity (18 \pm 2 nM), always measured in the context of KRKRKAKLSF, a 4-fold increase when compared to the original pocket (Table 2).

The crystal structure of the new design (Lock 2, Figure 3A,B) with the bound peptide KRKRKAKLSF demonstrated that the peptide is bound in a similar way as before. The asymmetric unit contains six molecules, all of which bind the

peptide in exactly the same way. Among them, chain B shows the best electron density for the LSF moiety. There, both leucine and serine are well-defined, but the terminal phenylalanine has an elevated B-factor, indicating flexibility of the side chain. The mutations R36Q and N37Q make additional hydrogen bonds to the backbone and to the serine side chain, respectively, whereas S40M increases the contact area between the peptide and the dArmRP. The mutations W117I and L114F increased the hydrophobic interaction of the peptide; also, the threonine is involved in a hydrogen bond to Q37.

L114F increases the hydrophobic interaction of the peptide and provides a better hydrophobic pocket for the leucine. Furthermore, the interaction of the peptide C-terminus to R36 is shifted to R33 (Figure 3A,B). The mutation S78A is not directly involved in peptide binding but has been observed in all Rosetta interface designs of dArmRPs so far and might lead to a better packing of the helices.

Further Redesign of the Peptide Additionally Increases the Affinity. The structure of the dArmRP with a new Lock 2 pocket in the complex with the KRKRKAKLSF peptide showed that this alternative solution to the original pocket can bind the peptide with higher affinity. Yet, key interactions like the cation– π stacking between Arg36 and the phenylalanine of the peptide were not observed in the new design. For this reason, and because of the flexibility of the terminal phenylalanine, which might not fit perfectly in the pocket, we decided to do a second round of computational design on Lock 2, this time changing the peptide sequence instead of the pocket to investigate whether another peptide sequence would bind even better to the new lock.

The design was carried out analogously to the first round, and a predominant outcome of the design was the sequence LTW, with 82 out of 100 designed sequences. Besides LTW, especially LSW (10%) and ITW (6%) were predicted to bind. We decided to experimentally test the top 5 sequences for their ability to bind to the new pocket.

Interestingly, most of the new sequences showed a higher affinity (lower dissociation constant $K_{\rm D}$) than the original LSF motif, and one sequence showed the highest affinity among all other peptide motifs. The KRKRKAKITW peptide binds with an affinity of 7 ± 0.3 nM, a more than 2-fold decrease of the $K_{\rm D}$ (Table 1). The crystal structure, which was obtained in the same condition and space group as for the LSF peptide, explains this by showing that ITW fits deeper into the pocket than LSF (Figure 3C,D). The electron density is again not identical among the six protein-peptide complexes of the asymmetric unit, but all of them exhibit the same binding. Chain D was used for further analysis. The isoleucine is bound in a similar way as the leucine before. Furthermore, a cation $-\pi$ interaction between the tryptophan and R33 is observed. Like the serine of the peptide, also the threonine is involved in a hydrogen bond to Q37.

These interactions, combined with the fact that, with isoleucine and tryptophan, two larger hydrophobic residues were introduced, can explain the higher affinity of the ITW peptide compared to that of the LSF peptide. This observation is also consistent with the increase in shape complementarity (SC) between the peptide and dArmRP. It should be noted that for small ligands like peptides, the SC can be affected by edge effects, but when we compare the three similar interfaces, this effect should be minor.³² The original pocket (Lock 1) binding the LSF motif has an SC of 0.76, which is comparable to typical antibody–peptide complexes.³² Importantly, both

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Figure 3. Crystal structures of Lock 2 in a complex with the AKLSF and AKITW peptide moieties. dArmRPs are in green. Mutated positions are in yellow. The peptide is in cyan, and the $2F_o-F_c$ map contoured at 1 σ is in black. (A) Chain B of Lock 2 with AKLSF bound at 2.1 Å resolution. (B) Schematic interaction map of complex shown in A with the coloring according to B. Only side chain interactions are shown, and hydrogen bonds (length in Å) are in green. Map created with LigPlot+.⁵⁴ (C) Chain D of Lock 2 with AKITW bound at 2.0 Å resolution. (D) Schematic interaction map of the complex shown in C with the coloring according to C. Only side chain interactions are shown, and hydrogen bonds (length in Å) are in green. Map created with LigPlot+.⁵⁴

Table 3. Kinetic Binding Constants of the Y _{III} M ₅ A _{II} :(KR) ₅ Complex and the Th	hree Terminal Designed I	Peptide Pockets Bound
to the Respective Peptides			

complex	$k_{\rm on}~({\rm M}^{-1}~{\rm s}^{-1})$	$k_{\rm off}~({ m s}^{-1})$	$K_{\rm D}~({\rm nM})$ calculated	$K_{\rm D} \pm {\rm SD} \ ({\rm nM})$ from anisotropy
$Y_{III}M_5A_{II}:(KR)_5$	$9.7 \times 10^8 \pm 1.3 \times 10^8$	1.1 ± 0.1	1	1.1 ± 0.8^{a}
Lock 1:KRKAKRKLSF	$3.3 \times 10^8 \pm 0.0 \times 10^8$	46.6 ± 3.1	166	83 ± 10
Lock 2:KRKRKAKLSF	$3.0 \times 10^8 \pm 0.3 \times 10^8$	5.5 ± 1.3	20	19 ± 2
Lock 2:KRKRKAKITW	$3.2 \times 10^8 \pm 0.0 \times 10^8$	1.6 ± 0.0	5	7 ± 0.3
^{<i>a</i>} As determined in ref 15.				

LSF (0.809, chain B) and ITW (0.848, chain D) peptides show a higher SC when bound to the new pocket (Lock 2).

Lock 2 Pocket Is Specific for Isoleucine over Leucine. By testing all the different peptide motifs, we found a remarkable difference in affinity between the leucine and isoleucine variants. Whereas the LTW peptide binds with an affinity of 23 ± 1 nM, ITW binds with 7 ± 0.3 nM, which is a difference of a factor of 3 between these very similar amino acids in a pocket that was not specifically designed for isoleucine. By comparison of the structures with both the LSF and the ITW peptides, it becomes obvious that isoleucine fits better into the optimized pocket (SI Figure S3). Leucine has a branch at the C γ atom that leaves a hole opposite Gly75, which is filled by isoleucine, which has a branch at the C β atom. Furthermore, the branch at the C γ atom of leucine could also result in a clash with I117. Consequently, the observed rotamer of this leucine residue is the only one that is not involved in any potential clashes between the side chain and the



Figure 4. Register shifts in dArmRP:peptide complexes observed by single-molecule FRET in solution and reduced by incorporating Lock 2. (A) Transfer efficiency histograms of the $Y_{III}M_5A_{II}$:(KR)₅ complex (top) and the Lock 2:ITW complex (bottom). dArmRP was labeled with an acceptor dye (CF660R) at position 51, and the peptide with a donor dye (Cy3B) at the N-terminus. Measured transfer efficiency histograms (gray) are compared to a fit (red line) with three populations calculated from probability distribution analysis (PDA³⁶), assuming that the peak widths are shot noise-limited. Subpopulations share the same peak width and position and are shaded in green, blue, and orange, respectively. Their fitted mean transfer efficiency, E_{fit} is given on the top and their relative contributions to both histograms on the right. (B) Cartoon representations of $Y_{III}M_5A_{II}$:KR₅ in the -1 sliding state (top), the optimal bound state (middle), and the +1 sliding state (bottom). The sterically accessible volumes of the donor and acceptor dyes are depicted as green and red clouds. Next to each structure, the calculated mean transfer efficiencies are given.

surrounding amino acids. This observation is very encouraging for the design of sequence-specific pockets.

Both Design Rounds Resulted in a Decreased Off-Rate. To understand the nature of the interaction of peptides with dArmRPs better, we developed a stopped-flow Förster resonance energy (FRET) assay to determine the binding and dissociation rates for both the dArmRPs (containing Lock 1 and Lock 2) and the peptides (KRKAKRKLSF and KRK-AKRKITW). For this purpose, the dArmRP and the peptide were fused to the fluorescent proteins mCherry and sfGFP, respectively. The association was probed by following the increase in mCherry fluorescence upon binding when sfGFP was excited. The dissociation rate constant was determined in a competition assay, where a preformed dArmRP–mCherry/ peptide–sfGFP complex was rapidly mixed with an excess of unlabeled peptide, and the decrease of acceptor fluorescence was monitored.

Table 3 and SI Figures S4–S7 show that all dArmRP– peptide interactions share the same very high association rate constant (>10⁸ M⁻¹ s⁻¹), consistent with an electrostatically driven interaction of an unstructured peptide without rotational restrictions. For the interaction of Y_{III}M₅A_{II} with (KR)₅ ($k_{on} = (9.7 \pm 1.3) \times 10^8 \text{ M}^{-1} \text{ s}^{-1}$), a dArmRP with five identical binding repeats each recognizing a lysine–arginine (KR) dipeptide unit, an even higher association rate than the dArmRPs with the new pockets (by a factor of 3) is seen.

These rates are close to the limit set by diffusion³³ and are probably accelerated by electrostatic attraction between the positively charged peptide and the negatively charged dArmRP, which explains also the higher k_{on} of the Y_{III}M₅A_{II}: (KR)₅ complex compared to the LSF- or ITW-containing constructs. It should be noted that the fluorescent fusion proteins carry only a miniscule net charge of -3 at the pH of the experiment. Furthermore, the constant KRKRKAK motif of all Lock–peptides may contribute the majority to the fast association, which explains why all rate coefficients are nearly identical. The dissociation rate constants of dArmRPs containing Lock 1 and Lock 2 show larger differences, ranging from fast dissociation of the Lock 1:KRKRKAKLSF peptide complex (46.6 ± 3.1 s⁻¹) to a 30-fold slower dissociation in the optimized Lock 2:KRKRKAKITW peptide complex (1.6 ± 0.03 s⁻¹). This observation strongly indicates that the increase in binding affinity is mainly driven by a decrease of k_{off} which we attribute to the additional interactions within the peptide lock.

Designed Pocket Prevents Sliding in Solution. The different binding registers between peptide and dArmRP, whose elimination was the driving force behind this study, were so far only observed in crystal structures. It might thus be possible that these register shifts are induced by crystal packing forces. Therefore, we conducted single-molecule FRET experiments to probe whether register shifts also occur in solution.³⁴ Two peptide-dArmRP complexes were investigated: first, the symmetric Y_{III}M₅A_{II}:(KR)₅ complex, where register shifts have been observed in the crystal structure; second, the Lock 2:KRKRKAKITW complex, where the peptide should be locked into one preferential binding register. In both cases, the peptide was labeled with a donor dye (Cy3B) at an N-terminal cysteine residue and the ArmRP with an acceptor dye (CF660R) at position 51 (mutated to cysteine). The measured transfer efficiency is directly related to the distance between the two dyes, which will be affected by register changes (Figure 4). Single-molecule FRET captures "distance snapshots" (time resolution of ~ 1 ms) of individual molecules as they diffuse through the laser focus, so each binding register will generate a distinct transfer efficiency peak if it is populated in solution.

The transfer efficiency (*E*) histograms of the $Y_{III}M_5A_{II}$: (KR)₅ and the Lock 2:KRKRKAKITW complex are shown in Figure 4. Both histograms share a very similar average peak position. However, the peak is much broader for $Y_{III}M_5A_{II}$: (KR)₅, indicating that in this complex, a broader distribution of states is populated. For a quantitative description, we first conducted a recurrence analysis of single particles (RASP)³⁵ to test how many subspecies are required to describe the transfer efficiency histograms (see SI Figure S8). RASP revealed that at least three species are contributing to the histogram of $Y_{III}M_5A_{II}$: (KR)₅, potentially corresponding to one preferred binding mode and sliding of the peptide to either side of the dArmRP.

We hypothesized that the optimal binding mode is populated to a larger extent in the Lock 2:KRKRKAKITW complex, and the two shifted binding modes contribute less, leading to a narrower E distribution compared to that of Y_{III}M₅A_{II}:(KR)₅. Nonetheless, for each binding register, the underlying interdye distance distributions should be the same in both complexes, as the peptide backbone is bound in the same way. Therefore, we fitted both histograms with three populations using photon distribution analysis (PDA³⁶), which calculates the shape of the transfer efficiency peaks by taking into account shot noise and the statistical fluctuation of recorded numbers of donor and acceptor photons about their mean values. We obtained a very good fit to the recorded E histograms by assuming the underlying E distributions to be very narrow and sharing the same peak positions for both dArmRP-peptide complexes (see Figure 4). The resulting fit suggests that the optimal binding mode (shaded in blue) is occupied only to 50% in the Y_{III}M₅A_{II}:(KR)₅ complex. In the Lock 2:KRKRKAKITW complex, however, it dominates the distribution with 68% occupancy. We note that this value increases to 85% when we allow slight peak broadening beyond shot noise (a common observation in many single-molecule FRET studies;³⁷ see SI Figure S9), so the value of 68% likely represents a lower limit for the population of the optimal state in the Lock 2:KRKRKAKITW complex.

To compare the FRET results with the atomic models for the dArmRP-peptide complexes, we calculated the transfer efficiencies expected from the complex structures. The $Y_{III}M_5A_{II}$:(KR)₅ structure (PDB ID: 5AEI) served as a model for the optimal binding mode, and the sliding states (register shifts) were modeled with Coot.³⁸ The dyes can sample a large range of interdye distances as they are attached to the protein by long flexible linkers. For each structure, we calculated the geometrically accessible volumes for both dyes³⁹ and estimated the mean transfer efficiencies, assuming fast interdye distance fluctuations. The analysis yields transfer efficiencies of 0.74, 0.83, and 0.87 for the -1, optimal, and +1states, respectively, close to the measured values (0.76, 0.84, and 0.92; typical accuracy ± 0.03). In conclusion, the singlemolecule FRET experiments support the presence of sliding in solution for the symmetric Y_{III}M₅A_{II}:(KR)₅ complex and suggest that it is significantly reduced for the restrained Lock 2:KRKRKAKITW complex, in line with the crystallographic results

Binding Pockets Can Be Transferred to an Internal Position. The peptide lock, located in the N-terminal region of the ArmRP, was designed to bind the C-terminal amino acids of the peptide. Such a terminal peptide lock would prohibit peptide register shifts and can be used with a randomization strategy in the middle of the protein in selections for a new amino acid side chain located in the middle of the peptide. To test the modularity of our system, we moved the module, recognizing the new amino acids (K)LSW or (K)ITW, orthogonal to KRKR.

To test whether the pocket can be moved to the middle of the dArmRP, we constructed two different internal locks. On the basis of the structural similarity of the N-cap and the internal repeats, we took the structure PDB-ID 5MFM, a dArmRP with six internal repeats binding to a $(KR)_5$ peptide, and grafted the binding residues of the new pockets onto the internal repeats two to four, keeping to their relative positions like at the N-terminus. For the Lock 1 pocket, which binds to LSF, we knew that only the leucine and the phenylalanine were involved in binding. Therefore, we designed the new peptide with a lysine at the position of the serine, having the sequence KRKRKLKFKR. For the improved pocket, also the threonine displayed an interaction in the crystal structure. Therefore, we also tested the sequence KRKAKITWKR analogously to the terminal AKITW motif from the original design.

Both new dArmRPs bind to the respective peptides with an affinity of 298 ± 45 nM (internal Lock 1) or 102 ± 19 nM (internal Lock 2), and an alanine scan on the lock binding residues of the peptide revealed that the lock is indeed involved in binding (Table 4). However, the overall affinity is

 Table 4. Dissociation Constants of the Respective Peptides

 Binding to the Internal Lock Variants

peptide	internal Lock 1 $K_{\rm D} \pm SD$ (nM)	internal Lock 2 $K_{\rm D} \pm {\rm SD}$ (nM)
KRKRKLKFKR	298 ± 45	
KRKRK A KFKR	892 ± 47	
KRKRKLKAKR	880 ± 28	
KRKAKITWKR		102 ± 19
KRKAKATWKR		733 ± 12
KRKAKIAWKR		69 ± 5
KRKAKIT A KR		171 ± 20

lower than for the initial lock designs $(83 \pm 10 \text{ nM} \text{ for Lock } 1 \text{ and } 7 \pm 0.3 \text{ nM} \text{ for Lock } 2)$, where the respective amino acids and the pockets were placed at the termini of the peptide and the dArmRP.

The structures of the internal pockets in complex with their target peptides show that the peptides are bound differently on the surface of the dArmRP, because of their different designs (Figure 5). The internal Lock 1 binds the peptide in the conserved binding mode, with every second peptide bond being involved in bidentate hydrogen bonds to the conserved N37 ladder. Additionally, all arginines and lysines are bound in their respective pockets, and the side chains from the new peptide fit into the internal pockets. This binding mode is therefore as expected, since the peptide backbone is kept in place by the interaction to the N37 ladder. Yet, it deviates slightly from the binding mode of the terminal Lock 1 pocket. There, the C-terminus of the peptide is free to move deeper into the hydrophobic pockets, which is not possible in the internal position, as the backbone is fixed by binding to N37.

In contrast to the results with Lock 1, internal Lock 2 exhibits a different binding between a terminal and an internal location. The peptide is only bound up to the ITW moiety. The ITW moiety shows a similar conformation in the internal position when compared to ITW in the terminal position, although the terminal tryptophan is not bound as deep into the



Figure 5. Crystal structures of the internal Lock pockets with the bound peptides. dArmRPs are in green. Mutated positions are in yellow. The peptide is in cyan, and the $2F_o-F_c$ map contoured at 1 σ is in black. (A) Chain A of internal Lock 1 with the bound peptide. Shown is the KLKFKR motif of the peptide (3.17 Å). (B) Chain A of internal Lock 2 with the bound peptide. Shown is the AKITW motif of the peptide. Terminal KR residues were not fully resolved in the electron density, and only the $C\alpha$ atom of the lysine could be fitted (2.8 Å).

pocket as in the terminal position. However, no clear electron density is visible for the additional arginine and lysine residues which succeed the ITW sequence, indicating that they do not bind into their respective pockets, and the whole peptide stretch is effectively locked into position by just the ITW moiety. Furthermore, the interaction of the backbone with the conserved N37 is not possible in the internal Lock 2 pocket. At this position, the N37 was mutated to a glutamine, which interacts with the threonine of the ITW motif and not with the backbone of the peptide. By this interaction, the whole peptide is locked into this binding register at the expense of constraining the peptide backbone to the protein. This also explains the lower affinity, compared to a peptide of the same composition and length but with a terminal ITW motif, where the lysine and arginine can still bind.

CONCLUSIONS

Modular binding scaffolds have the potential to overcome time-intense and expensive selections for individual peptide binders, as a preselected set of specific binding modules can be reassembled to bind arbitrary peptide targets.^{17,18} We had observed that register shifts, i.e., multiple binding modes, can exist, which could interfere with the desired position-specific recognition of individual amino acids.¹⁹ Furthermore, a specific binding register is mandatory during selections where a new target amino acid is facing a randomized pocket. Here, we show that this problem of peptide register shifts of repeating units can be overcome by introducing a designed peptide lock. We created the new binding site neither by complete de novo design nor by selection (as the latter would be dependent on already having a functional lock), but we grafted a natural hydrophobic binding pocket of β -catenin onto our dArmRP scaffold. In our study, we reached the following conclusions:

(i) We found that one of the two designed pockets (Lock 2) was able to differentiate between isoleucine and leucine by a factor of 3 (ITW:LTW). This result shows that dArmRPs have the potential to differentiate between very similar amino acids, and this finding is important for the further design for specificity.

(ii) By developing a stopped-flow-based FRET assay, we noted very high k_{on} values. This can be explained by the

mechanism of binding, which involves an elongated and short unstructured peptide stretch, furthermore involving charge interactions between a protein and a peptide. This increases the likelihood of successful collisions, as observed in other systems.^{40,41}

(iii) Single-molecule FRET studies indicated multiple binding modes in solution for the repetitive system and thus showed that different binding modes (registers) are not just crystallization artifacts but do also exist in solution. In contrast, a much larger fraction of the locked peptide is bound in the desired register, probably closer to 85%, if we consider that there may be a slight peak broadening beyond shot noise. This underlines that we have successfully created a peptide lock with only one predominant binding mode.

(iv) We found that the two new pockets can be moved from an N-terminal to an internal position, thereby creating the first designed modular peptide binders in which the modules can be easily interchanged. Both the biochemical and structural analysis of these shuffled dArmRP showed an interaction with the respective peptide, with the structures displaying no peptide register shift, as desired, with the same interactions (Lock 2) as at the terminus.

In summary, we describe the design of a hydrophobic binding pocket for dArmRP repeat proteins to solve a particular problem, namely, the sliding and flipping of repetitive peptides on repetitive dArmRPs. A combination of grafting and computational redesign resulted in a sequencespecific pocket that is orthogonal to those recognizing positively charged amino acids. The new design can discriminate between two very similar amino acids, isoleucine and leucine. With this study, we also give a first example of a workflow for the computational design of pockets for dArmRPs. Furthermore, we show that this new pocket prevents multiple binding modes and can be freely moved to an internal position. This demonstrates the potential of dArmRPs as modular peptide binders. The collection of pockets for other amino acids will now be extended by a combination of selection and de novo design, testing new algorithms, score functions, and different constraints.

METHODS

Cloning, Protein Expression, and Purification. Cloning was carried out as described.^{9,13} DNA fragments coding for the respective proteins were ordered from IDT or Genewiz and subcloned into a pQE30LIC 3C-based vector as described.¹⁵ Protein expression and purification were done as described previously,¹⁶ or alternatively, expression was done in an autoinduction medium for 15 h at 25 °C.

Crystallization and Structure Determination. Proteins were concentrated to 20-150 mg mL⁻¹ in 10 mM Tris/HCl at pH 8 and 100 mM NaCl using Amicon centrifugal concentrators (Amicon Ultra Centrifugal Filters, Merck Millipore), and a 1.5-fold molar excess of peptide was added to the Lock 1 protein. Screening for suitable crystallization conditions was done in commercially available 96-well sparse matrix screens (Hampton Research and Molecular Dimensions). Per condition, the mother liquor was mixed in three different ratios (1:1, 2:1, and 3:1 or 5:1) with the protein in 300–400 nL drops and 75 μL of reservoir solution; plates were incubated at 4 °C. For structure PDB-ID 6S9L, a fine-screen was set up on the basis of the initially found conditions in a 96-well format by changing the pH along the column and the precipitant along the row.

Prior to data collection, crystals were flash-frozen in liquid nitrogen directly or after incubation for 10 s in a mother liquor containing 20% (v/v) ethylene glycol. Diffraction data were collected at beamline X06SA (Paul Scherrer Institute, Villigen, Switzerland) equipped with an Eiger 16 M detector (Dectris) at 1 Å. Crystallization conditions and data collection and refinement statistics are summarized in SI Table S1. Data were processed with XDS, XSCALE, and XDSCONV.⁴³ Phases were determined by molecular replacement using PHASER,⁴⁴ with structures of dArmRPs containing five or six internal repeats. Model building was done in Coot³⁸ and refinement using REFMAC5,45 PHENIX refine,46 and BUSTER.47 The final resolution of the data sets was determined using paired refinement in pdb redo,⁴⁸ based on ref 49.

Determination of Binding Constants. All experiments were conducted in PBS, supplied with 0.01% (v/v) Tween 20. Fluorescence anisotropy data were measured according to ref 15 with two to three measurements per point. Binding kinetics were determined using a FRET-based binding assay. Increasing concentrations of mCherry-dArmRP fusions were mixed with a constant concentration of sfGFP-peptide fusions in a 1:1 volume ratio in 120 µL in a PiStar-180 stopped-flow fluorometer (Applied Photophysics Ltd.) equipped with a mercury-xenon lamp. sfGFP was excited using a wavelength of 436 nm and a bandwidth of 10 nm. Fluorescence was collected through a 590 nm long-pass filter. For obtaining $k_{\rm on}$, three kinetic association curves at different mCherry-dArmRP concentrations were recorded. A single-exponential curve was fit to the data, and the resulting rates were plotted against the concentration of mCherry-dArmRP (120, 160, 200 nM). The final k_{on} was then determined by fitting a linear function to these data points. An additional point at 80 nM was also recorded but not used for the fit, as the concentration was only 2 times above the peptide-sfGFP concentration. For determining k_{off} the peptide-protein complex was preincubated and mixed with two different concentrations of competitor, nonlabeled dArmRP. A single-exponential decay was fit to the data.

Single-Molecule FRET Experiments. For labeling of dArmRPs, Y_{III}M₅A_{II} (a dArmRP with five internal KR binding repeats; Y_{III} and A_{II} are the third- and second-generation capping repeats, respectively), and Lock 2, a cysteine was introduced at position D51 via sitedirected mutagenesis. Protein expression and purification were done as described above with 1 mM 2-mercaptoethanol (Pierce) in all purification buffers. Proteins were further purified via size exclusion chromatography in PBS, 1 mM DTT, and consecutive anion exchange chromatography on a MonoQ 5/50 GL column to remove the DTT in 20 mM NaH₂PO₄/Na₂HPO₄, at pH 7.2. Proteins were eluted with a gradient from 0 to 1000 mM NaCl. Following the elution, the proteins were labeled using CF660R-maleimide dye (Biotium) in 1.3fold molar excess for 16 h at 4 °C. After labeling, the excess dye was removed with a second anion exchange step. Peptides (CSAGGK-

RKRKRKR and CSAGGKRKRKAKITW) were ordered from Life-Tein as a lyophilized powder and dissolved in 20 mM NaH₂PO₄/ Na₂HPO₄, at pH 7.2, and 1 mM TCEP. They were further purified by reverse-phase HPLC on a Reprosil Gold 200 C18 column (Dr. A. Maisch GmbH), using a H_2O + 0.1% (v/v) TFA/acetonitrile gradient. Fractions containing the full-length peptide were pooled and lyophilized. For labeling, the peptides were dissolved in 100 mM KH₂PO₄/K₂HPO₄, at pH 7.0, and a 1.3-fold excess of Cy3Bmaleimide (GE Healthcare) was used for labeling the N-terminal cysteine. The reaction was incubated at room temperature (RT) for 3 h, followed by a second HPLC step to remove excess dye. The labeled peptides were again lyophilized and stored at -20 °C until further use. For the measurement, they were redissolved in 50 mM NaH₂PO₄/Na₂HPO₄, pH 7.0, supplemented with 0.001% (v/v) Tween 20.

Measurements were recorded for 1 h with Cy3B-labeled peptide concentrations of 50 pM and CF660R-labeled dArmRP concentrations of 5 nM, in 50 mM NaH₂PO₄/Na₂HPO₄, at pH 7.0, supplemented with 0.001% (v/v) Tween 20. A plastic sample chamber (μ -slide Angiogenesis, ibidi) was used to minimize surface adhesion.

All single-molecule experiments were conducted on a custom-built confocal instrument equipped with a 532 nm continuous-wave laser (LaserBoxx LBX-532-50-COL-PP, Oxxius) to excite the donor dye.⁵ Fluorescence photons were collected through a high numerical aperture objective (UPlanApo 60×/1.20-W, Olympus), subsequently separated from the scattered photons with a triple-band mirror (zt405/530/630rpc, Chroma), and split onto four channels according to their wavelength and polarization. Dichroic mirrors were used to separate donor and acceptor emissions (T635LPXR, Chroma). Donor photons were filtered with ET585/65m bandpass filters (Chroma) before detection on one of two τ -SPAD avalanche photodiodes (PicoQuant). Acceptor photons were filtered with LP647RU long pass filters (Chroma) and detected with SPCM-AQRH-14 singlephoton avalanche diodes (PerkinElmer). The arrival time of every photon was recorded; the photon time trace was binned at 0.75 ms, and all bins with at least 80 photons in them were retained as a burst. Identified bursts were corrected for the quantum yield of the dyes, the detection efficiency of the detectors, crosstalk, and direct excitation of the acceptor.⁵¹ Bursts during which acceptor photobleaching likely occurred were eliminated.⁵² The bursts were binned in a transfer efficiency (E) histogram according to the corrected number of photons detected in the donor (n_D) and acceptor (n_A) channels: $E = \frac{n_{\rm A}}{2}$

 $n_A + n_{\Gamma}$

In RASP, the single-molecule data are filtered for pairs of photon bursts that were measured so close in time that they most likely were emitted from one and the same molecule visiting the confocal volume twice. As long as the molecule does not change its conformation, both bursts will sample the same underlying transfer efficiency distribution of a subpopulation. This way, we are able to reconstruct individual subpopulations from a histogram with overlapping peaks.

RASP analysis of the transfer efficiency histogram was conducted as described before.³⁵ Burst pairs separated by less than 5 ms were selected for the analysis. From the time correlation of the bursts, we calculate that after 5 ms the probability that the bursts originate from the same molecule is still 85%. We filter for groups of burst pairs where the transfer efficiency of the first burst is in an interval of $\Delta E =$ $E_2 - E_1 = 0.05$, ranging from E_1 of 0.65 to 1 in steps of 0.025. The second bursts of each pair are used to construct "pure" species histograms. Those 15 resulting histograms were normalized to an area of 1 and globally fitted with one to four Gaussian peak functions to calculate the χ^2 value used to assess the quality of the fit.

Probability distribution analysis (PDA)³⁶ was used to analyze transfer efficiency histograms. PDA assumes an intrinsic distribution of experimental transfer efficiencies, which is further broadened by shot noise due to the limited number of photons detected per burst. Shot-noise broadening is modeled using the photon statistics of the experimental burst size distribution. Here, we applied PDA, assuming that the peptide-dArmRP complex can exist in three states that have the same intrinsic transfer efficiency distributions for both the $Y_{III}M_5A_{II}$:(KR)₅ complex and the Lock 2:ITW complex. Unlike the original PDA approach,³⁶ however, we do not calculate the intrinsic distribution from a hypothetical normal distribution of interdye distances. Instead, we model the intrinsic transfer efficiency distribution, p(E), for each subpopulation directly in transfer efficiency space in the form of a β distribution:

$$p(E) = \frac{(1-E)^{(1-\varepsilon)\nu-1}E^{\varepsilon\nu-1}}{B(\varepsilon\nu, (1-\varepsilon)\nu)}$$

where B(x,y) is the β function, $\varepsilon = \langle E \rangle$ is the mean of the distribution, and its variance is given by $\sigma^2 = \varepsilon(1 - \varepsilon)/(1 + v)$ with the parameter v > 0. In Figure 4, we assumed an infinitely narrow intrinsic transfer efficiency distribution $(v \to \infty)$, and in SI Figure S8 we left v as a free fit parameter but constrained it to be identical for all three subpopulations.

Geometrically accessible volumes of dyes were calculated from different models of the Y_{III}M₅A_{II}:(KR)₅ complex (models were generated in Coot based on PDB-ID 5AEI), which served as model structures for the ligand-bound and ligand-free states, respectively. All calculations were performed using the freely available "FRET Positioning and Screening" software (http://www.mpc.hhu.de/en/ software/fps.html), in which Cy3B and CF660R were parametrized with the following parameters (linker length/width/dye radii in Å): 37/4.5/(3.4/8.2/3) for Cy3B and 19/4.5/(8.1/4.2/2.1) for CF660R. The long linker length for Cy3B arises because we assume the five Nterminal amino acids of the peptide to be flexible, as they have no binding pocket on the dArmRP (assuming a stretched chain; adding 19 Å to the Cy3B linker length of 18 Å). To take into account the rapid fluctuations in interdye distance that occur on time scales similar to the excited state lifetime of the donor, we modeled the dynamics as diffusive motion in the potential of mean force corresponding to the interdye distance distribution from the accessible volume calculations using the diffusion coefficient of the free dyes.⁵

ASSOCIATED CONTENT

Supporting Information

The Supporting Information is available free of charge at https://pubs.acs.org/doi/10.1021/acschembio.9b00928.

Figures: sequence alignment, surface representation, determinations of kinetic binding constants, fits of kinetic binding constants, RASP analysis, transfer efficiency histograms; tables of data collection and refinement statistics, and Rosetta Scripts: code for redesigns (PDF)

Accession Codes

Protein models have been deposited at the Protein Data Bank under the accession codes 6S9L (Lock 1:(KR)₄KLSF), 6S9M (Lock 2_(GS)₆_KRKRKAKLSF), 6S9N (Lock 2_(GS)₆_KRKRKAKITW), 6S9O (internal-Lock_2_(GS)₆KRKRKLKFKR), and 6S9P (internal-Lock_2_(GS)₆_KRKAKITWKR).

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Author Contributions

The manuscript was written through contributions of all authors. All authors have given approval to the final version of the manuscript.

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Notes

The authors declare no competing financial interest.

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ABBREVIATIONS

dArmRP, designed Armadillo repeat protein; nArmRP, natural Armadillo repeat protein; DARPin, designed Ankyrin repeat protein; GFP, green fluorescent protein

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