

# ScrepYard: An online resource for disulfide-stabilized tandem repeat peptides

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#### Abstract

Receptor avidity through multivalency is a highly sought-after property of ligands. While readily available in nature in the form of bivalent antibodies, this property remains challenging to engineer in synthetic molecules. The discovery of several bivalent venom peptides containing two homologous and independently folded domains (in a tandem repeat arrangement) has provided a unique opportunity to better understand the underpinning design of multivalency in multimeric biomolecules, as well as how naturally occurring multivalent ligands can be identified. In previous work, we classified these molecules as a larger class termed secreted cysteine-rich repeat-proteins (SCREPs). Here, we present an online resource; ScrepYard, designed to assist researchers in identification of SCREP sequences of interest and to aid in characterizing this emerging class of biomolecules. Analysis of sequences within the ScrepYard reveals that two-domain tandem repeats constitute the most abundant SCREP domain architecture, while the interdomain "linker" regions connecting the functional domains are found to be abundant in amino acids with short or polar sidechains and contain an unusually high abundance of proline residues. Finally, we demonstrate the utility of ScrepYard as a virtual screening tool for discovery of putatively multivalent peptides, by using it as a resource to identify a previously uncharacterized serine protease inhibitor and confirm its predicted activity using an enzyme assay.

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#### KEYWORDS

bioactive, bivalent, disulfide-rich, multivalent, peptide, SCREPs, secreted proteins, tandem-repeat

#### **1** | INTRODUCTION

Multivalency is a common property of biomolecules that describes the interaction between two molecules through multiple nonoverlapping binding interfaces. The advantage of multivalency is two-fold (1) higher specificity due to a larger interaction interface, and (2) enhanced binding kinetics and thermodynamics that result in high avidity (Mammen et al., 1998; Vauquelin & Charlton, 2013). Nowhere is this better recognized than in the adaptive immune system where antibodies use multivalency as a key mechanism in responding to infections through the dimeric nature of the antigen recognizing regions and the symmetry in the Y-shaped structure (Uvyn & De Geest, 2020). Mimicry of this process has resulted in the field of antibody therapeutics, which have had a tremendous impact on contemporary pharmaceutical development (Miller & Lanthier, 2015). Despite the success of antibody therapeutics, there are a number of limitations; these include a requirement of a good (unique and accessible) antigen, relatively poor thermal and chemical stability, and that antigen recognition may or may not lead to the desired (or any) functional outcome (Rodgers & Chou, 2016). Where antibodies are limited, small molecules often excel, with the caveat of poor selectivity that can potentially lead to serious side-effects. Peptides offer an attractive middle ground, providing higher specificity than small molecules due to their larger binding interface whilst being as functionally potent as antibodies. Indeed, peptides have received substantial attention over the past few decades, demonstrating an exceptional capacity for use as molecular probes which target many therapeutically relevant biomolecules (Dutertre & Lewis, 2010; Muttenthaler et al., 2021; Pennington et al., 2018). They are also increasingly being developed into novel therapeutics, with approximately 80 peptide drugs now approved for use, and over 150 peptides currently undergoing clinical trials (Muttenthaler et al., 2021).

Disulfide-rich peptides (DRPs) have emerged as a particularly attractive class of peptides due to their covalent intramolecular disulfide bonds. These bonds act as crossbraces to increase structural stability and backbone rigidity, resulting in resistances to proteolysis and extreme physicochemical conditions (i.e., extremes of pH and temperature) (Gongora-Benitez et al., 2014). The majority of DRPs characterized to date are highly potent neurotoxins isolated from animal venoms and consist of a single domain (Mobli et al., 2017). However, the therapeutic potential of many potent single domain DRPs are limited due to poor selectivity. For example, the analgesic potential of several voltage-gated sodium channel inhibitors is overshadowed by their effect on other physiologically crucial ion channels (Deuis et al., 2017; Zhang et al., 2018). Interestingly, there are several reports of naturally occurring multi-domain DRPs that display a multivalent mode-of-action (Bohlen et al., 2010; Chassagnon et al., 2017; Guyot et al., 2020; Van de Locht et al., 1995; Van de Locht et al., 1996). All of these characterized multi-domain DRPs contain а tandem repeat (TR) architecture, where the individual domains share high internal sequence homology. Previous bioinformatics studies of these TR-DRPs revealed that they belong to the larger molecular class that we have defined as secreted cysteine-rich repeat proteins (SCREPs) (Maxwell et al., 2018).

To date, three venom derived TR-DRPs have been characterized in detail; including two spider derived ion channel modulating toxins; DkTx ( $\tau$ -theraphotoxin-Hs1a; UniProtKB ID P0CH43) from Cyriopagopus schmidti (Bohlen et al., 2010) and  $\pi$ -hexatoxin-Hi1a (henceforth Hi1a; UniProtKB ID A0A1L1QJU3) from Hadronyche infensa (Chassagnon et al., 2017), and the serine protease inhibitor rhodniin (UniProtKB ID Q06684) from Rhodnius prolixus (Van de Locht et al., 1995). All three TR-DRPs use bivalency-simultaneously binding to two receptor sites-as a mechanism to enhance and prolong their pharmacological effects (Bohlen et al., 2010; Chassagnon et al., 2017; Van de Locht et al., 1995). The larger interaction interface observed in the bivalency of SCREPs (Bae et al., 2016; Gao et al., 2016) demonstrates their capacity for improved target selectivity compared to their single domain counterparts, such as the improved selectivity of DkTx compared with  $\tau$ -theraphotoxin-Pc1b UniProt P0C245) (Vanillotoxin-2; ID (Bohlen et al., 2010). This provides an opportunity to leverage existing knowledge of venom derived DRPs in the search for peptides with higher specificity toward therapeutic targets. Additionally, the relatively slow dissociation rates of bivalent DRPs make them ideal molecular probes for studying channel structure (Gao et al., 2016).

However, despite their attractiveness, there is currently no resource designed for mining or browsing SCREPs. Common databases dedicated to sequence repeats often focus on genomic DNA sequences (Boby et al., 2005; Gelfand et al., 2007; Le Fleche et al., 2001), or short amino acid repeats (Kalita et al., 2006). For example, PRDB (Jorda & Thierry, 2012) defines repeats as short periodic amino acid sequences that are directly adjacent to one another, while RepeatsDB (Paladin et al., 2017) uses structural information obtained from the Protein Data Bank to define protein repeats. Databases containing large numbers of DRPs such as Cono-Server (Kaas et al., 2010) and the Knottin database (Postic et al., 2018) are likely to contain examples of SCREPs, but they do not include any curation relating to peptide domain organization (architecture). Here, we present ScrepYard, an online database of SCREPs extracted using a refined and automated datamining pipeline. Open access to this database is provided to facilitate discovery and further investigation of SCREPs, extending the available resources for uncovering the underlying mechanisms that drive their fascinating multivalent activity.

#### 2 | DATABASE CONSTRUCTION

The following section will outline the construction of the ScrepYard database. This process is comprised of three distinct stages: SCREP datamining, SCREP architecture annotation, and the upload of data to ScrepYard (Figure 1).



**FIGURE 1** Flowchart outlining the construction of the ScrepYard database. The process can be divided into three stages, (1) SCREP datamining (green), (2) SCREP architecture annotation (purple), and (3) the compiling and upload of SCREP data to ScrepYard (orange). Key processes for each step shown on the right.

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#### 2.1 | Datamining

#### 2.1.1 | Generating the initial dataset

Three datasets are downloaded from UniProtKB(Bairoch et al., 2005) (1) those that are manually curated (Swiss-Prot subset) and non-Swiss-Prot sequences (UniProt-TrEMBL) that contain either (2) the annotation "signal peptide" or (3) have a subcellular annotation of "secreted". All three datasets are filtered to exclude sequences with subcellular location annotation of "intramembrane," "topological domain," and "transmembrane." The outputs of the three filters are merged and used as the initial dataset.

### 2.1.2 | Data refinement and SCREP extraction

The initial dataset is subsequently refined by applying a keyword filter to remove known non-SCREPs based on their annotations. Currently the keyword list consists of "intracellular" (Maxwell et al., 2018), "disulfide isomerase" (Wilkinson & Gilbert, 2004), "double CXXCH motif" (Chivers et al., 1996), "ferredoxin" (Schurmann & Buchanan, 2008), "sulfur" (Baghshani & Abadi, 2014; Lill, 2009), "zinc" (Brandt et al., 2009), "iron" (Lill, 2009), "cytochrome" (Meunier et al., 2004), "thioredoxin" (Arner & Holmgren, 2000), and "dehydrogenase" (Brandt et al., 2009). This heuristic approach allows for continual optimization of the pipeline with the addition of new keywords as these are identified and ongoing updates to the database.

After the keyword filter, SignalP [v-5.0 (Armenteros et al., 2019) is used to recognize and remove the signal peptide from each sequence in the dataset, generating mature protein sequences. In some cases, secreted proteins are sequenced from native material such as venom secretions, and do not contain a signal peptide region, for example, the spider toxin DkTx (UniProt ID P0CH43). Proteins that are not recognized by SignalP are directly grouped together with the mature sequences. Finally, the SCREP processing algorithm (SPA) is applied to remove all sequences that contain >500 or <20 amino acids (AAs) and sequences that contain <4 cysteine residues. The upper limit is set to avoid collecting much larger proteins, for example, transmembrane receptors, and the lower limit is set to avoid possible false identification of small non-domain repetitive elements. All remaining sequences are then processed to identify regions with internal sequence homology by use of an iterative BLAST function [see also Maxwell et al. (Maxwell et al., 2018)].

#### 2.2 | Architecture annotation

#### 2.2.1 | Generation of domain information

The dataset of extracted SCREP sequences requires further processing to accurately characterize each SCREP architecture including the specific domain types occurring in each SCREP, the order in which they appear, the sequence length of each domain, and the inter-domain linkers. The first step in SCREP annotation generates domain information. We utilize InterProScan (v-5.48), a consortium of several protein databases that predict domains using sequence-based recognition methods (Jones et al., 2014). As InterProScan consists of multiple databases, a single domain may be identified multiple times with slight differences in domain boundaries. To refine the InterProScan output data, the series of identified domains for each SCREP sequence is clustered by the database used, for example, Pfam, Prosite, and so on. In each cluster, identified regions are sorted according to their start and end positions. If an overlap exists between annotated domains, preference is given to the smallest recognized domain. The database-cluster with the highest number of recognized domains is then selected as the representative series of domain annotations for the SCREP candidate. If the number of domain annotations are identical, the database-cluster is selected according to a database preference list: Pfam (Mistry et al., 2021) > Prosite (Sigrist et al., 2010) > SMART (Letunic et al., 2021) > CDD (Lu et al., 2020) > SUPERFAMILY (Gough et al., 2001). After extracting the nonredundant domain annotations, each domain within a SCREP is numbered in sequential order according to its location from N- to C-terminus.

#### 2.2.2 | Defining TR architecture

For each SCREP, a series of internal BLAST functions (default parameter, e-value <10) are performed between all identified domains to determine interdomain sequence homology. Domains are defined as TR if a BLAST alignment can be found between neighboring domains (e-value <10) or are deemed nonhomologous (e-values >10) and defined as non-TR (nTR) domains. After defining the number of domains and whether they are TR of nTR domains, the SCREP architecture is annotated according to the sequential order of TR / nTR domains, (i.e., all three domain SCREPs may be annotated as TR1-TR2-TR3, TR1-TR2-nTR3, and nTR1-TR2-TR3) distinguishing between all possible combinations of TR and nTR domains. Finally, the sequence length of various SCREP elements including the N- and C-termini, the individual domains, and the interdomain linker regions are calculated based on the identified domain boundaries. In SCREPs containing more than one linker, that is, containing  $\geq$ 3 domains, each linker is sequentially numbered in the same way as the domains described above. Finally, we note that our approach to generate SCREP architecture annotation relies on the use of InterProScan, and in instances where ordered regions are not recognized by this tool, no annotations are produced in the ScrepYard output.

#### 2.3 | Data upload

#### 2.3.1 | SCREP database generation

To remove any duplicate SCREPs from ScrepYard, CD-HIT [v-4.8.1 (Li et al., 2001)] is used with a threshold of 0.999. CD-HIT is only applied to sequences that originate from TrEMBL (Boeckmann et al., 2003) as they have not been manually curated and may contain errors resulting in sequence duplication and fragmentation. All manually curated SCREPs that originate from SwissProt (Boutet et al., 2016) are maintained without applying CD-HIT. All SCREP domain annotations and other relevant information, such as taxonomy and cysteine content, is compiled, formatted, and uploaded to ScrepYard.

#### 2.3.2 | ScrepYard updates

The content in ScrepYard is automatically updated every 3 months. For each update, all newly released and recently modified sequences from UnitProtKB are processed. Any existing SCREPs that are found as new entries in the updated UniProtKB dataset are removed from ScrepYard and re-processed (this is to account for any slightly modified SCREP sequences). The newly processed data are then merged with the existing SCREPs database. Previous database iterations are archived on the Nectar Research Cloud (Barker et al., 2019) for 1 year, after which archived data are stored on a local server at the Centre for Advanced Imaging, University of Queensland, Australia.

The SCREP recognition process relies heavily on existing third-party software, including blast+, InterProScan, SignalP, and CD-HIT. To ensure the accuracy of SCREP datamining and annotation, we also perform software updates as required. After any software updates, the entire ScrepYard database is rebuilt.

## 3 | DATABASE UTILITY AND DISCUSSION

## 3.1 | Database content—SCREP architectures

In the latest update of ScrepYard (Dec 2022), 183,518 sequences were identified as putative SCREPs from the total secreted protein dataset (18,791,263 sequences)— three times as many SCREPs as the previous published extraction (May 2018), which comprised 60,935 putative

SCREPs from 8,006,061 secreted protein sequences (Maxwell et al., 2018). The growth in number of sequences shows the remarkably rapid expansion of available sequences within UniProtKB, further emphasizing the need for automated processing tools to extract sequences of interest.

The data can be broadly broken down into two categories based on their putative domain annotations, "InterProScan-identified" (49.3%) and "unknown architecture" (50.7%) (Figure 2a). The large proportion of unknown domains reflect the abundance of



**FIGURE 2** Distribution of SCREP architectures and linker analysis of two-domain SCREPs. (a) The inner circle demonstrates the two major clusters of SCREPs; "InterProScan-identified" (SCREPs with predicted domain types) and "unknown architectures" (SCREPs with unknown domain types). The outer circle demonstrates the different architecture types; unknown architectures (50.7%), pure domain repeats (PDR) (37.6%), and combinatorial domain repeats (CDR) (11.6%), dividing the PDR's and CDR's into a distribution based on the length of repeating domains. (b) The frequency distribution of linker lengths within the TR1-TR2 dataset. Most linkers are <20 AAs in length (85.57%), with the remaining linkers (14.43%) extending between 20 and 100 AAs. (c) A heat map of amino acid composition for all known two-domain SCREP for linker lengths between 1–20 AAs. AAs are sorted left-to-right in order of decreasing side hydrophobicity (Monera et al., 1995). (d) A grouped heatmap displaying the abundance of domain specific linker lengths in bacteria, fungi, plants, and metazoans. Each kingdom contains the four highest occurring domain types, with the frequency of each linker length displayed. The coloring indicates the relative level of abundance for each domain type

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uncharacterized domain types within UniProtKB, sometimes referred to as the "dark proteome" (Perdigao et al., 2015). Within the "unknown architecture" dataset we find a taxonomic bias, where 69.5% of prokaryotic SCREPs contain unrecognizable domains compared to just 39.1% of eukaryotic SCREPs (Figure S1). Given the uncertainties associated with the "unknown architecture" dataset, we have restricted the below analysis of the database to the "InterProScan-identified" dataset only.

The "InterProScan-identified" dataset can be further broken down into two groups based on their architecture. A pure domain repeat (PDR) is defined as having an architecture that contains only TR domains (e.g., TR1-TR2 or TR1-TR2-TR3), while combinatorial domain repeats (CDRs) represent all other architectures (e.g., nTR1-TR2-TR3 or TR1-TR2-nTR3; see also Figure 2a). Within PDRs, the TR1-TR2 architecture is most common, accounting for 24.9% of the entire database. CDRs make a smaller fraction of the database accounting for 11.7% of all sequences. For both PDRs and CDRs the most abundant architectures are those with the fewest number of domains, and in general there is a decrease in the number of SCREPs with a certain architecture, as a function of decreasing number of domains within that architecture (Figure S2).

The cysteine density (percentage of cysteine residues) within a SCREP is a defining feature of this class of molecules, and in general we would expect that a higher cysteine density would correlate with disulfide-directed, and hence thermodynamically more stable, folds. Overall, we find that SCREPs from eukaryotic kingdoms have a higher cysteine density (Figure S3), consistent with the more sophisticated disulfide processing machinery in these higher organisms (Gruber et al., 2006; Van Anken & Braakman, 2005). Given the central importance of this feature, we have made it possible to directly refine search results within ScrepYard by defining a minimum and maximum cysteine density. We note that while we have taken the inclusive approach of retaining any sequences within ScrepYard with a potential domain repeat that contains a single disulfide bond, this does not necessarily satisfy the requirement of "cysteine-rich". The cysteine-density filter, thus, allows the user to search or view a subset of the database (default values within the advanced search are >4 cysteines and >10% cysteine density).

#### 3.2 | Database content—Linker analysis

An important yet poorly characterized aspect of multivalency is how multiple domains are linked together, and what effect the "linker" region has on binding and function. The peptide linker is crucial in ensuring that each domain is positioned for optimal engagement with their molecular target (Bae et al., 2016; Bobrovnik, 2007; Klein et al., 2014; Mack et al., 2012). Elements of the linker, such as flexibility/rigidity and its effect on spatial positioning of the domains, play an important role in defining the intermolecular binding kinetics (Soler & Fortuna, 2017). Under evolutionary pressure, naturally occurring multivalent ligands have yielded linkers of a specific length that have a suitable amount of structural rigidity for enhanced target engagement (Bohlen et al., 2010; Chassagnon et al., 2017; Van de Locht et al., 1995; Van de Locht et al., 1996). These evolved linkers are consequently also likely to be dependent on the molecular target of the peptide. ScrepYard has been designed to be enriched in sequences that contain multivalent ligands. Analysis of linker sequences in ScrepYard may thus provide insights into the basic design principles that have emerged as a product of an evolutionary process in naturally occurring multivalent peptide ligands, thereby aiding rational engineering of synthetic multivalent peptides.

To further investigate the potential of ScrepYard to provide insights into linker properties of multivalent peptide ligands, we selected a subset of SCREPs with a twodomain TR architecture (Figure 2a). We subsequently filtered this subset to remove any sequences that contain a cysteine residue in the inter-domain linker sequence as this may indicate incorrectly defined domain boundaries and/or unrecognized domain regions (Figure 2b). Although there may be some genuine cases of cysteine containing linkers, to verify this requires individual assessment of existing experimental data. Our preliminary analysis of two-domain SCREPs from SwissProt (Boutet et al., 2016) with linkers  $\leq 20$  AAs that contain a cysteine, reveal a total of 40 SCREPs. One interesting example where four cysteine residues form two disulfide bonds within the linker was observed for the doubleantistasin like peptide (UniProt ID P15358) (Lapatto et al., 1997). This suggests that there may be a small population of SCREPs with functionally relevant cysteine residues within the linker region, warranting further analysis which extends beyond the scope of our current investigation. Additionally, SCREPs that are posttranslationally processed into two separate domains, via protease cleavage, were identified and removed from our analysis (2153 SCREPs in total). Sequences containing a dibasic site, that is, "KK," "KR," "RK," "RR," within the linker region indicate cleavage from subtilisin-like proprotein convertases (SPCs) (Rholam & Fahy, 2009). An example is human endothelin (UniProt ID P05305), which harbors two homologous endothelin-like domains. During posttranslation modification, the gene product is

cleaved by a Furin enzyme at a K91-R92 motif (between the two TR domains), yielding mature endothelin peptide and a second endothelin domain with unknown function (Turner & Murphy, 1996). This example highlights cases where TRs are posttranslationally cleaved to yield monovalent disulfide rich peptides, and as such contain linkers that do not contribute to multivalent binding.

Next, we restricted the data to proteins that had a cysteine density  $\geq$ 5%, to enrich for potential disulfidestabilized protein structures. This dataset is here simply referred to as the two-domain SCREPs. Within this dataset we found that the linker length (number of AAs) has an asymmetric Gaussian distribution with a maximum at approximately 10 AAs in length and the majority of peptides containing a linker between 1 and 20 AAs (85.57%) (Figure 2b). As the potential for the existence of an unrecognized domain within the linker increases with linker length (regardless of the presence of a cysteine), subsequent analysis of the amino acid composition (Figure 2c) and the distribution of linker lengths within various taxonomic groups and domain types (Figure 2d) was performed using a subset of peptides containing a linker of <20 AAs. The linkers of these two-domain SCREPs appear to consist primarily of amino acids with short or polar sidechains highly enriched in proline and alanine residues (Figure 2c). The observed linker composition of these SCREPs aligns with previous findings of AA occurrence within naturally occurring linker regions (Chen et al., 2013; George & Heringa, 2002).

The secondary structure prediction tool [MobiDBLite (Necci et al., 2017)] was then used to predict the presence of disorder within these linkers. Overall, we find that disordered linkers are more prevalent in SCREPs with a bacterial origin (12.3% of bacterial linkers compared with 0.2% of eukaryotic linkers) (Table S1). Although the exact functional purposes of these disordered linkers are unknown, their presence demonstrates natural variability of structural rigidity. We can only speculate that increased disorder would lead to lower avidity, or higher receptor promiscuity, which may reflect the differences observed between the prokaryotes and the more complex eukaryotic organisms.

Next, we investigated if there was a relationship between the linker length and the domain type. It is known that some DRP domain types are associated with specific functions (e.g., protease-inhibiting Kunitz domains). In these cases, if the second domain has evolved to bind to a common and adjacent receptor site, there may be evolutionary pressure to restrict the length and composition of the interdomain linker (Handl et al., 2007; Tran et al., 2020). In this analysis we find three general patterns, (1) domains with a broad distribution of linker lengths, (2) domains that have either short or longer linkers, or (3) domain types with a highly conserved linker length (with a sharp distribution, i.e., length  $\pm$  1 residue). Examples of the three types are as follows:

- 1. The fungal chitin domains and the metazoan ShKt domain types appear to have a broad distribution of linker lengths between 1–20 AAs.
- 2. The PsiF bacterial domains, and the metazoan Kunitz and WAP domains appear to favor shorter linker lengths (<12 AA's), while the Gnk2 plant domain has a cluster of linkers with a longer length (>14 AAs).
- 3. Domains with highly conserved linker lengths include; the CFEM fungal domain, the short (1–3 AA linkers) bacterial sel1-like repeats, the 2-residue linkers in Bowman-Birk plant domain, the 7-residue linkers in proteinase inhibitor plant domain (Prot.I.), and the 16-residue linkers in phospholipase inhibitor (Phos.I.) domain.

Examples of where a correlation between linker length and molecular target may exist can also be found in the bivalent serine-protease inhibitors rhodniin (a Kazal-type SCREP; UniProt ID Q06684) and ornithodorin (a peptide with two Kunitz domains with low sequence similarity; UniProt ID P56409). Despite their domains being structurally different, both rhodniin and ornithodorin bind to the same two regions of thrombin and have very similar linker lengths of 9 and 10 AAs, respectively (Van de Locht et al., 1995; Van de Locht et al., 1996). Therefore, we speculate that in some circumstances linker length may be indicative of molecular target (in this case more so than the 3D structure of the individual domains). Domain types with broad linkerlength distributions may indicate that these domains have undergone functional divergence, interacting with structurally diverse targets. Conversely, the highly conserved lengths such as that observed within the phospholipase inhibitor domain (Phos.I.), suggest interactions with either a limited number of molecular targets, or a family of targets with a high degree of structural similarity. Evidently, the elucidation of correlations between linker length and molecular target may serve as a powerful method in discovering novel multivalent ligands of known receptors.

#### 3.2.1 | Identifying bioactive SCREPs

In addition to annotating the SCREP architectures of natural multivalent peptides, ScrepYard has been devised to aid researchers to mine SCREP sequences to identify multivalent versions of their well-characterized singledomain counterparts. Our approach relies on the observation that the individual domains of two-domain bivalent SCREPs reported to date align well with existing single domain DRPs (Bohlen et al., 2010; Chassagnon et al., 2017). In addition, evidence suggests DRPs that target the same receptor tend to convergently evolve similar primary structures (Undheim et al., 2016), meaning that within a fold type, there is a high probability that a SCREP with a particular function shares a relatively high degree of sequence similarity with a single-domain DRP with the same function. For example, there is high sequence identity between the single-domain PcTx-1 isolated from the venom of the spider Psalmopoeus cambridgei (Escoubas et al., 2003) and the two-domain SCREP Hi1a isolated from the distantly related spider Hadronyche infensa (Chassagnon et al., 2017) (71% and 56% sequence identity with TR1 and TR2 of Hi1a, respectively). Both PcTx-1 and Hi1a have been confirmed to modulate the acid sensing ion channel 1a (ASIC1a) (Berkut et al., 2015; Chassagnon et al., 2017; Escoubas et al., 2003), with Hi1a exhibiting higher avidity than PcTx-1 due to a bivalent mode-of-action (Chassagnon et al., 2017). To apply this evolution-guided mining approach, we propose that the wealth of functional data available for single-domain DRPs [such as those curated in ToxProt (Jungo et al., 2012)] may serve as an ideal starting point to identify SCREPs with a putative multivalent mode-of-action.

As proof of principle, we employed a batch sequence analysis method aimed at identifying toxins with known activity that share sequence identity with SCREPs. A dataset of experimentally validated bioactive toxins was extracted from the ToxProt (Jungo et al., 2012) database, using this as a set of query sequences we performed a BLAST search between the known toxins and the SCREPs database. Using this method, we identified 9325 SCREPs which display varying degrees of sequence similarity with known single domain toxins. From these data, we selected the single-domain DRP Kalicludine-3, a dualfunction toxin isolated from the sea anemone Anemonia sulcata (UniProt ID Q9TWF8) that inhibits trypsin—a serine protease from the PA clan superfamily-and voltage-sensitive potassium channels (Schweitz et al., 1995). Kalicludine-3 was subsequently used as a query sequence to further demonstrate the utility of the ScrepYard BLAST search.

The output shows that Kalicludine-3 has high sequence homology with d-Gs1a; a putative double domain SCREP from the marine gastropod *Gemmula speciosa* (UniProt ID A0A098LW49) (Figure 3a). Thus, to determine if d-Gs1a shares the same bioactivity as Kalicludine-3, a d-Gs1a gene was synthesized and cloned into an *E. coli* expression vector for recombinant production (Figure S4). Following successful production, we used NMR spectroscopy to assess the folding of the peptide, and found a highly dispersed NH-fingerprint region,



**FIGURE 3** Sequence based identification, NMR confirmation of structural order and trypsin inhibition assay of d-Gs1a (A0A098LW49). (a) Alignment between Kalicludine-3 with each domain of d-Gs1a. Conserved residues between Kalicludine-3 and d-Gs1a are highlighted in red, while cysteines are highlighted in yellow. (b) 1D <sup>1</sup>H-NMR spectrum of d-Gs1a demonstrating well resolved and dispersed signal within the NH region, a characteristic feature of a well-defined globular fold. (c) Trypsin assay in the presence of d-Gs1a (0.1  $\mu$ M and 0.25  $\mu$ M) demonstrating inhibition of digestion of a trypsin substrate which fluoresces upon enzymatic cleavage (increased absorbance correlates with enzyme activity). All trypsin assays were performed in triplicate with 0.5  $\mu$ M trypsin.

consistent with a well-defined globular fold (Figure 3b). As Kalicludine-3 is a known serine protease inhibitor, a trypsin inhibition assay was performed to test the function of d-Gs1a. As suspected, we find that the recombinant d-Gs1a peptide shares activity with Kalicludine-3, showing potent trypsin inhibition in a concentration dependent manner, and being able to achieve full inhibition at sub-stoichiometric ratios (Figure 3c). However, screening for activity against a wide panel of voltage-gated potassium channels, TRPV1, and ASIC1a channels (Supplementary methods), revealed that d-Gs1a does not share the dual functionality of other, venom-derived, Kunitz-type peptides (Figure S5).

#### 4 | CONCLUSION

Naturally occurring multivalent peptides represent a valuable source of bioactive ligands, with a potential to be developed into novel biologics in the pharmaceutical and agrochemical industries. These molecules benefit from an evolutionary refinement process that offers unique insights into the underlying design principles of multivalency in peptides (Bohlen et al., 2010; Chassagnon et al., 2017). ScrepYard has been designed to be enriched for multivalent peptide ligands and provides researchers with the necessary tools to mine this resource using a variety of search and browse functions. To demonstrate the utility of this resource, we show how analyses of sequences within the database provide new insights into the significance of interdomain peptide sequences in defining peptide function. We further outline a targeted mining approach that enables the identification of novel SCREPs using the known sequence and bioactivity of previously studied receptor ligands. Using this approach, we identify a previously unknown two-domain protease inhibitor from the marine gastropod Gemmula speciosa. The construction and demonstrated utility of this resources promises to improve our understanding of multivalency while uncovering molecules of pharmaceutical and agricultural relevance.

#### AUTHOR CONTRIBUTIONS

Junyu Liu: Conceptualization (supporting); data curation (lead); formal analysis (lead); investigation (equal); methodology (supporting); writing – original draft (equal); writing – review and editing (equal). Michael J Maxwell: Conceptualization (supporting); formal analysis (supporting); investigation (equal); writing – original draft (equal); writing – review and editing (equal). Thom Cuddihy: Methodology (supporting); software (equal); writing – review and editing (supporting). Theo

Crawford: Investigation (equal); writing - original draft (supporting); writing - review and editing (equal). Madeleine Bassetti: Software (supporting); visualization (equal); writing - review and editing (supporting). Cameron Hyde: Methodology (supporting); software (equal); writing - review and editing (supporting). Steve Peigneur: Funding acquisition (supporting); investigation (equal); methodology (supporting); writing - review and editing (equal). Jan Tytgat: Funding acquisition (supporting); investigation (equal); methodology (supporting); writing - review and editing (equal). Eivind Undheim: Conceptualization (equal); formal analysis (supporting); funding acquisition (equal); project administration (equal); supervision (supporting); writing review and editing (equal). Mehdi Mobli: Conceptualization (equal); formal analysis (supporting); funding acquisition (equal); project administration (equal); supervision (lead); writing – original draft (supporting); writing - review and editing (equal).

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#### **CONFLICT OF INTEREST**

The authors have no conflicts of interest to declare.

#### DATA AVAILABILITY STATEMENT

The ScrepYard database is freely available at http://www. screpyard.org. The ScrepYard web application is independent and supports most browsers. The datasets generated and analysed during the current study are available in the SCREP repository at https://screpyard.org/database. 10 of 12 WILEY-

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#### SUPPORTING INFORMATION

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