

Research Publication Repository

http://publications.wehi.edu.au/search/SearchPublications

This is the author's version of the work. It is posted here by permission of the AAAS for personal use, not for redistribution.

Publication details:	Ribeiro AJM, Das S, Dawson N, Zaru R, Orchard S, Thornton JM, Orengo C, Zeqiraj E, Murphy JM, Eyers PA. Emerging concepts in pseudoenzyme classification, evolution, and signaling. <i>Science Signaling</i> . 2019 12(594): eaat9797
Published version is available at:	https://doi.10.1126/scisignal.aat9797

Changes introduced as a result of publishing processes such as copy-editing and formatting may not be reflected in this manuscript.

1 Emerging concepts in pseudoenzyme classification, evolution and signaling 2

- 3 António J. M. Ribeiro1, Sayoni Das2, Natalie Dawson2, Rossana Zaru1, Sandra Orchard1, Janet
- 4 T. Thornton₁, Christine Orengo₂, Elton Zeqiraj₃, James M. Murphy_{4,5} and Patrick A. Eyers_{6*}

European Molecular Biology Laboratory, European Bioinformatics Institute (EMBL-EBI),
 Wellcome Genome Campus, Hinxton, Cambridge CB10 1SD, UK

- 7 2 Structural and Molecular Biology, UCL, Gower St, London, WC1E 6BT
- 8 3 Astbury Centre for Structural Molecular Biology, Molecular and Cellular Biology, Faculty of
- 9 Biological Sciences, Astbury Building, Room 8.109, University of Leeds, Leeds, LS2 9JT, UK
- ⁴ Walter and Eliza Hall Institute of Medical Research, 1G Royal Parade, Parkville, Victoria
 3052, Australia
- ¹² ⁵ Department of Medical Biology, University of Melbourne, Parkville, Victoria 3052, Australia
- 13 6 Department of Biochemistry, Institute of Integrative Biology, University of Liverpool,
- 14 Liverpool, L69 7ZB, UK
- 15 * Corresponding Author: <u>patrick.eyers@liverpool.ac.uk</u>
- 16

17 **ONE SENTENCE SUMMARY:**

- 18 Introduction of a framework to support the rapidly expanding field of pseudoenzyme analysis
- 19

20 **GLOSS:**

Enzymes are protein biological catalysts that speed-up the rate of chemical reactions in all living organisms. Enzymes regulate both metabolism and cell signaling, and bioinformatics has revealed surprising variations within the enzyme catalytic machinery, which converts a significant percentage into pseudoenzymes that lack a catalytic function. In this resource, we review recent progress in the identification, classification and functional annotation of pseudoenzymes, revealing them as key regulators of many fundamental biological processes.

27

28 **ABSTRACT:**

The 21st century is witnessing an explosive surge in our understanding of pseudoenzyme-29 30 driven regulatory mechanisms in biology. Pseudoenzymes are proteins that have sequence 31 homology with enzyme families but which are proven or predicted to lack enzyme activity due 32 to mutations in otherwise-conserved catalytic amino acids. The best-studied pseudoenzymes are pseudokinases, although examples from other families are emerging at a rapid rate as 33 experimental approaches "catch-up" with an avalanche of freely-available informatics data. 34 Kingdom-wide analysis in prokaryotes, archaea and eukaryotes reveals that between 5-10% 35 of enzyme families contain pseudoenzymes, with significant expansions and contractions 36

37 seemingly associated with specific signaling niches. Pseudoenzymes can allosterically 38 activate canonical enzymes, act as scaffolds to control assembly of signalling complexes and 39 their localization, serve as molecular switches or regulate signalling networks through substrate or enzyme sequestration. Molecular analysis of pseudoenzymes is rapidly 40 41 advancing knowledge of how they perform their own non-catalytic functions, and permitting the discovery of surprising, and previously unappreciated, functions of their intensively-studied 42 enzyme counterparts. Excitingly, upon further examination, some pseudoenzymes have also 43 44 proven to possess novel enzyme activities that could not have been predicted a priori. 45 Pseudoenzymes can be targeted and manipulated by small molecules and therefore represent 46 new therapeutic targets (or anti-targets, where intervention should be avoided) in various diseases. In this review, which brings together broad bioinformatics and cell signaling 47 48 approaches in the field, we highlight a selection of new and recent findings relevant to a 49 contemporary understanding of pseudoenzyme-based biology.

50 **INTRODUCTION:**

51 Genomic sequencing and annotation and the subsequent mining of datasets from varied 52 organisms confirms that most well-characterized enzyme families encode pseudoenzyme homologs, which are predicted to be enzymatically-inactive due to the loss of at least one key 53 catalytic amino acid residue. This basic description, garnered from primary sequencing data, 54 has allowed the bioinformatic identification of pseudoenzyme genes (all of which are 55 transcribed as mRNAs and translated into proteins) in >20 different protein families [1], 56 including well-studied paradigms amongst the pseudokinases, pseudophosphatases and 57 58 pseudoproteases [2-10]. As detailed in Table 1, subtle changes in catalytic and substrate-59 binding sites have likely led to the appearance of pseudoenzymes from classical enzyme 60 templates, almost certainly following gene-duplication events. Not surprisingly, therefore, 61 pseudoenzymes share a similar overall protein fold when compared with catalytically-active enzymes from the same family [11, 12]. Of further interest, detailed comparison between 62 pseudoenzymes and related enzyme counterparts can also unearth physiological non-63 catalytic biological functions of enzymes, driven by the adoption of regulated inactive 64 65 conformation(s). Such additional functions are also broadly encapsulated in the related concept of protein "moonlighting" [13], where proteins are able to 'multi-task', by performing 66 different cellular functions as a consequence of their distinct environmental interactomes. 67 68 Although an absence of conserved catalytic residues in a pseudoenzyme does not unequivocally prove catalytic deficiency, very high sequence and/or structural conservation 69 70 suggests that pseudoenzyme sequences have been functionally selected across all the existing branches of life, and have been preserved to regulate specific aspects of cell biology 71 72 through catalytically-independent mechanisms. Despite the clear presence of 73 pseudoenzymes in a significant percentage of proteomes, we still understand very little about 74 their individual function, especially relative to their active enzyme counterparts. Two recent 75 international pseudoenzyme meetings, which were held in Liverpool, UK (2016) and Sardinia, 76 Italy (2018), served as complementary hubs to bring together expertise from across the 77 breadth of international biological communities, and to discuss and dissect how computational, theoretical and experimental data can be combined to make rapid progress in this new field. 78 79 Below, we review and evaluate and discuss some of these bioinformatic, structural and 80 biochemical approaches, which are rapidly revealing the extent, evolution and distribution of pseudoenzymes across the kingdoms of life. 81

82

84 Experimentally-established functional classes of pseudoenzymes

Pseudoenzymes are found amongst many metabolic and signaling classes of enzyme superfamilies (see Table 1 for an annotated selection). In terms of signaling outputs, pseudoenzyme actions can be rationalised based on four types of functional mechanisms: (i) regulating catalytic outputs of conventional (canonical) enzymes; (ii) acting as integrators of signaling events and/or toggling between signaling states as molecular switches; (iii) controlling assembly and localization of signaling hubs; or (iv), binding substrates/subunits to control the activity of conventional enzymes [14].

92 The first of these mechanisms, for which an increasing number of examples are available in 93 the protein kinase, phosphatase and ubiquitin signaling literature, all retain clear "enzyme-like" overall folds. These pseudoenzymes evolved the ability to regulate catalysis of a bona fide 94 enzyme-associated partner in order to generate a graded biological output in which the 95 96 pseudoenzyme and enzyme interaction remains the crucial controlling factor. The kinase-97 pseudokinase pair is perhaps the most well-known example from this pseudoenzyme class [15]. However, this recurring theme has recently been exemplified in the ubiquitin field, with 98 99 the analysis of the BRCC36 isopeptidase complex (BRISC) and the BRCA1-A complex, in 100 which the catalytically-active JAMM (JAB1/MPN/Mov34 metalloenzyme) domain Lys63-101 specific deubiquitinase (DUB) BRCC36 is physically partnered with the pseudo-DUBs Abraxas1 and Abraxas2 [16]. The BRISC complex permits cooperation between a newly 102 discovered "moonlighting" function of the metabolic enzyme serine hydroxymethyltransferase 103 104 2 (SHMT2), the pseudo-DUB Abraxas2 and three tandem pseudo-E2 (UEV) domains of 105 BRCC45 (Fig. 1). Complex assembly is regulated by SHMT2, which requires pyridoxal-5'phosphate (PLP) to form a tetramer that catalyzes its canonical role in nucleotide and amino 106 acid metabolism. Catalytically-inactive SHMT2 dimers, but not active, PLP-bound tetramers, 107 inhibit BRISC DUB activity, and intracellular PLP abundance controls this newly discovered 108 moonlighting SHMT2 function, which leads to the co-ordinated regulation of immune signaling 109 in cells [17]. 110

A second class of pseudoenzymes act as binary "switches" by integrating signals in the form of posttranslational modifications (PTMs) or by binding to metabolic ligands (perhaps ancient substrates), triggering inter-conversion between active and inactive conformations. Several of these are presented in more detail in Table 1.

The third category of pseudoenzymes possess distinct biological functions by operating as protein interaction scaffolds, generating (sub)cellular focal points that nucleate assembly of protein complexes, or regulate the localization of a binding partner. Examples are shown in Table 1, with the most profligate example to date being pre-mRNA-processing-splicing factor 119 8 (PRPF8), which contains four pseudoenzyme domains (Table 1) that act as scaffolds for 120 assembly of the enzymatic spliceosome complex [16, 18]. The further study of protein interactions made by pseudoenzyme domains ('pseudoenzyme interactomes') will provide 121 interesting information about the role of these proteins in the cell. Well-known examples 122 include the interaction of the STYX pseudophosphatase with components of the canonical 123 124 ERK pathway [19] and independent targeting to FBXW7, a subunit of the SCF Ubiquitin E3 ligase [20]. In a broad sense, molecular interaction data analysis can now be facilitated by 125 126 using the detailed curation model employed by the International Molecular Exchange (IMEx) 127 Consortium of molecular interaction databases [21], which accurately maps interaction data to specific binding-regions of proteins, and captures the effects of site-directed mutagenesis 128 129 on potential protein:protein interactions [22].

130 The final major pseudoenzyme category contains examples in which the protein has 131 repurposed canonical features of a fold that is shared with active enzyme relatives, probably 132 to act as competitors for either substrate binding (as catalytic "traps") or higher order complex assembly. A good example of this fourth category are pseudophosphatases, which usually 133 134 retain the ability to bind to "substrates", such as phosphorylated peptides, but are no longer 135 able to enzymatically process them, providing a trapping or localization mechanism in situ [5, 23]. The same might also be true of pseudoproteases [2], which possess reasonable binding 136 affinity for substrates (and can therefore control their fate), but do not cleave their substrates, 137 138 performing a catalytically-independent regulatory role. Outside of these notable pseudoenzyme families, relatively few examples of competitor and/or 139 "decov" 140 pseudoenzymes have been identified to date. However, new data confirm their expansion within plant, fungal and pathogen proteomes [24-27], including, for example, a new viral 141 142 pseudoenzyme that provides the pathogen with an advantage by directly competing with host cell defence mechanisms [28]. 143

Below, we present the fruits of the findings from the first two international meetings held between members of the pseudoenzyme community, including new state-of-the-art computational approaches that can exploit both genomic and proteomic (and in the future metabolomic) datasets. Together, these are revealing a vast number of pseudoenzymes embedded within genomes, and helping to create the first advanced analytical frameworks to probe and understand how pseudoenzymes function in a multitude of biological niches.

Bioinformatic resources for pseudoenzyme analysis: How computational scientists can identify and curate pseudoenzymes

152 The UniProt knowledgebase (<u>www.uniprot.org</u>) provides the scientific community with a 153 comprehensive, high quality and freely-accessible resource of protein sequence and

154 functional information for many species [29]. For enzymes, which represent between 20 and 155 40% of most proteomes, UniProtKB provides additional information about Enzyme Classification (EC), catalytic activity, cofactors, enzyme regulation, kinetics and pathways, all 156 based on critical assessment of published experimental data. Bioinformatic and structural data 157 are also used to enrich the annotation of the sequence with the identification of active sites 158 159 and binding sites. Whereas the annotation of enzymes is very well defined, the curation of pseudoenzymes has proven much more challenging. The main issue resides in their 160 identification, especially when the only evidence usually available to define a pseudoenzyme, 161 162 is the *absence* of critical active site residues. Indeed, to infer a lack of catalytic activity based solely on sequence analysis can be misleading. Experimental evidence can also be difficult to 163 interpret, especially when it contradicts sequence analysis-based prediction and catalysis may 164 be present or absent among orthologs. During the curation process, all the available evidence 165 must be manually assessed to decide whether the protein could be considered a 166 pseudoenzyme or not (Fig. 2A). Another important challenge is how to translate this 167 168 information into consistent, meaningful annotation to facilitate the subsequent retrieval of 169 pseudoenzymes by users. Following the two international pseudoenzyme conferences held in 170 2016 (Liverpool, UK) and 2018 (Sardinia, Italy), the annotation of pseudoenzymes in UniProt 171 underwent a revision to reflect advances in the pseudoenzyme field. For example, the UniProt 172 curators improved the usage of protein names to directly reflect their inactive status. In addition to providing the enzyme family to which they are related, as well as the position of the 173 "catalytically inactive" domain, UniProt explains the reason why a protein is considered a 174 175 pseudoenzyme in a "Caution" comment in the enzyme function section (Fig. 2B). Importantly, 176 the source of the evidence used to infer the lack of catalytic activity-experimental, sequence analysis or orthology-based—has also been added to make pseudoenzyme definitions as 177 178 evidence-based as currently practicable (see Table 1).

179 As an experimental test-bed for the curation and annotation of procedures to create pseudoenzyme databases, UniProtKB have recently reviewed and updated the annotation of 180 the complete kinome [30] and phosphatome of the model worm C. elegans, which contain 181 438 kinases and 237 phosphatases, respectively. Among 208 kinase genes whose function 182 has been experimentally characterized, 41 are annotated as pseudokinases, although 183 184 supporting evidence is often lacking beyond predicted sequence information. Indeed, for > 185 90% of worm pseudokinases, a lack of catalytic activity is based only on sequence analysis 186 (Fig. 3A). For genes that lack functional characterization, the sequence analysis predicted lack 187 of kinase activity can be supported by experimental evidence in orthologs. However, this is not always possible, especially for families that contains mainly *C.elegans* specific genes. This 188 is the case for the two most abundant kinase families, CK1 and TK. For the CK1 family, 11 of 189 the 83 members (13%) and for the TK-KIN-16 group 4 of the 15 members (26%) are predicted 190

191 to be inactive based on the loss of the active site. The TK-Fer group is predicted to contain 192 only active members. A similar situation is found in the C.elegans phosphatome. Of the 237 phosphatases, 8 are annotated as "inactive" pseudoenzymes, although only one 193 pseudophosphatase possesses documented experimental evidence supporting the facile 194 195 informatics-based prediction of catalytic sterility (Fig. 3B). These two examples illustrate the 196 need to improve experimental characterization of pseudoenzymes, as inferences based only 197 on sequence analysis could result in the incorrect attribution of inactivity, as the well-known 198 example of the with-no-lysine (WNK1; UniProtKB Q9JIH7) family of serine/threonine-protein 199 kinases ably demonstrates [31, 32]. Based on sequence analysis, WNK enzymes would be predicted to be inactive; however, under various experimental conditions tested, they are 200 201 clearly catalytically active [6]. The WNK members demonstrate the importance of biochemical 202 and structural analysis in understanding enzymes and pseudoenzymes in enzyme superfamilies, such as phosphotransferases; in the case of the WNK "pseudokinases", this 203 204 has led to their more accurate re-classification to atypical kinases.

205 Perhaps the most difficult challenge facing the curation field is the vast amount of protein 206 sequences that continue to be made available through genome sequencing projects. For 207 example, in just one month between UniProt releases in October and November of 2018, over 208 3 million new protein sequences were added to the database. In order to provide users with 209 information on these proteins (without experimental evidence in nearly all cases), automatic annotation rule-based systems have been devised to enrich the annotation of protein 210 sequences based on protein family membership. These identification systems, based on 211 related manually curated protein entries, provide users with basic predicted functional 212 213 annotation; an example being UniProtKB H9J2B7, a predicted tyrosine-protein kinase receptor based on automatic sequence detection analysis. Despite the development of rules and tools 214 aimed at identifying differences and similarities in natural variation, such rules may not 215 recognize pseudoenzymes even within a closely related enzyme family. 216

217

218

219 Evolutionary biology as a driver in the pseudokinase field

Whereas several workflows exist for annotating enzymes, the "rules" for automatic annotation of pseudoenzymes are currently lacking. Focused progress needs to be made on this issue across multiple different enzyme families in order to fully appreciate pseudoenzyme diversity and potential biological ubiquity. Reliable identification of catalytic site residues that can be used to predict a loss of enzymatic activity remains difficult. However, increased experimental characterization of pseudoenzymes in the last decade, and new evolutionary analyses of

226 catalytic sites and catalytic domains, continues to provide invaluable information that improves 227 manual curation and expands the automatic identification/annotation of pseudoenzymes. A 228 case-in-point are the pseudokinases, for which a kinome-wide database has now been assembled across all known eukaryotic, bacterial, and archaeal proteomes [24]. This analysis, 229 230 published in Science Signaling, has revealed that pseudokinases are present across all three 231 domains of life; in total, approximately 30,000 eukaryotic, 1,500 bacterial, and 20 archaeal 232 pseudokinase sequences were classified into 86 distinct pseudokinase families, including ~30 233 well conserved pseudokinase families that were not previously reported [24]. The rich diversity 234 of pseudokinases that occurs across the kingdoms of life exhibits notable family-specific expansions in animals, plants, fungi, and bacteria, where pseudokinases (and 235 pseudoenzymes in general) had previously received cursory attention. Pseudokinase 236 expansions are often accompanied by domain shuffling, which appears to have promoted new 237 roles in plant innate immunity, modulation of plant-fungal interactions, and bacterial signaling. 238 Mechanistically, the ancestral kinase fold, an ideal template for the generation of new 239 functions in pseudoenzymes, has diverged in many distinct ways through the enrichment of 240 241 unique signature sequence motifs, generating in-turn a slew of new pseudokinase families. 242 The catalytic kinase domain is also repurposed for non-canonical nucleotide binding or 243 atypical catalysis, or to stabilize unique, catalytically-inactive kinase conformations associated 244 with catalytically-independent types of signaling (see below). To conveniently compare these complex datasets, an annotated, searchable collection of all known predicted pseudokinase 245 sequences, and their evolutionary relationships, has been captured in the freely available 246 247 protein kinase ontology (ProKinO) [24, 331 248 (http://vulcan.cs.uga.edu/prokino/hierarchy/ProteinPseudokinaseDomain).

250 Structural biology as a key driver in the pseudokinase and pseudoenzyme fields

Structural studies continue to play a pivotal role in advancing our knowledge of functional 251 252 evolution among enzyme superfamilies, most notably the kinase superfamily, which is composed of a very broad range of small molecule, antibiotic, glycan and protein kinases [34]. 253 254 Notable examples of these important pseudoenzymes are discussed here briefly, beginning 255 with the remarkable finding that nucleotides can bind in different modes in predicted 256 pseudoenzymes, in some cases leading to unusual (and completely unexpected) types of 257 catalysis. For example, an inverted conformation of ATP was observed in the pseudokinase 258 FAM20A (Fig. 4A) [35, 36], which is a catalytically-inactive pseudoenzyme regulator of the conventional protein kinase FAM20C [37]. However, a structure of SelO revealed a similar 259 atypical ATP-binding mode, but rather than being catalytically-inactive, SelO has evolved the 260 261 capacity to "AMPylate", rather than phosphorylate, a broad spectrum of protein substrates (Fig. 4B) [38]. Additionally, several (pseudo)kinases have been confirmed to operate as 262 phosphorylate sugar (as opposed to amino acid) residues, including FAM20B [37, 39] (Fig. 263 4C) and SgK196/Protein O-mannose kinase (POMK) [40] (Fig. 4D). Intriguingly, the bacterial 264 pseudokinase SidJ has recently been shown to catalyse protein glutamylation of the SidE 265 family of ubiquitin E3 ligases, inhibiting their catalytic output [41, 42]. 266

Although not classically considered to be a specific catalytic motif, the variable glycine-rich 267 loop sequence in kinases (Gly-X-Gly-X-X-Gly) serves as a flap-like structural feature to 268 269 promote and lock ATP-binding. Interestingly, this motif has been found to be dispensable for 270 catalytic activity in an atypical catalytically-active coccidian kinase, which has been termed 271 WNG1 (with no glycine-1) [43]. A recent study suggests a catalytically active, druggable 272 conformation in Drosophila BubR1 that can directly phosphorylate the motor protein CENP-E [44]. This likely differentiates it from human BubR1, which is reported to be an inactive 273 pseudokinase [45] lacking a Gly-rich loop and that not bind detectably to ATP [46]. Although 274 275 it is more common that proteins annotated as pseudokinases based on the absence of the 276 traditional catalytic residues (which do not, by definition, include the Gly-rich loop) are indeed catalytically-defective when scrutinized at the appropriate biochemical level, these examples 277 underscore the versatility of the kinase fold to evolve diverse functions, especially when 278 279 released from the evolutionary pressure of maintaining a catalytically-competent fold [47]. 280 More broadly among pseudoenzymes, such functions likely speak to a non-catalytic substrate binding to an ancestral enzyme, which has evolved to perform a catalytic function, as has 281 been demonstrated during in vitro evolution experiments [48, 49] and by using structural 282 features to evolve canonical activity in the pseudokinase CASK, where 4 amino acid 283 284 substitutions are required to regenerate a Mg-ATP dependent enzyme [50]. In a similar vein, 285 improving the curation of (non-kinase) pseudoenzyme families will provide the scientific community with valuable information to understand the evolution of these proteins, the aetiology of related diseases and the development and repurposing of pseudoenzymetargeted drugs, a useful bonus in carefully-conducted pseudoenzyme studies [7, 51, 52].

289

A new approach for the identification of pseudoenzymes using UniProt

To help take the pseudoenzyme field forwards, we have designed and tested a simple 291 292 computational pipeline to identify and annotate pseudoenzymes using sequence alignments, UniProt annotations assembled from the primary literature (The UniProt Consortium, 2017), 293 and information from the Mechanism and Catalytic Site Atlas (M-CSA), a database that 294 currently contains defined catalytic residues and mechanistic data for 964 enzymes [53]. We 295 296 have limited our current analysis to those enzymes for which we have a good or detailed 297 knowledge of their catalytic mechanism (such as those in M-CSA) and therefore know the 298 specific residues involved. To identify pseudoenzymes amongst these data, we start by finding 299 all the SwissProt sequences that are homologous to entries in M-CSA, by using phmmer with 300 an E-value cut-off of 10-10 [54]. This value is reasonably stringent and will act as a filter to include only close homologues. More distant relatives are better identified using the structure-301 based approaches developed by Orengo and colleagues, as described below and in related 302 303 work [12, 55]. Our procedure yields a broad collection of sequences from all the domains of life, although the sample can be biased by the (current) uneven representation of sequences 304 in SwissProt and enzymes in M-CSA. After their identification, we categorize each homologue 305 as enzyme or non-enzyme according to its annotation in SwissProt. There are at least three 306 307 types of annotation in SwissProt that we can use to identify enzymes: EC numbers, UniProt keywords, and GO terms. These annotations provide us with three possible rules. A sequence 308 can be categorized as an enzyme if it (i) has at least one EC number; (ii) is annotated with at 309 least one catalytic UniProt keyword (Oxidoreductase, Transferase, Hydrolase, Lyase, 310 Isomerase, Ligase, Translocase, or their hierarchical children); or (iii) is annotated with a 311 312 catalytic GO term ("catalytic activity" and its child terms). Unfortunately, the three rules are not 313 currently entirely consistent. For example, there are approximately 3,000 sequences with a 314 catalytic keyword but no EC number, and about 35,000 sequences possess a catalytic GO term but have no catalytic keyword. These differences may be due to out-of-date annotation, 315 316 no available EC number in the EC classification, or GO terms that are automatically transferred 317 based on homology, which will label pseudoenzymes as enzymes. Overall, the rule (ii) based on the UniProt keywords seems to be the most comprehensive without using extended 318 annotation based uniquely on homology, and so we have applied it for the purpose of this 319 review. There are no pseudoenzymes in 669 of the enzyme families curated in M-CSA. 320 321 However, in 237 of the enzyme families where pseudoenzymes are found, these account for less than 10% of the sequences in the family. The number of families where pseudoenzymes 322

are more common than 10% is 88 (the sum of the last 9 columns in Figure 5). However, the percentage of families identified as including pseudoenzymes is currently a *minimum* number - more can and will be found with deeper searches. Below, we highlight how we can use data from individual enzyme families to try to understand how pseudoenzymes have evolved, an important central question in evolutionary biology.

329 Evolution of pseudoenzymes

330 Knowing the number of pseudoenzymes associated with each M-CSA enzyme entry is an 331 important starting point for understanding how these proteins came to "lose" (or repurpose) their catalytic function, assuming that they evolved from canonical enzymes in the first place. 332 The reconstruction of their evolutionary trees is necessary to tackle some of the fundamental 333 questions that the field would need to answer. These include (i) how common are the 334 evolutionary 'jumps' that either transform enzymes into pseudoenzymes or pseudoenzymes 335 into enzymes?; (ii) which of these transformations is more common?; and (iii) can we identify 336 the conserved mutation(s) in the catalytic residues that drive these changes? Below, we have 337 attempted to answer the last question by generating and annotating phylogenetic trees for all 338 the enzymes in M-CSA. In the phylogenetic tree for β -amylase (**Fig. 6**), which contains two 339 pseudoenzymes, the homologues belong mostly to plants, except for five that possess a 340 341 bacterial origin. The red and green circle on the tree represents the point where one 342 pseudoenzyme diverged from an ancestral enzyme. The two existing pseudoenzymes evolved from this pseudoenzyme ancestor, so there is only one "loss-of-function" event 343 344 represented in the tree. The green circle at the base of the tree means that the last common 345 ancestor for all the homologues in the tree was most likely an enzyme, consistent with the directionality of evolution from enzyme to pseudoenzyme. The pseudoenzymes, which are 346 both classified as "inactive" in UniProtKB and have no associated EC numbers, have the most 347 348 mutations in their catalytic residues of all the homologues identified. Furthermore, the 349 pseudoenzymes are the only homologues in the tree with substitutions replacing the catalytic 350 residue, Glu₃₈₁. This acidic side-chain acts as the general base that activates the nucleophilic water molecule to a reactive hydroxide ion, while other residues such as Thr₃₄₃, Leu₃₈₄, and 351 Asp102 have stabilization roles that are not critical for the reaction, and so can (and are) readily 352 replaced during evolution without affecting enzyme function. It is currently impossible to say if 353 these mutations were the original cause for the loss of function, or if they happened after the 354 355 loss of function, which may have occurred through other means such as simple accumulation of mutations without corrective selective pressure. After the loss-of-function event, a lack of 356 corrective selection pressure will lead to the accumulation of mutations in the catalytic 357 358 residues. For this enzyme set, the mutation of catalytic residues alone does not lead 359 necessarily to loss-of-function, as some of the other "active β-amylases have catalytic 360 mutations. Gain-of-function may also be generated through means other than catalytic residue 361 mutations, as illustrated by the β/α -amylase protein (P21543), which has an additional domain 362 with the α -amylase activity (Fig. 6, dark blue).

363

Use of GO terminology to estimate the global extent of pseudoenzymes

365 There are currently a large number of entries in UniProt (about 35,000) that do not have a catalytic keyword but do possess a catalytic GO term. This inconsistency is brought about by 366 GO annotation that is automatically extended from enzymes to their homologues, regardless 367 368 of the presence or absence of actual catalytic activity. Consequently, many of the sequences annotated with a catalytic GO term are themselves pseudoenzymes. To illustrate this 369 370 contradiction, we also performed an analysis where we removed all the sequences where annotation is contradictory. This led to a smaller total number of pseudoenzymes identified 371 372 within the M-CSA entries homologues set, as would be expected (fig. S1), and the number of 373 enzyme families with pseudoenzyme members drops from 325 (32.7% of all the families) to 132 (13.3%). Clearly, this type of analysis remains rather subjective unless backed up with 374 supporting data, and represents one of the major experimental challenges in the 375 pseudoenzyme field, where biochemical and cellular analysis are often complex, time-376 377 consuming and costly.

378

381

Identifying pseudoenzymes in proteomic databases by exploiting the Class, Architecture, Topology, Homology (CATH) resource

Pseudoenzymes have now been identified in many major enzyme families across the tree of 382 life; predictably, this has been predominantly through computational sequence-based 383 384 analyses [30, 56] (Table 1). Most studies have traditionally compared the sequence of a 385 relative of unknown structure and functional residues against sequences of relatives, which 386 have been structurally characterized and annotated with known catalytic residues that have 387 been confirmed experimentally [14]. However, one study has indicated that homologous 388 proteins that share a common core domain structure can often acquire new enzyme functions by changes both in the nature of their catalytic residues and in the absolute position of these 389 catalytic residues in the protein scaffold [55]. As a result, it can be very difficult to provide 390 391 confident predictions of pseudoenzymes by only looking at deviations from known catalytic sites, most notably if this is done so in the absence of experimental characterization and/or 392 phenotypic or mutational information. More complex approaches are therefore deemed 393 394 suitable wherever possible.

As an extension of the approach described above using enzyme terminology, we have also systematically investigated the distribution of pseudoenzymes in the protein universe, using protein families from the CATH-Gene3D resource [57, 58], which links protein domain sequences to structures and experimental functions. CATH-Gene3D classifies ~460,000 domain structures and ~95 million protein domain sequences into ~6,200 evolutionary superfamilies [57]. These can then be sub-classified into Functional Families (FunFams) that share highly similar structures and functions, based on sequence patterns – specificity402 determining positions (SDPs) and other conserved positions (Fig. 7A) [59]. Although it is 403 difficult to achieve complete separation in some extremely diverse superfamilies, the accuracy of these functional family classifications has been validated by comparison against 404 experimental data and by endorsement through blind independent assessment in the Critical 405 406 Assessment of protein Function Annotation (CAFA) algorithm-based evaluation of functional 407 annotations [60]. The number of functional families reflects the functional diversity of a particular superfamily and can be used to explore how protein function is modulated in diverse 408 409 superfamilies.

410 We examined 383 enzyme superfamilies in CATH/Gene 3D v.4.2 that contain well-known (experimentally-validated) catalytic domains [53, 55], and identified the proportion of functional 411 families that have enzyme annotations and compared them to those that lack any enzyme 412 annotation. These are highly populated superfamilies accounting for 64% of sequences in all 413 CATH enzyme superfamilies and 60% of all sequences in CATH. A functional family was 414 415 considered to have enzyme annotations if it has at least one relative that has an EC annotation 416 in UniProtKB [29] and an experimental Gene Ontology (GO) [61] annotation for 'catalytic 417 activity'. For a third (131) of these enzyme superfamilies, all functional families were annotated 418 as enzymes (Fig. 7B). However, approximately 252 enzyme superfamilies (two-thirds) had 419 varying proportions of functional families that had no enzyme annotations in the EC 420 classification or GO (Fig. 7C), suggesting that these are very likely to be pseudoenzymes. To collate this information in a searchable manner, we have created the first on-line list of putative 421 422 pseudoenzyme superfamilies (https://uclorengogroup.github.io/cathpseudoenzymes/index.html). In order to explore some of these pseudoenzyme-containing 423 424 families, we also developed a protocol that first generates structure-guided multiple sequence alignments of multiple functional families within a superfamily that share very similar structures 425 (such as, wherein relatives superimpose with <RMSD 5Å), and are therefore likely to share 426 catalytic mechanisms. For these structural clusters, grouping structurally similar relatives from 427 different functional families, we next examined whether there was at least one functional family 428 429 within them with known catalytic residues [53], and at least one family with no enzyme 430 annotation. A functional family lacking enzyme annotation was identified as a putative 431 pseudoenzyme family if we observed a loss or significant change in known catalytic site 432 residues (identified in the enzyme family) in all relatives of the putative pseudoenzyme family, 433 determined using the comprehensive structural cluster alignment. In other words, the putative 434 pseudoenzyme family has conserved residues in the active site that differ significantly in their 435 physicochemical nature or 3-dimensional location to the known catalytic residues of the enzyme family. Clearly, the final confirmation of the sequence as a pseudoenzyme can only 436 come from exhaustive experimental testing, as is now becoming much more commonplace in 437 major signaling superfamilies, such as the phosphatases and pseudophosphatases, kinases 438

439	and pseudokinases, and proteases and pseudoproteases [5, 14]. Finally, analysis of some of
440	these structural clusters in superfamilies enabled us to identify families containing previously
441	reported pseudoenzymes, such as the uclear ransfer actor 2 (NTF2) and calsequestrin [62].
442	For example, the CATH superfamily 3.10.450.50 contains ~38% of functional families with
443	enzyme annotations, whereas 62% of the families do not have any enzyme annotations. The
444	distinct functional family containing NTF2 is structurally very similar (< 4Å RMSD, same
445	structural cluster) to the enzyme families containing scytalone hydratases and steroid delta-
446	isomerases but lacks the catalytic machinery of those enzyme relatives (Fig. 8). Another
447	interesting example is N-Myc downstream-regulated gene 2 (NDRG2), which is structurally
448	similar to the enzyme families in the α/β -hydrolase superfamily (CATH 3.40.50.1820) and is
449	thought to be a tumor suppressor [63].
450	
451	
452	
453	
454	
455	
456	
457	

460 The future of pseudoenzyme research

The evolutionary conservation and prevalence of genes that encode pseudoenzymes is now 461 abundantly clear in the natural world, from well-studied model organisms to newly sequenced 462 genomes from previously uncharacterized species. However, one major barrier to accurate 463 464 pseudoenzyme cataloguing prior to experimental triage is the broad diversity of terminologies 465 previously used in their literature description, differing definitions of what constitutes a bona fide pseudoenzyme, and an historical prejudice that proteins lacking catalytic activity are of 466 unlikely to serve biologically-important functions. Over the last 20 years, various studies have 467 referred to pseudoenzymes as "non-enzymes", "prozymes", "dead" or "catalytically-468 defective/inactive" or, even more confusingly, "atypical" or "non-canonical" enzymes. We 469 therefore propose the broad adoption of "pseudoenzyme" as the essential descriptor that will 470 471 behave as a critical rate-determining step in allowing the direction of this relatively new field 472 to be strategically plotted and accurately delivered. Formally, we propose that pseudoenzymes be defined as "the predicted catalytically-defective counterparts of enzymes 473 owing to an absence of one or more catalytic residue". It should come as no surprise that in 474 475 some cases in which this simple description is applied, residual catalytic activities have been reported amongst proteins defined as pseudoenzymes (mainly pseudokinases), including the 476 isolated HER3/EGFR3, JAK2 (JH2), CASK and TRIB2 pseudokinase domains [7, 50, 64, 65]. 477 478 However, because these very modest catalytic activities are either dispensable for biological 479 function, or not conserved in vitro among paralogous proteins across species (such as 480 HER3/EGFR3[2) [24], this vestigial (or residual) catalytic activity is probably not a defining 481 feature of their biological function. Technically, establishing and guantifying such activities 482 remains enormously challenging, because ab initio prediction of substrates and co-factors is not facile and even trace co-purifying/contaminating proteins in recombinant pseudoenzyme 483 preparations (which may lack auxiliary endogenous factors) can lead to erroneous attributions 484 485 of catalytic functions.

486

Other cases discussed in this review pose a more serious definitional challenge: diverse 487 proteins predicted to be 'catalytically-dead' pseudokinases, like SelO, SidJ, and 488 SgK196/POMK, have instead evolved distinct, and guantifiable, catalytic functions, illustrating 489 a potential weakness of ab initio prediction of protein function based on a comparative 490 assessment of sequence. However, these important findings serve to highlight the versatility 491 of protein domains as ancestral folds for the creation of newly evolved activities. Moreover, 492 studying these "odd" folds, and various adaptions at the amino acid level, can potentially lead 493 to paradigm-shifting discoveries, including the three examples listed above. Rather than being 494 495 defined solely as pseudoenzymes by virtue of the unconventional positions of their catalytic 496 residues, such catalytic proteins can more accurately be considered as enzymes that exhibit atypical or novel catalytic mechanisms. Historically, this is best exemplified by WNK1, which 497 guided by structural studies was revealed to contain a compensatory ATP-positioning lysine 498 499 in the β 2 rather than the β 3 strand found in the vast majority of protein kinases [32]. The discovery of catalytically active WNG1, one of several secreted Toxoplasma atypical kinases 500 501 that together form a clade (termed ROP33, 34 and 35) comes as an additional pleasant 502 surprise [43], especially given that the related pseudokinase Bradyzoite pseudokinase 1 503 (BPK1), which possesses no detectable nucleotide binding in vitro [46], was previously shown to be involved in both the development and infectivity of *Toxoplasma* cysts [66]. The discovery 504 505 of non Gly-rich loop-containing atypical protein kinases adds an interesting new twist to the 506 WNK1 paradigm, because it has generally been assumed that the **complete** absence of a 507 Gly-rich loop would preclude ATP binding and catalysis amongst phosphotransferases [46, 67-69]. However, it is now abundantly clear from these, and other, examples that a detailed 508 509 knowledge of protein structure aligned with robust biochemical assays continues to be 510 essential for understanding and validating the novel catalytic or pseudoenzyme-based mechanisms that are at play. 511

512

513 A rich diversity of pseudoenzymes in genomes revealed by bioinformatics

Databases such as UniProtKB, and the biocurators that populate new and existing entries with 514 information extracted from the biomedical literature, play a critical role in cataloguing 515 experimentally-evaluated pseudoenzymes. These characterised proteins, however, represent 516 only a small percentage of the pseudoenzymes believed to be present in even well-studied 517 model genetic organisms. Moreover, their identification is critical in enabling the recognition 518 of pseudoenzymes in all proteomes, and for the development of more sophisticated 519 520 recognition criteria, which should allow these proteins to be distinguished from catalytically active family members. While bioinformatic studies can readily predict a lack of conventional 521 522 activities based on founder members of a particular enzyme class or mechanism, they cannot 523 exclude the acquisition of novel enzymatic functions or predict the biological nature of the non-524 catalytic functions a pseudoenzyme might perform, which is the major reason to study them. Nonetheless, the evolution of non-catalytic functions by pseudoenzymes provides a window 525 526 into formerly unrecognized regulatory functions mediated by conventional enzymes, such as 527 ATP-binding site occupation in the absence of catalysis in PKC [70], and the stabilization of NMYC by catalytically-inactive Aurora A [71]. While this "moonlighting" allows enzymes to 528 perform additional functions beyond their first-characterized catalytic function(s), it is likely that 529 530 gene duplications have allowed pseudoenzymes to evolve specific functions, often within the same pathway as the parental enzyme. Duplication allows a parental enzyme to perform an 531 essential function, since it relieves the selective pressures on the duplicated enzyme that 532

533 would otherwise constrain active site geometries for catalysis and substrate recognition. Divergence arising from relief of such selective pressure is very well illustrated by structural 534 plasticity among pseudokinase domains. For example, the pseudokinase domains of mouse 535 and human MLKL exhibit divergence in the structures of their "pseudoactive" site clefts and 536 the relative contributions of pseudoactive site residues to ATP binding, which is a relatively 537 538 common feature of many catalytically-inactive pseudokinases [47, 72]. Additionally, the 539 positions of one of the core elements in active protein kinases, the αC helix, whose Glu positions the ATP-binding lysine in β 3 of the N-lobe for catalysis, is highly variable among both 540 kinase and pseudokinase domains. Indeed, in some cases, such as the pseudokinase 541 Sqk223, the α C helix is absent in crystal structures [73] and/or has evolved alternative 542 functions, such as a serving as a regulated platform for peptide-binding [74, 75] amongst the 543 Tribbles pseudokinases [76], which can themselves be targeted with small molecules 544 originally developed as inhibitors of canonical kinases [52, 77]. 545

546 We believe that bioinformatics approaches, including those under development discussed 547 here, will permit the identification of many new pseudoenzymes amongst the large number of 548 enzyme families in annotated proteomes. Initially, this will be by leveraging our knowledge of 549 protein structure and catalytic mechanisms among active enzyme counterparts in order to 550 predict deficiencies in catalytic residues within their pseudoenzyme cousins. Furthermore, the 551 continued massive expansion of the sequence repositories, particularly with many hundreds of millions of novel enzyme sequences coming from metagenome studies, combined with 552 increasing protein structure data, will enhance the power of those bioinformatic methods that 553 554 analyse the presence or absence of highly sequence-conserved residue positions in experimentally uncharacterised functional families. Expanding the size and diversity of family 555 membership gives clearer, more accurate sequence patterns. These can then be more easily 556 compared against sequence patterns of known enzyme families in the same superfamily in 557 order to more accurately predict loss of essential catalytic machinery. In addition, they might 558 also identify novel (and experimentally testable) catalytic machinery, spatially located in the 559 560 active site pocket that is potentially indicative of a shift or change in signaling function. Such 561 approaches will facilitate broad mining of protein sequences to identify candidate pseudoenzymes, but will ultimately rely on a combination of structural and biochemical 562 studies, ideally including the mapping of enzyme <-> pseudoenzyme evolutionary trajectories, 563 564 to formally evaluate catalytic-deficiency and/or the acquisition of non-canonical catalytic 565 mechanisms.

567 Improved use of mechanistic data to better understand pseudoenzyme evolution

In this review, we have shown how Uniprot data can be used to identify pseudoenzymes 568 associated with a dataset of enzymes of interest. One such dataset is the M-CSA, which 569 contains information about the catalytic residues and the reaction mechanisms of 964 570 enzymes. By creating annotated phylogenetic trees, we are able to show where in the 571 572 evolutionary past the loss of function events occur and which catalytic residues mutations are 573 associated with those events. In the particular example shown, we observe that not all catalytic 574 mutations lead to loss of function, and that the use of this rule to all enzyme families, in general, 575 may be too simplistic. Our future major aim is to analyse the phylogenetic trees of all enzymes 576 in M-CSA that possess pseudoenzyme relatives. By using annotation specific to M-CSA, we will understand if the loss of function events are related with particular catalytic residue 577 578 functions and the specific chemistry the enzyme catalyses. For example, a mutation in a 579 residue acting as an electrostatic stabilizer may be tolerated, while a mutation in a nucleophile may not be. Furthermore, this tolerance to mutations may be dependent on the specific 580 reaction and the types of chemical groups in substrate(s), which adds another level of 581 complexity. 582

583

584 **Overview of different approaches to identify and characterise pseudoenzymes**

585 In this review, three ways are described to identify and characterize pseudoenzymes across 586 protein families. The first is the manual curation by UniProt using the literature or manually 587 checked computational annotation, so that although this method is labour intensive, it is the 588 most accurate of the three. Efforts are presented to characterize all the kinases and 589 phosphatases in C. elegans, for which about 20% and 3% of these family members are pseudoenzymes, respectively. The pseudoenzymes identified and annotated in this manner 590 will eventually form the 'gold standard' from which other automated methods will develop. The 591 592 second method uses sequence homology to identify close relatives and existing SwissProt data to identify pseudoenzymes, which are then annotated with the catalytic residues of the 593 original enzyme, to check their conservation. This sequence-based method identifies 594 pseudoenzymes in about one third of the enzymatic families in M-CSA. The final method uses 595 the CATH structural database as a starting point to identify related proteins, and UniProt and 596 Go annotation to categorize them as either enzyme or non-enzyme. Protein structure is more 597 598 conserved during evolution than sequence, hence, structural methods can see further into the 599 past to uncover more ancient relationships. Therefore, as expected, this method detects more 600 families with pseudoenzymes than the previous one, identifying about two thirds of a set of 601 enzymatic CATH superfamilies as pseudoenzyme-containing families. The Achilles heel of the 602 second and third methods is their reliance on the lack of catalytic annotation as a useful

603 criterion for defining non-enzymes. This is a general problem in the pseudoenzyme field, which 604 must use negative evidence, or lack of observed experimental catalytic activity, as a benchmark. The problem is compounded when using databases, where lack of annotation 605 does not mean that catalytic activity was evaluated. More experiments are therefore needed 606 607 to test more broadly for enzymatic activity and manual curation of the absence of evidence needs to be distinguished from clear experimental proof of a lack of catalysis. Until then, 608 comparative computational approaches (including those described above) are the most 609 610 powerful method for pseudoenzyme identification that we currently possess.

611

612 **CONCLUSIONS**:

The next decade will be an exciting, and potentially transformative, period of rapid 613 development in the pseudoenzyme field, as experimental and bioinformatic findings rapidly 614 merge, creating new databases that bring together exploitable information for specialists and 615 616 non-specialists alike. Feeding this information into experimental workflows will rapidly lead to 617 a revolution in our understanding of enzyme and pseudoenzyme evolution and inform 618 fundamental fields ranging from protein folding and enzyme mechanism to cell signalling, 619 metabolism, and drug discovery. Such endeavours will require much broader comparative 620 studies between enzyme families from diverse species, rather than the piecemeal approaches 621 currently favored for studying enzymes and pseudoenzymes in isolation away from their physiological environment. Beyond readily searchable comparative datasets, a major 622 outcome for such studies will also be the creation of benchmarks for studying, and predicting, 623 624 the effects of evolutionary and disease-associated mutations that take place in enzymes and 625 pseudoenzymes, especially when these changes are conserved in molecular "hot-spots". More data together with more sophisticated methods will facilitate the development of highly 626 627 accurate tools that can identify pseudoenzymes computationally. The assembly of these new sets of "rules" (and their allied rule-breakers) will then create truly useful, and biologically 628 informative, outputs. This will help to loosen current phenomenologically complex 629 630 classifications that can constrain, and often blur, the numerous strands of enzyme-based research taking place. In particular, the fruits of these labors are likely be a complete species-631 level catalog of pseudoenzymes across hundreds of distinct enzyme families and 632 633 superfamilies and the prioritization of biochemical, cellular and guided-evolution frameworks 634 to study pseudoenzymes that are of interest across scientific disciplines.

635 636

637 **REFERENCES AND NOTES:**

Murphy, J.M., H. Farhan, and P.A. Eyers, *Bio-zombie: the rise of pseudoenzymes in biology*.
 Biochem Soc Trans, 2017. 45(2): p. 95-104.

- Adrain, C. and M. Freeman, *New lives for old: evolution of pseudoenzyme function illustrated by iRhoms.* Nat Rev Mol Cell Biol, 2012. **13**(8): p. 489-98.
- 6423.Eyers, P.A. and J.M. Murphy, The evolving world of pseudoenzymes: proteins, prejudice and643zombies. BMC Biol, 2016. 14(1): p. 98.
- 6444.Jacobsen, A.V. and J.M. Murphy, The secret life of kinases: insights into non-catalytic signalling645functions from pseudokinases. Biochemical Society Transactions, 2017. **45**(3): p. 665-681.
- 6465.Reiterer, V., P.A. Eyers, and H. Farhan, Day of the dead: pseudokinases and647pseudophosphatases in physiology and disease. Trends Cell Biol, 2014. 24(9): p. 489-505.
- 648 6. Eyers, P.A. and J.M. Murphy, *Dawn of the dead: protein pseudokinases signal new adventures*649 *in cell biology.* Biochemical Society Transactions, 2013. **41**(4): p. 969-74.
- Kung, J.E. and N. Jura, *Prospects for pharmacological targeting of pseudokinases*. Nature
 reviews. Drug discovery, 2019.
- 652 8. Chen, M.J., J.E. Dixon, and G. Manning, *Genomics and evolution of protein phosphatases*.
 653 Science signaling, 2017. **10**(474).
- Boudeau, J., et al., *Emerging roles of pseudokinases*. Trends in cell biology, 2006. 16(9): p. 44352.
- 65610.Zeqiraj, E. and D.M. van Aalten, *Pseudokinases-remnants of evolution or key allosteric*657*regulators?* Current opinion in structural biology, 2010. **20**(6): p. 772-81.
- Pils, B. and J. Schultz, *Inactive enzyme-homologues find new function in regulatory processes.*J Mol Biol, 2004. **340**(3): p. 399-404.
- Todd, A.E., C.A. Orengo, and J.M. Thornton, Sequence and structural differences between *enzyme and nonenzyme homologs.* Structure, 2002. 10(10): p. 1435-51.
- 13. Jeffery, C.J., *The demise of catalysis, but new functions arise: pseudoenzymes as the phoenixes*of the protein world. Biochemical Society Transactions, 2019. 47(1): p. 371-379.
- 66414.Murphy, J.M., P.D. Mace, and P.A. Eyers, Live and let die: insights into pseudoenzyme665mechanisms from structure. Current opinion in structural biology, 2017. 47: p. 95-104.
- 66615.Zeqiraj, E., et al., Structure of the LKB1-STRAD-MO25 complex reveals an allosteric mechanism667of kinase activation. Science, 2009. **326**(5960): p. 1707-11.
- Walden, M., et al., *Pseudo-DUBs as allosteric activators and molecular scaffolds of protein complexes*. Biochemical Society Transactions, 2018. 46(2): p. 453-466.
- 670 17. Walden, *Metabolic control of SHMT2 oligomerization regulates ubiquitin-dependent cytokine*671 *signaling.* Nature, 2019. In Press.
- Bertram, K., et al., *Cryo-EM structure of a human spliceosome activated for step 2 of splicing.*Nature, 2017. **542**(7641): p. 318-323.
- 67419.Reiterer, V., et al., Pseudophosphatase STYX modulates cell-fate decisions and cell migration675by spatiotemporal regulation of ERK1/2. Proceedings of the National Academy of Sciences of676the United States of America, 2013. **110**(31): p. E2934-43.
- Reiterer, V., et al., *The pseudophosphatase STYX targets the F-box of FBXW7 and inhibits SCFFBXW7 function.* The EMBO journal, 2017. **36**(3): p. 260-273.
- Del-Toro, N., et al., *Capturing variation impact on molecular interactions in the IMEx Consortium mutations data set.* Nature communications, 2019. **10**(1): p. 10.
- 681 22. Orchard, S., et al., *Protein interaction data curation: the International Molecular Exchange*682 (*IMEx*) consortium. Nature methods, 2012. 9(4): p. 345-50.
- 68323.Reiterer, V. and H. Farhan, STYX: a pseudophosphatase that regulates MAPK signaling and684ubiquitin ligases Biochem Soc Trans, 2017. In press.
- Kwon, A., et al., *Tracing the origin and evolution of pseudokinases across the tree of life.*Science signaling, 2019. **12**(578).
- Wang, J., et al., *Reconstitution and structure of a plant NLR resistosome conferring immunity*.
 Science, 2019. **364**(6435).
- Wang, J., et al., *Ligand-triggered allosteric ADP release primes a plant NLR complex*. Science,
 2019. **364**(6435).

691 27. Klymiuk, V., A. Fatiukha, and T. Fahima, Wheat tandem kinases provide insights on disease-692 resistance gene flow and host-parasite co-evolution. The Plant journal : for cell and molecular 693 biology, 2019. 694 28. Olson, A.T., et al., A poxvirus pseudokinase represses viral DNA replication via a pathway 695 antagonized by its paralog kinase. PLoS pathogens, 2019. 15(2): p. e1007608. 696 29. Consortium, T.U., UniProt: a worldwide hub of protein knowledge. Nucleic acids research, 697 2019. 47(D1): p. D506-D515. 698 30. Zaru, R., M. Magrane, and C. O'Donovan, From the research laboratory to the database: the 699 Caenorhabditis elegans kinome in UniProtKB. The Biochemical journal, 2017. 474(4): p. 493-700 515. 701 31. Xu, B., et al., WNK1, a novel mammalian serine/threonine protein kinase lacking the catalytic lysine in subdomain II. The Journal of biological chemistry, 2000. 275(22): p. 16795-801. 702 703 32. Min, X., et al., Crystal structure of the kinase domain of WNK1, a kinase that causes a 704 hereditary form of hypertension. Structure, 2004. 12(7): p. 1303-11. 705 33. Gosal, G., K.J. Kochut, and N. Kannan, ProKinO: an ontology for integrative analysis of protein 706 kinases in cancer. PloS one, 2011. 6(12): p. e28782. 707 Kannan, N., et al., Structural and functional diversity of the microbial kinome. PLoS biology, 34. 708 2007. **5**(3): p. e17. 709 35. Cui, J., et al., Structure of Fam20A reveals a pseudokinase featuring a unique disulfide pattern 710 and inverted ATP-binding. eLife, 2017. 6. 711 36. Tagliabracci, V.S., et al., A Single Kinase Generates the Majority of the Secreted 712 Phosphoproteome. Cell, 2015. 161(7): p. 1619-32. Zhang, H., et al., Structure and evolution of the Fam20 kinases. Nature communications, 2018. 713 37. 714 **9**(1): p. 1218. 715 Sreelatha, A., et al., Protein AMPylation by an Evolutionarily Conserved Pseudokinase. Cell, 38. 716 2018. **175**(3): p. 809-821 e19. 717 39. Wen, J., et al., Xylose phosphorylation functions as a molecular switch to regulate 718 proteoglycan biosynthesis. Proceedings of the National Academy of Sciences of the United 719 States of America, 2014. 111(44): p. 15723-8. 720 40. Zhu, Q., et al., Structure of protein O-mannose kinase reveals a unique active site architecture. 721 eLife, 2016. 5. 722 Black, M.H., et al., Bacterial pseudokinase catalyzes protein polyglutamylation to inhibit the 41. 723 SidE-family ubiquitin ligases. Science, 2019. 364(6442): p. 787-792. 724 42. Qiu, J., et al., A unique deubiquitinase that deconjugates phosphoribosyl-linked protein 725 ubiquitination. Cell research, 2017. 27(7): p. 865-881. 726 43. Beraki, T., et al., Divergent kinase regulates membrane ultrastructure of the Toxoplasma 727 parasitophorous vacuole. Proceedings of the National Academy of Sciences of the United 728 States of America, 2019. 116(13): p. 6361-6370. 729 44. Huang, Y., et al., BubR1 phosphorylates CENP-E as a switch enabling the transition from lateral 730 association to end-on capture of spindle microtubules. Cell research, 2019. 731 45. Suijkerbuijk, S.J., et al., The vertebrate mitotic checkpoint protein BUBR1 is an unusual 732 pseudokinase. Developmental cell, 2012. 22(6): p. 1321-9. 733 46. Murphy, J.M., et al., A robust methodology to subclassify pseudokinases based on their 734 nucleotide-binding properties. The Biochemical journal, 2014. 457(2): p. 323-34. 735 47. Murphy, J.M., et al., Insights into the evolution of divergent nucleotide-binding mechanisms 736 among pseudokinases revealed by crystal structures of human and mouse MLKL. The 737 Biochemical journal, 2014. 457(3): p. 369-77. 738 Clifton, B.E., et al., Evolution of cyclohexadienyl dehydratase from an ancestral solute-binding 48. 739 *protein.* Nature chemical biology, 2018. **14**(6): p. 542-547. 740 49. Kaltenbach, M., et al., Evolution of chalcone isomerase from a noncatalytic ancestor. Nature 741 chemical biology, 2018. 14(6): p. 548-555.

742 50. Mukherjee, K., et al., Evolution of CASK into a Mg2+-sensitive kinase. Science signaling, 2010. 743 **3**(119): p. ra33. 744 51. Byrne, D.P., D.M. Foulkes, and P.A. Eyers, Pseudokinases: update on their functions and 745 evaluation as new drug targets. Future medicinal chemistry, 2017. 9(2): p. 245-265. 746 Foulkes, D.M., et al., Covalent inhibitors of EGFR family protein kinases induce degradation of 52. 747 human Tribbles 2 (TRIB2) pseudokinase in cancer cells. Science Signaling, 2018. 11(549). 748 53. Ribeiro, A.J.M., et al., Mechanism and Catalytic Site Atlas (M-CSA): a database of enzyme 749 reaction mechanisms and active sites. Nucleic acids research, 2018. 46(D1): p. D618-D623. 750 54. Potter, S.C., et al., HMMER web server: 2018 update. Nucleic acids research, 2018. 46(W1): p. 751 W200-W204. 752 55. Furnham, N., et al., Large-Scale Analysis Exploring Evolution of Catalytic Machineries and 753 Mechanisms in Enzyme Superfamilies. Journal of molecular biology, 2016. 428(2 Pt A): p. 253-754 267. 755 Kwon, Tracing the origin and evolution of pseudokinases across the tree of life. Science 56. 756 Signaling, 2019. In Press. 757 57. Sillitoe, I., et al., CATH: expanding the horizons of structure-based functional annotations for 758 genome sequences. Nucleic acids research, 2019. 47(D1): p. D280-D284. 759 58. Lewis, T.E., et al., Gene3D: Extensive prediction of globular domains in proteins. Nucleic acids 760 research, 2018. 46(D1): p. D1282. 761 59. Das, S., et al., Functional classification of CATH superfamilies: a domain-based approach for 762 protein function annotation. Bioinformatics, 2015. **31**(21): p. 3460-7. 763 60. Jiang, Y., et al., An expanded evaluation of protein function prediction methods shows an 764 *improvement in accuracy.* Genome biology, 2016. **17**(1): p. 184. 765 61. Ashburner, M., et al., Gene ontology: tool for the unification of biology. The Gene Ontology 766 Consortium. Nature genetics, 2000. 25(1): p. 25-9. Sharir-Ivry, A. and Y. Xia, Nature of Long-Range Evolutionary Constraint in Enzymes: Insights 767 62. 768 from Comparison to Pseudoenzymes with Similar Structures. Molecular biology and evolution, 769 2018. **35**(11): p. 2597-2606. 770 63. Hwang, J., et al., Crystal structure of the human N-Myc downstream-regulated gene 2 protein 771 provides insight into its role as a tumor suppressor. The Journal of biological chemistry, 2011. 772 **286**(14): p. 12450-60. 773 Bailey, F.P., et al., The Tribbles 2 (TRB2) pseudokinase binds to ATP and autophosphorylates in 64. 774 a metal-independent manner. The Biochemical journal, 2015. **467**(1): p. 47-62. 775 65. Ungureanu, D., et al., The pseudokinase domain of JAK2 is a dual-specificity protein kinase that 776 negatively regulates cytokine signaling. Nature structural & molecular biology, 2011. 18(9): p. 777 971-6. 778 66. Buchholz, K.R., P.W. Bowyer, and J.C. Boothroyd, Bradyzoite pseudokinase 1 is crucial for 779 efficient oral infectivity of the Toxoplasma gondii tissue cyst. Eukaryotic cell, 2013. 12(3): p. 780 399-410. 781 67. Hanks, S.K. and T. Hunter, Protein kinases 6. The eukaryotic protein kinase superfamily: kinase 782 (catalytic) domain structure and classification. FASEB journal : official publication of the 783 Federation of American Societies for Experimental Biology, 1995. 9(8): p. 576-96. 784 68. Walker, J.E., et al., Distantly related sequences in the alpha- and beta-subunits of ATP 785 synthase, myosin, kinases and other ATP-requiring enzymes and a common nucleotide binding 786 *fold.* The EMBO journal, 1982. **1**(8): p. 945-51. 787 69. Stefely, J.A., et al., Mitochondrial ADCK3 employs an atypical protein kinase-like fold to enable 788 coenzyme Q biosynthesis. Molecular cell, 2015. 57(1): p. 83-94. 789 70. Cameron, A.J., et al., PKC maturation is promoted by nucleotide pocket occupation 790 independently of intrinsic kinase activity. Nature structural & molecular biology, 2009. 16(6): 791 p. 624-30.

- 79271.Otto, T., et al., Stabilization of N-Myc is a critical function of Aurora A in human793neuroblastoma. Cancer cell, 2009. 15(1): p. 67-78.
- 794 72. Murphy, J.M., et al., *The pseudokinase MLKL mediates necroptosis via a molecular switch* 795 *mechanism.* Immunity, 2013. **39**(3): p. 443-53.
- 73. Patel, O., et al., Structure of SgK223 pseudokinase reveals novel mechanisms of homotypic and
 797 heterotypic association. Nature communications, 2017. 8(1): p. 1157.
- 74. Murphy, J.M., et al., *Molecular Mechanism of CCAAT-Enhancer Binding Protein Recruitment*799 by the TRIB1 Pseudokinase. Structure, 2015. 23(11): p. 2111-21.
- 80075.Jamieson, S.A., et al., Substrate binding allosterically relieves autoinhibition of the801pseudokinase TRIB1. Science Signaling, 2018. **11**(549).
- 80276.Eyers, P.A., K. Keeshan, and N. Kannan, Tribbles in the 21st Century: The Evolving Roles of803Tribbles Pseudokinases in Biology and Disease. Trends in cell biology, 2017. 27(4): p. 284-298.
- 80477.Jamieson, S.A., et al., Substrate binding allosterically relieves autoinhibition of the805pseudokinase TRIB1. Sci Signal, 2018. 11(549).
- 78. Zeqiraj, E., et al., ATP and MO25alpha regulate the conformational state of the STRADalpha
 807 pseudokinase and activation of the LKB1 tumour suppressor. PLoS Biol, 2009. 7(6): p.
 808 e1000126.
- 809 79. Littlefield, P., et al., Structural analysis of the EGFR/HER3 heterodimer reveals the molecular
 810 basis for activating HER3 mutations. Sci Signal, 2014. 7(354): p. ra114.
- 80. Lupardus, P.J., et al., *Structure of the pseudokinase-kinase domains from protein kinase TYK2*812 *reveals a mechanism for Janus kinase (JAK) autoinhibition.* Proc Natl Acad Sci U S A, 2014.
 813 **111**(22): p. 8025-30.
- 81481.Rostislavleva, K., et al., Structure and flexibility of the endosomal Vps34 complex reveals the815basis of its function on membranes. Science, 2015. **350**(6257): p. aac7365.
- 81682.Edmund, A.B., et al., The pseudokinase domains of guanylyl cyclase-A and -B allosterically817increase the affinity of their catalytic domains for substrate. Sci Signal, 2019. 12(566).
- 818 83. Murphy, J.M., et al., *The pseudokinase MLKL mediates necroptosis via a molecular switch*819 *mechanism.* Immunity, 2013. **39**(3): p. 443-453.
- 84. Jacobsen, A.V., et al., *HSP90 activity is required for MLKL oligomerisation and membrane*821 *translocation and the induction of necroptotic cell death*. Cell Death Dis, 2016. **7**: p. e2051.
- 822 85. Kung, J.E. and N. Jura, *The pseudokinase TRIB1 toggles an intramolecular switch to regulate*823 *COP1 nuclear export.* The EMBO journal, 2019. **38**(4).
- 824 86. Ha, B.H. and T.J. Boggon, *The crystal structure of pseudokinase PEAK1 (Sugen kinase 269)*825 *reveals an unusual catalytic cleft and a novel mode of kinase fold dimerization*. J Biol Chem,
 826 2018. 293(5): p. 1642-1650.
- 827 87. Lecointre, C., et al., *Dimerization of the Pragmin Pseudo-Kinase Regulates Protein Tyrosine*828 *Phosphorylation.* Structure, 2018. **26**(4): p. 545-554 e4.
- 829 88. Ha, B.H. and T.J. Boggon, *The crystal structure of pseudokinase PEAK1 (Sugen kinase 269)*830 *reveals an unusual catalytic cleft and a novel mode of kinase fold dimerization.* The Journal of
 831 biological chemistry, 2018. **293**(5): p. 1642-1650.
- 832 89. Liu, L., et al., *Homo- and Heterotypic Association Regulates Signaling by the SgK269/PEAK1*833 *and SgK223 Pseudokinases.* The Journal of biological chemistry, 2016. **291**(41): p. 21571834 21583.
- Safari, F., et al., *Mammalian Pragmin regulates Src family kinases via the Glu-Pro-Ile-Tyr-Ala*(EPIYA) motif that is exploited by bacterial effectors. Proceedings of the National Academy of
 Sciences of the United States of America, 2011. **108**(36): p. 14938-43.
- 838 91. Tanaka, H., H. Katoh, and M. Negishi, *Pragmin, a novel effector of Rnd2 GTPase, stimulates*839 *RhoA activity.* The Journal of biological chemistry, 2006. **281**(15): p. 10355-64.
- Stefely, J.A., et al., *Cerebellar Ataxia and Coenzyme Q Deficiency through Loss of Unorthodox Kinase Activity.* Molecular cell, 2016. 63(4): p. 608-620.

- 842 93. Childers, W.S., et al., *Cell fate regulation governed by a repurposed bacterial histidine kinase.*843 PLoS Biol, 2014. **12**(10): p. e1001979.
- 84494.Ng, A.A., et al., The CMT4B disease-causing phosphatases Mtmr2 and Mtmr13 localize to the845Schwann cell cytoplasm and endomembrane compartments, where they depend upon each846other to achieve wild-type levels of protein expression. Hum Mol Genet, 2013. 22(8): p. 1493-847506.
- 848 95. Reiterer, V., et al., *The pseudophosphatase STYX targets the F-box of FBXW7 and inhibits*849 *SCFFBXW7 function*. EMBO J, 2017. **36**(3): p. 260-273.
- 850 96. Blanchetot, C., L.G. Tertoolen, and J. den Hertog, *Regulation of receptor protein-tyrosine*851 *phosphatase alpha by oxidative stress.* The EMBO journal, 2002. **21**(4): p. 493-503.
- 852 97. Chang, D.W., et al., *c-FLIP(L) is a dual function regulator for caspase-8 activation and CD95-*853 *mediated apoptosis.* EMBO J, 2002. **21**(14): p. 3704-14.
- 85498.Adrain, C., et al., Tumor necrosis factor signaling requires iRhom2 to promote trafficking and855activation of TACE. Science, 2012. 335(6065): p. 225-8.
- 85699.Christova, Y., et al., Mammalian iRhoms have distinct physiological functions including an857essential role in TACE regulation. EMBO Rep, 2013. 14(10): p. 884-90.
- Klemencic, M., et al., *Phylogenetic Distribution and Diversity of Bacterial Pseudo- Orthocaspases Underline Their Putative Role in Photosynthesis.* Frontiers in plant science,
 2019. 10: p. 293.
- Piao, S., et al., Crystal structure of a clip-domain serine protease and functional roles of the clip
 domains. The EMBO journal, 2005. 24(24): p. 4404-14.
- 863102.Geng, X., et al., Caenorhabditis elegans caspase homolog CSP-2 inhibits CED-3 autoactivation864and apoptosis in germ cells. Cell death and differentiation, 2009. 16(10): p. 1385-94.
- 865 103. Geng, X., et al., Inhibition of CED-3 zymogen activation and apoptosis in Caenorhabditis
 866 elegans by caspase homolog CSP-3. Nature structural & molecular biology, 2008. 15(10): p.
 867 1094-101.
- 868104.Eddins, M.J., et al., Mms2-Ubc13 covalently bound to ubiquitin reveals the structural basis of869linkage-specific polyubiquitin chain formation. Nat Struct Mol Biol, 2006. 13(10): p. 915-20.
- 870 105. Lingaraju, G.M., et al., *Crystal structure of the human COP9 signalosome*. Nature, 2014.
 871 512(7513): p. 161-5.
- Worden, E.J., C. Padovani, and A. Martin, *Structure of the Rpn11-Rpn8 dimer reveals mechanisms of substrate deubiquitination during proteasomal degradation*. Nature structural
 & molecular biology, 2014. 21(3): p. 220-7.
- Pathare, G.R., et al., *Crystal structure of the proteasomal deubiquitylation module Rpn8- Rpn11.* Proceedings of the National Academy of Sciences of the United States of America,
 2014. 111(8): p. 2984-9.
- 878108.Schafer, I.B., et al., The structure of the Pan2-Pan3 core complex reveals cross-talk between879deadenylase and pseudokinase. Nature structural & molecular biology, 2014. **21**(7): p. 591-8.

109. Ceccarelli, D.F., et al., FAM105A/OTULINL Is a Pseudodebuiquitinase of the OTU-Class that
 Localizes to the ER Membrane. Structure, 2019.

- Strickland, M., et al., *Tsg101 chaperone function revealed by HIV-1 assembly inhibitors*. Nat
 Commun, 2017. 8(1): p. 1391.
- 111. Walden, M., et al., *Metabolic control of BRISC-SHMT2 assembly regulates immune signalling*.
 Nature, 2019. **570**(7760): p. 194-199.
- 112. Mandel, C.R., et al., *Polyadenylation factor CPSF-73 is the pre-mRNA 3'-end-processing endonuclease.* Nature, 2006. 444(7121): p. 953-6.
- Hillen, H.S., et al., *Mechanism of Transcription Anti-termination in Human Mitochondria*. Cell,
 2017. **171**(5): p. 1082-1093 e13.
- Lazzaretti, D., et al., *The bicoid mRNA localization factor Exuperantia is an RNA-binding pseudonuclease.* Nat Struct Mol Biol, 2016. 23(8): p. 705-13.

- Rosenberg, O.S., et al., Substrates Control Multimerization and Activation of the Multi-Domain
 ATPase Motor of Type VII Secretion. Cell, 2015. 161(3): p. 501-12.
- 894116.Foster, R., et al., Identification of a novel human Rho protein with unusual properties: GTPase895deficiency and in vivo farnesylation. Mol Cell Biol, 1996. 16(6): p. 2689-99.
- Wennerberg, K., et al., *Rnd proteins function as RhoA antagonists by activating p190 RhoGAP*.
 Curr Biol, 2003. **13**(13): p. 1106-15.
- 898 118. Stiegler, A.L. and T.J. Boggon, *PseudoGTPase domains in p190RhoGAP proteins: a mini-review.*899 Biochem Soc Trans, 2018. 46(6): p. 1713-1720.
- 900 119. Stiegler, A.L. and T.J. Boggon, *The N-Terminal GTPase Domain of p190RhoGAP Proteins Is a* 901 *PseudoGTPase*. Structure, 2018. **26**(11): p. 1451-1461 e4.
- 902 120. Basilico, F., et al., *The pseudo GTPase CENP-M drives human kinetochore assembly*. Elife, 2014.
 903 3: p. e02978.
- 904 121. Schroeder, C.M., et al., A Ras-like domain in the light intermediate chain bridges the dynein
 905 motor to a cargo-binding region. Elife, 2014. 3: p. e03351.
- Ranok, A., et al., Structural and thermodynamic insights into chitooligosaccharide binding to human cartilage chitinase 3-like protein 2 (CHI3L2 or YKL-39). J Biol Chem, 2015. 290(5): p.
 2617-29.
- 909 123. Schimpl, M., et al., *Human YKL-39 is a pseudo-chitinase with retained chitooligosaccharide-*910 *binding properties.* Biochem J, 2012. 446(1): p. 149-57.
- 911 124. Wong, W., et al., Structure of Plasmodium falciparum Rh5-CyRPA-Ripr invasion complex.
 912 Nature, 2019. 565(7737): p. 118-121.
- 125. Kuzina, E.S., et al., Structures of ligand-occupied beta-Klotho complexes reveal a molecular
 mechanism underlying endocrine FGF specificity and activity. Proceedings of the National
 Academy of Sciences of the United States of America, 2019.
- 916 126. Velez, N., C.A. Brautigam, and M.A. Phillips, *Trypanosoma brucei S-adenosylmethionine* 917 *decarboxylase N terminus is essential for allosteric activation by the regulatory subunit* 918 *prozyme.* J Biol Chem, 2013. 288(7): p. 5232-40.
- 919 127. Volkov, O.A., et al., *Relief of autoinhibition by conformational switch explains enzyme*920 *activation by a catalytically dead paralog.* Elife, 2016. 5.
- Willert, E.K., R. Fitzpatrick, and M.A. Phillips, *Allosteric regulation of an essential trypanosome polyamine biosynthetic enzyme by a catalytically dead homolog.* Proc Natl Acad Sci U S A,
 2007. **104**(20): p. 8275-80.
- 924129.He, S., et al., Viral pseudo-enzymes activate RIG-I via deamidation to evade cytokine925production. Mol Cell, 2015. 58(1): p. 134-46.
- 926130.Nguyen, S., et al., Allosteric activation of trypanosomatid deoxyhypusine synthase by a
catalytically dead paralog. J Biol Chem, 2013. 288(21): p. 15256-67.
- 131. Afanador, G.A., D.R. Tomchick, and M.A. Phillips, *Trypanosomatid Deoxyhypusine Synthase*Activity Is Dependent on Shared Active-Site Complementation between Pseudoenzyme
 Paralogs. Structure, 2018. 26(11): p. 1499-1512 e5.
- 132. Richter, V., et al., *Structural and functional analysis of MiD51, a dynamin receptor required for mitochondrial fission*. The Journal of cell biology, 2014. **204**(4): p. 477-86.
- 933 133. Yao, J., et al., *Human pantothenate kinase 4 is a pseudo-pantothenate kinase*. Protein science
 934 : a publication of the Protein Society, 2019.
- 935 134. Allali-Hassani, A., et al., Structural and chemical profiling of the human cytosolic
 936 sulfotransferases. PLoS biology, 2007. 5(5): p. e97.
- 135. Rao, F.V., et al., Structure of a bacterial putative acetyltransferase defines the fold of the human O-GlcNAcase C-terminal domain. Open Biol, 2013. 3(10): p. 130021.
- 939 136. Bozatzi, P. and G.P. Sapkota, *The FAM83 family of proteins: from pseudo-PLDs to anchors for* 940 *CK1 isoforms.* Biochem Soc Trans, 2018. **46**(3): p. 761-771.
- 941 137. Fulcher, L.J., et al., *The DUF1669 domain of FAM83 family proteins anchor casein kinase 1 isoforms.* Sci Signal, 2018. **11**(531).

- 943 138. Banumathi, S., et al., Structure of the neurotoxic complex vipoxin at 1.4 A resolution. Acta
 944 Crystallogr D Biol Crystallogr, 2001. 57(Pt 11): p. 1552-9.
- 139. Larson, H.N., H. Weiner, and T.D. Hurley, *Disruption of the coenzyme binding site and dimer*interface revealed in the crystal structure of mitochondrial aldehyde dehydrogenase "Asian"
 variant. J Biol Chem, 2005. 280(34): p. 30550-6.
- 140. Teoh, M.L., P.J. Walasek, and D.H. Evans, *Leporipoxvirus Cu,Zn-superoxide dismutase (SOD) homologs are catalytically inert decoy proteins that bind copper chaperone for SOD.* J Biol
 Chem, 2003. 278(35): p. 33175-84.
- 951141.Liu, W., et al., Mechanism for KRIT1 release of ICAP1-mediated suppression of integrin952activation. Molecular cell, 2013. 49(4): p. 719-29.
- 142. Elkhateeb, E., et al., *The role of mouse 2',5'-oligoadenylate synthetase 1 paralogs*. Infection,
 genetics and evolution : journal of molecular epidemiology and evolutionary genetics in
 infectious diseases, 2016. 45: p. 393-401.
- 143. Courtney, S.C., et al., *Identification of novel host cell binding partners of Oas1b, the protein conferring resistance to flavivirus-induced disease in mice.* Journal of virology, 2012. 86(15): p.
 7953-63.
- 959

960 ACKNOWLEDGEMENTS:

- 961 We thank all the scientists and support staff who organized, attended and contributed to the
- scientific input at "Pseudoenzymes 2016: from signalling mechanisms to disease", which took
- 963 place at the Liverpool Maritime Museum, UK, in September 2016 and "Pseudoenzymes 2018:
- 964 From molecular mechanisms to cell biology", which took place in May 2018 in Sardinia, Italy,
- 965 both of which informed this review.

966 **FUNDING**:

- This work was initially funded by a Royal Society Research Grant (to PAE) and subsequently by North West Cancer Research grants CR1088 and CR1097 (to PAE). We thank the Biochemical Society and EMBO for awarding dedicated conference and workshop funding to support the development of the pseudoenzyme field.

971 **AUTHOR CONTRIBUTIONS:**

- 972 AB, SD, ND, RZ, SO, JT and CO performed bioinformatics analyses, and created databases
- and websites. EZ, JMM and PAE assembled the information in Table 1. All authors co-wrote
- 974 the manuscript, and all authors approved the final version prior to submission.

975 **COMPETING INTERESTS:**

- The authors declare that they have no conflicts of interest related to the content of this review.
- 978

979 FIGURE LEGENDS:

Figure 1. Regulated assembly of the macromolecular BRISC-SHMT2 pseudoenzyme-980 containing complex. Top, schematic of the SHMT2 dimer-tetramer oligomerization 981 982 transition, which is regulated by pyridoxal-5'-phosphate (PLP), the active co-factor form of vitamin B₆. The SHMT2 dimer, which is inactive as a methyltransferase (left) specifically 983 interacts with, and inhibits, the pseudoenzyme-containing BRISC complex (bottom), 984 985 revealing a new 'moonlighting' role for SHMT2. The PLP-bound SHMT2 tetramer, which is active as a methyltransferase, is unable to bind or inhibit the BRISC complex. Bottom, 986 schematic of the multimeric BRISC-SHMT2 DUB complex, which contains the active DUB 987 BRCC36 (MPN+) and the inactive Abraxas2 pseudo-DUB (MPN-). BRCC45 contains three 988 989 pseudo-E2 domains (UEVs), and is discussed further in Table 1 and in the text.

Figure 2. Computational annotation of enzymes and pseudoenzymes in UniProt. (A)
Workflow for expert curation of pseudoenzymes. Reported evidence is based on experimental
data, ortholog similarity and sequence analysis. (B) Worked example of the "inactive"
annotation for the *C.elegans* pseudophosphatase egg-4 (UniProt KB 001767), including the
protein name, the protein-coding evidence (status) and the caution comment.

995

Figure 3. Analyzing the (pseudo)kinome and (pseudo)phosphatome in a model worm.

- (A) Calculated percentage of pseudokinases in the *C. elegans* kinome (grey). (B) Percentage
 of phosphatases in the *C. elegans* phosphatome (grey).
- 999

Figure 4. Diversity in ATP-binding mode and the acquisition of non-canonical catalytic 1000 1001 functions amongst bioinformatically-annotated pseudokinases. (A) The catalyticallyinactive pseudokinase, FAM20A, still binds ATP (red sticks), but in an inverted conformation 1002 [35] (PDB, 5wrs). Like the related FAM20B, the position of the α C helix (orange) differs from 1003 1004 that in canonical protein kinases. (B) The highly atypical annotated pseudokinase SelO [38] 1005 (PDB, 6eac), can catalyse protein AMPylation via an unusual catalytic mechanism. Like 1006 FAM20A (A), SelO binds ATP in an inverted conformation, but in addition, it catalyses AMP 1007 transfer to protein substrates. ATP analog AMP-PNP is shown as red sticks; Mg2+ and Ca2+ are shown as yellow and green spheres, respectively. (C) The predicted pseudokinase, 1008 FAM20B (PDB, 5xoo), is actually a catalytically-active xylose kinase [37] involved in 1009 1010 proteoglycan synthesis. The sugar substrate is shown as green sticks; an adenine (red sticks) is also modelled in the structure. (D) Like FAM20B, the predicted pseudokinase, 1011 SgK169/protein O-mannose kinase (POMK; PDB, 5gza) is actually a sugar kinase [40]. 1012 SgK169/POMK closely resembles a typical protein kinase fold with conventional α C helix 1013 (orange) position, nucleotide-binding mode (red sticks) and Mg2+ cofactors (yellow spheres), 1014 1015 and excitingly, this structure captures the protein bound to its sugar substrate (green sticks).

1016 Figure 5. Estimating the ratio of pseudoenzymes in known enzyme families.

A family is defined here as the group of Swiss-Prot sequences that are homologous to one enzyme/entry in M-CSA. Each entry in M-CSA corresponds to one enzyme with a unique enzyme mechanism, so the same EC number can be represented more than once if evolved independently with distinct mechanisms. Sequences are categorised as enzymes if they have a catalytic UniProt keyword and as pseudoenzymes otherwise. The green bar corresponds to enzyme families that contain only enzymes, and the blue bars to enzyme families that contain a variable percentage of pseudoenzymes.

- 1024
- 1025

1027 Figure 6. Annotated phylogenetic tree for soybean β -amylase.

1028 Phylogenetic tree for homologous sequences of soybean β -amylase (P10538) annotated 1029 with the catalytic residues identified in M-CSA, their EC numbers and PFAM domains. The 1030 tree contains 20 homologues from plants, of which 6 are chloroplastic and 5 are bacterial. 1031 The two "inactive" pseudoenzymes identified (Q9FM68 and Q8VYW2) belong to Thale 1032 Cress (*Arabidopsis thaliana*), and one of these (Q9FM68) is known to have physiological 1033 regulatory functions in this organism.

1034

Figure 7. Analysis of the distribution of pseudoenzymes. (A) Sequences in CATH
superfamilies are sub-classified into functional families (FunFams) predicted to share similar
structures and functions, and which can be used to understand protein function evolution. (B)
Distribution of the number of enzyme superfamilies (containing catalytic domains) that have
varying proportions of functional families with enzyme annotations. (C) Showing the number
of putative pseudoenzyme families in enzyme superfamilies (containing catalytic domains).
These remain to be confirmed by further analysis and experimental testing.

Figure 8: Pseudoenzymes in the NTF2 family. An example comparing the known 1042 pseudoenzyme family of NTF2 (shown in blue) and the related enzyme family of scytalone 1043 hydratases (shown in orange) in CATH superfamily 3.10.450.50. The established catalytic 1044 1045 residues for the Scytalone hydratase family (EC 4.2.1.94) are shown in red. The structural alignment between the structures 1stdA00 and 1ounA00 from the enzyme and pseudoenzyme 1046 1047 families was generated using CATH-superpose (https://cathtools.readthedocs.io/en/latest/tools/cath-superpose/). On the right-hand panel, the highly 1048 conserved residues in the two families are shown for structurally equivalent positions lying in 1049 the active site of the enzyme family. The height of the characters reflects the degree of 1050 1051 conservation and the colours change according to physicochemical characteristics.

Figure 1

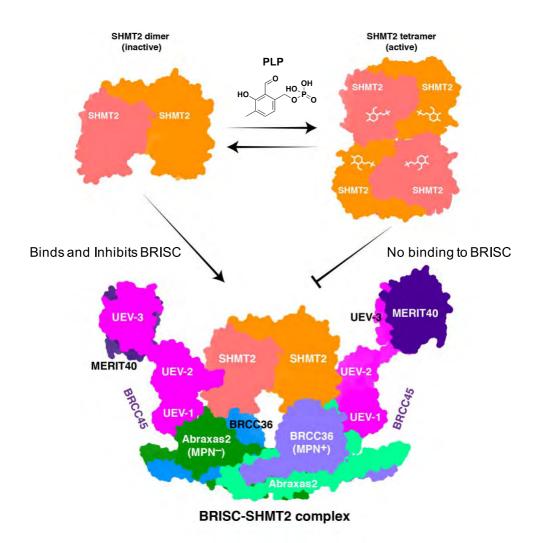


Figure 2

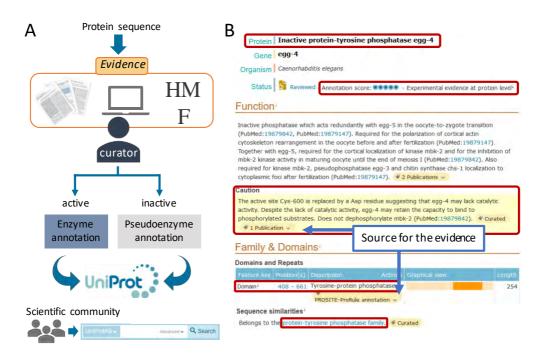




Figure 3

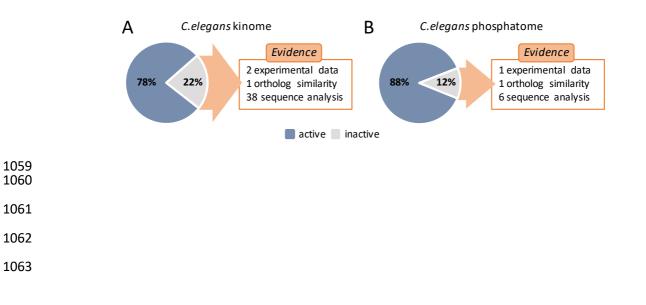
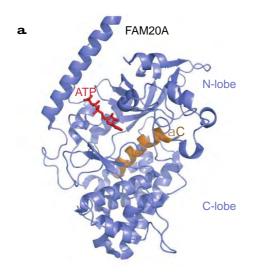
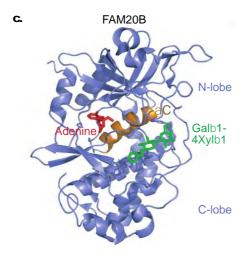
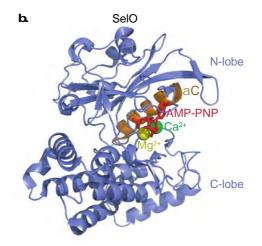


Figure 4







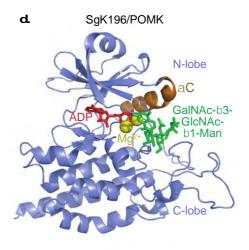
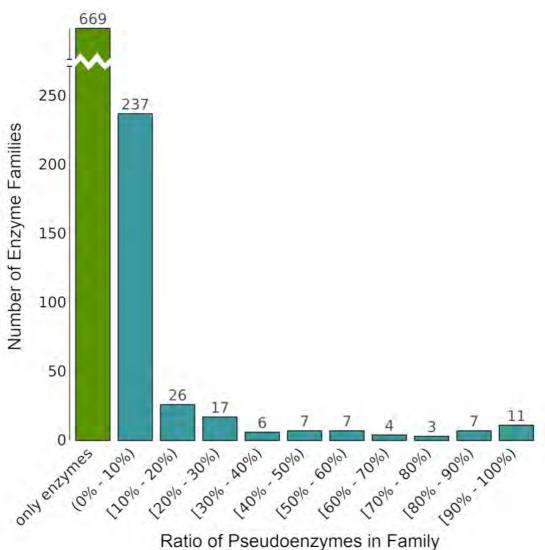


Figure 5



Ratio of Pseudoenzymes in Family

Figure 6

		Is Enzyme	Asp102	Glu187	Thr343	Glu381	Leu384	Ы	
-	Japanese rice Q10RZ1 Beta-amylase 2, chloroplastic		Asp183	Glu267	Thr427	Glu465	Leu468	3.2.1.2	-
	Japanese rice Q9AV88 Beta-amylase 1, chloroplastic		Asp163	Glu247	Thr408	Glu446	Leu449	3.2.1.2	
	Thale cress O23553 Beta-amylase 3, chloroplastic		Asp175	Glu259	Thr418	Glu456	Leu459	3.2.1.2	
	Thale cress Q9FM68 Inactive beta-amylase 4, chloroplastic		None	Glu269	Pro435	Arg473	Ser476		-
j	Thale cress Q8VYW2 Inactive beta-amylase 9		Gin172	Glu255	Pro411	Gin449	Ser452		
	Thale cress Q9FH80 Beta-amylase 8		Asn345	Glu429	Val584	Glu623	lle626	3.2.1.2	
	Thale cress O65258 Beta-amylase 2, chloroplastic		Asp185	Glu269	Thr424	Glu465	Leu468	3.2.1.2	4 3
۲ <u>ــــــــــــــــــــــــــــــــــــ</u>	Thale cress O80831 Beta-amylase 7		Asp338	Glu422	Val577	Glu618	Leu621	3.2.1.2	-
	Maize P55005 Beta-amylase		Asp99	Glu184	Thr340	Glu378	Leu381	3.2.1.2	-
	rye P30271 Beta-amylase		None	None	Thr36	Glu74	Leu77	3.2.1.2	I -
	Barley P16098 Beta-amylase		Asp99	Glu184	Thr340	Glu378	Leu381	3.2.1.2	-
ų į.	P82993 Beta-amylase		Asp99	Glu184	Thr340	Glu378	Leu381	3.2.1.2	-
L	bread wheat P93594 Beta-amylase		Asp99	Glu184	Thr340	Glu378	Leu381	3.2.1.2	-
· · · · · · · · · · · · · · · · · · ·	Thale cress Q8L762 Beta-amylase 6		Asp167	Glu252	Thr409	Glu447	Leu450	3.2.1.2	+ +
·····	white clover O65015 Beta-amylase		Asp102	Glu187	Thr343	Glu381	Leu384	3.2.1.2	-
	cowpea O64407 Beta-amylase	0	Asp102	Glu187	Thr343	Glu381	Leu384	3.2.1.2	-
	Alfalfa O22585 Beta-amylase		Asp102	Glu187	Thr343	Glu381	Leu384	3.2.1.2	
5 L	Soybean P10538 Beta-amylase		Asp102	Glu187	Thr343	Glu381	Leu384	3.2.1.2	-
	Sweet potato P10537 Beta-amylase		Asp103	Glu188	Thr345	Glu383	Leu386	3.2.1.2	+ +
	Thale cress P25853 Beta-amylase 5		Asp104	Glu189	Thr345	Glu383	Leu386	3.2.1.2	-
	Bacteria P36924 Beta-amylase	ē	Asp128	Glu202	Thr360	Glu397	Leu400	3.2.1.2	- 8
	- Bacteria P06547 Beta-amylase		Asp126	Glu199	Thr358	Glu395	Leu398	3.2.1.2	- 15-
	Bacteria P21543 Beta/alpha-amylase		Asp125	Glu198	Thr357	Glu394	Leu397	3.2.1.1, 3.2.1.2	
- L	Bacteria P96513 Beta-amylase		Asp125	Glu198	Thr357	Glu394	Leu397	3.2.1.2	-
	P19584 Thermophilic beta-amylase		Thr122	Glu195	Thr353	Glu392	Leu395	3.2.1.2	
	Thale cress Q9LIR6 Beta-amylase 1, chloroplastic	i.	Asp195	Glu279	Thr439	Glu477	Leu480	3.2.1.2	-

Figure 7

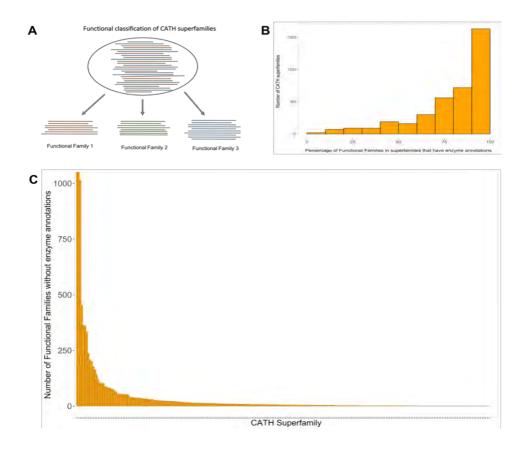


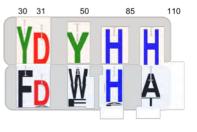
Figure 8





Scytalone dehydratase (1stdA00)





Nuclear Transport Factor 2 (1ounA00)

Table 1: The remarkable diversity of pseudoenzymes. Examples of pseudoenzymes from across the kingdoms of life, organized by class and function. Pseudoenzymes are highlighted in blue, whereas relevant conventional enzymes are in black. A broad selection of well-studied pseudoenzymes are discussed; the list is not meant to be comprehensive.

1091

Class	Function	Examples	References
Pseudokinase	Allosteric regulation of conventional protein kinase	STRADα:LKB1, HER3:HER2, JAK JH2:JAK JH1, FAM20A:FAM20C, where a pseudokinase domain regulates activity of a conventional kinase in a cognate pair. Also a common mechanism in plant pseudokinases (including tandem pseudokinases) such as RKS1	[25-27, 35, 56, 78-80]
	Regulator of a phospholipid kinae	Vps15 is a probable pseudokinase that forms part of a multimeric complex for regulation of the Vps34	[81]
	Allosteric regulation of other enzymes	Phosphatidylinositol 3-kinase	[82]
	012,1100	Pseudokinase domains of Guanylyl cyclase-A and -B regulate activity of the tandem guanylyl cyclase domains	[82]
	Molecular switch	Phosphorylation of the MLKL pseudokinase domain triggers exposure of the executioner four-helix bundle domain and cell death	[47, 83]
	Protein interaction domain	MLKL pseudokinase domain is regulated by binding to the RIPK3 kinase domain and HSP90:Cdc37 co-chaperones	[73, 84]
	Scaffold for assembly of signalling complexes	Tribbles (TRIB) pseudokinases nucleate assembly of a complex between a substrate (e.g. C/EBP α) and the E3	[74-76, 85]
		Ubiquitin ligase, COP1, whose intrasubcellular localization is controlled by TRIB1	[73, 86]
		SgK223 (Pragmin)/SgK269 (PEAK1) form higher-order signaling assemblies that include Src-family kinases	[73, 87-91]

[69, 92]

	Fundamental metabolic regulators of isoprenoid lipid production	UbiB pseudoenzyme family adopts an (inactive?) atypical protein kinase-like (PKL)-fold found in bacteria, archaea and eukaryotes. Human mitochondrial ADCK3 pseudokinase binds nucleotides such as ADP, but can be re-engineered into an ATP-dependent autophosphorylating enzyme. Also relevant to the yeast Coq8p ATPase.	
Pseudo-Histidine kinase	Protein interaction domain	Caulobacter DivL binds the response regulator, DivK, to regulate asymmetric cell division	[93]
Pseudophosphatase	Occlusion of conventional phosphatase access to substrate and subcellular localisation	STYX binds ERK1/2 kinase to occlude DUSP4 binding and anchors ERK1/2 in the nucleus	[23]
	Regulation of protein localisation in a cell	MTMR13 stabilizes lipid phosphatase MTMR2 and localizes MTMR2 to membranes	[94]
	Regulation of signalling complex assembly	STYX prevents substrate recruitment via FBXW7 to SCF E3 ligase complex	[95]
	Redox sensor as part of tandem enzyme:pseudoenzyme domains.	A Receptor Protein Tyrosine Phosphatase (RPTP) alpha redox-active Cys residue in the D2 (pseudophosphatase) domain controls the catalytic output of the D1 (canonical phosphatase) domain after exposure to ROS	[96]
Pseudoprotease	Allosteric regulator of conventional protease	cFLIP binds the cysteine protease, Caspase-8, to block apoptosis	[97]
	Regulation of protein localisation in a cell	Mammalian iRhom proteins bind and regulate membrane trafficking of receptors	[98, 99]
	Anoxygenic photosynthesis?	Mulitple pseudo-orthocaspases disclosed (prokaryotic)	[100]
	Activation of the prohenoloxidase cascade	Noncatalytic clip-domain SP family member PPAF-II from <i>H. diomphalia</i> binds and activates processed pro-	[101]
	Prevention of zymogen activation	phenoloxidases PPO1 and PPO2 Caspase-like pseudoproteases csp-2 and csp-3 prevent ced- 3 autoactivation in <i>C. elegans</i>	[102, 103]

Pseudodeubiquitinase (pseudoDUB)	Allosteric regulator of conventional DUB	Abraxas1 and 2 nucleate assembly of a higher order heterotetramer with BRCC36 active DUB (see also Fig. 1)	[104]
	Allosteric regulator of conventional enzyme	CSN6 supports activity of CSN5 and acts as a scaffold for the Cop9 signallosome complex	[105]
	Allosteric regulator of conventional DUB	PSMD7 (Rpn8 in yeast) supports the activity of the active DUB PSMD14 (Rpn11). Part of the 19S regulatory particle of the proteasome	[106, 107]
	Scaffold for assembly of enzyme complexes	PRP8 (Prpf8 in yeast) is central to the assembly of the spliceosome complex required for mRNA splicing. Prp8 contains three additional pseudoenzyme domains (pseudo-endonuclease, pseudo-reverse transcriptase, pseudo-RNaseH)	[16]
	Scaffold for assembly of enzyme complexes	USP39 (Sad1 in yeast) is a member of the spliceosome complex with presumed scaffolding roles	[16]
	Scaffold for assembly of enzyme complexes	elF3f and elF3h are both inactive and integral for the assembly of elF3 complex required for protein synthesis	[16]
	Scaffold for assembly of enzyme complexes	USP52 (Pan2 in yeast) serves as a scaffold for the Pan2- Pan3 deadenylating complex	[16, 108]

	Unknown function	FAM105A is an OTU domain containing protein that is closest to OTULIN (a.k.a. FAM105B or Gumby). Likely functions as a protein interactin domain the ER/Golgi. The MINDY4b pseudoenzyme is also part of the broader MINDY DUB family.	[16, 109]
Pseudo Ubiqutin Conjugating enzyme (pseudo-E2)	Allosteric regulator of conventional E2 ligase	The ubiquitin E2 variant (UEV) domain of Mms2 binds the active E2, Ubc13, to enable catalytic activity and dictate K63-linked ubiquitin linkages	[104]
	Regulation of protein localization in a cell	The Tsg101 UEV domain exerts a chaperone function, including during viral replication	[110]
	Scaffold for assembly of signaling complexes	BRCC45 contains three tandem UEV domains which support protein-protein interactions within the BRISC-SHMT2 complex required for IFN signalling and the BRCA1-A complex required for DNA damage repair	[111]
Pseudonuclease	Allosteric regulator of conventional nuclease	CPSF-100 is a component of the pre-mRNA 3' end processing complex containing the active counterpart, CPSF-73	[112]
	Nucleating and regulating assembly of protein:nucleic acid complexes	TEFM contributes to the elongation complex sliding clamp and sequesters ssDNA to maintain the mitochondrial transcription bubble	[113]
	Regulation of substrate localization	Exuperantia pseudo-exonuclease domain regulates cellular localization of the <i>bicoid</i> mRNA in <i>Drosophila</i> cells	[114]
PseudoATPase	Allosteric regulator of conventional ATPase	EccC comprises two pseudoATPase domains that regulate the N-terminal conventional ATPase domain	[115]

PseudoGTPase	Allosteric regulator of conventional GTPase	GTP-bound Rnd1 or Rnd3/RhoE bind p190RhoGAP to regulate the catalytic activity of the GTPase, RhoA	[116, 117]
	Allosteric control of conventional enzyme function	Three GTPase-like domains promote p190RhoGAP GAP activity	[118, 119]
	Scaffold for assembly of signaling complexes	CENP-M cannot bind GTP or switch conformations, but is essential for regulating kinetochore assembly	[120]
	Regulation of protein localization in a cell	Yeast light intermediate domain (LIC) is a pseudoGTPase, devoid of nucleotide binding, which connects the dynein motor protein to cargo proteins	[121]
Pseudochitinase	Substrate recruitment or sequestration	YKL-39 binds chitooligosaccharides via 5 binding subsites, but does not process these as substrates	[122, 123]
Pseudosialidase	Scaffold for assembly of signalling complexes	<i>P. falciparum</i> CyRPA nucleates assembly of the PfRh5/PfRipr complex that binds the host erythrocyte receptor, basigin, and mediates host cell invasion	[124]
Pseudoglycosidase	Scaffold for assembly of signalling complexes	β -Klotho tandem pseudoglycosidase domains act as a receptor for the hormone FGF21	[125]
Pseudolyase	Allosteric activation of conventional enzyme counterpart	Dead paralog binding to S-adenosylmethionine decarboxylase displaces an N-terminal inhibitory segment to activate catalytic activity	[126-128]
Pseudotransferase	Allosteric activation of cellular enzyme counterpart	Viral GAT recruits cellular phosphoribosyl- formylglycinamidine synthetase to deaminate RIG-I to block host antiviral defence	[129]
	Assembly of catalytically-active enzyme via composite active site	A composite active site is formed by dead and catalytic paralogs of <i>T. brucei</i> deoxyhypusine synthase heterotetramerize	[130, 131]

	Assembly of mitochondrial fission factor	MiD51 is an Inactive nucloetidyltransferase dynamin receptor that binds nucleotide diphosphates, and assembles Drp1, a regulator of mitochondrial fission	[132]
	Pantothenate kinase	Eukaryotic 'bifunctional' pantothenate kinase 4 (PANK4) phosphotransferases contain an N-terminal pantothenate pseudokinase domain fused to a phosphatase domain. Human PANK domain of PANK4 lacks catalytic Glu and Arg residues that are found in the PANK1-3 enzymes.	[133]
Pseudosulfotransferase	Unknown neuronal function?	Eukaryotic enzymatically-inactive soluble sulfotransferase (SULT4A1) fails to bind <i>in vitro</i> to the universal sulfate donor PAPS, or to sulfate a standard SULT substrate	[134]
Pseudo-histone acetyl transferase (pseudoHAT)	Possible scaffold for assembly of signalling complexes	Unlike bacterial <i>O. granulosus</i> counterpart, human C- terminal domain of putative bifunctional glycoside hydrolase and histone acetyltransferase (AT) O-GlcNAcase (OGA) lacks catalytic residues and detectable acetyl CoA binding	[135]
Pseudophospholipase	Scaffold for assembly of signalling complexes	DUF1669 pseudophospholipase D-like domains in FAM83 family proteins bind and regulate the canonical kinase CK1	[136, 137]
	Allosteric inactivation of conventional enzyme counterpart	Viper phospholipase A2 inhibitor attenuates the toxicity of the active phospholipase A2 paralog via heterodimerization	[138]
Pseudooxidoreductase	Allosteric inactivation of conventional enzyme counterpart	ALDH2*2 sequesters the active counterpart, ALDH2*1, to prevent its assembly into the tetrameric catalytic form	[139]
Pseudodismutase	Competition for enzyme's chaperone	Poxviral superoxide dismutase (SOD)-like proteins bind and sequester Copper chaperone for superoxide dismutase (CCS) to block activation of cellular SOD1	[140]
Pseudohydrolase	Integrin activation by protein with cryptic Nucleoside diiphosphate (Nudix) domain	KRIT1 antagonizes ICAP1 (negative regulator of integrin function) through a competitive mechanism. N-terminus of KRIT1 contains Nudix pseudoenzyme domain lacking canonical 'Nudix box'	[141]

	Pseudooligonucleotide synthetase	Anti-viral factor, loss induces resistance to West Nile Virus	Enzymatically-inactive Murine Oas1b (a 2'-5'-oligoadenylate synthease 1 paralog) possesses RNAse L-independent anti- viral properties towards flavivirus in mice.	[142, 143]
1092				
1093				
1094				