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dbPAF: an integrative database of protein phosphorylation in animals and fungi

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Protein phosphorylation is one of the most important post-translational modifications (PTMs) and regulates a broad spectrum of biological processes. Recent progresses in phosphoproteomic identifications have generated a flood of phosphorylation sites, while the integration of these sites is an urgent need. In this work, we developed a curated database of dbPAF, containing known phosphorylation sites in *H. sapiens*, *M. musculus*, *R. norvegicus*, *D. melanogaster*, *C. elegans*, *S. pombe* and *S. cerevisiae*. From the scientific literature and public databases, we totally collected and integrated 54,148 phosphoproteins with 483,001 phosphorylation sites. Multiple options were provided for accessing the data, while original references and other annotations were also present for each phosphoprotein. Based on the new data set, we computationally detected significantly over-represented sequence motifs around phosphorylation sites, predicted potential kinases that are responsible for the modification of collected phospho-sites, and evolutionarily analyzed phosphorylation conservation states across different species. Besides to be largely consistent with previous reports, our results also proposed new features of phospho-regulation. Taken together, our database can be useful for further analyses of protein phosphorylation in human and other model organisms. The dbPAF database was implemented in PHP + MySQL and freely available at <http://dbpaf.biocuckoo.org>.

As one of the most well-studied and ubiquitous post-translational modifications (PTMs) in proteins, phosphorylation is a conserved mechanism in both eukaryotes and prokaryotes, participates in almost all of biological processes, and reversibly determines the cellular dynamics and plasticity^{1–5}. In eukaryotes, phosphorylation mainly occurs on a subset of three types of amino acids, including serine, threonine and tyrosine residues. Phosphorylation events are dynamically but precisely regulated, and the core machinery of a phosphorylation system contains numerous protein kinases (PKs) for modifying proteins as “writers”, phospho-binding proteins for recognizing and interacting with phosphorylation sites as “readers”, and protein phosphatases (PPs) for demodifying substrates as “erasers”^{2,3,6,7}. The aberrant phosphorylation has been extensively reported to be highly associated with human diseases, such as cancers⁸, neurodegenerative diseases⁹ and metabolic disorders¹⁰. In this regard, the identification and functional analysis of phosphosites are fundamental for understanding the molecular mechanisms and regulatory roles of protein phosphorylation.

Recently, rapid progresses in high-throughput liquid chromatography/mass spectrometry (LC-MS) techniques have greatly advanced the identification of phosphorylation sites^{11–13}. For example, using a mouse model of multistage skin carcinogenesis, Zanivan *et al.* identified and quantified 3,457 proteins with 5,249 phosphorylation sites¹². Also, using a label-free technology, Lundby *et al.* totally quantified 31,480 phosphorylation sites in 7,280 proteins across 14 rat organs and tissues¹³. Recently, Humphrey *et al.* developed a new platform of EasyPhos, which can rapidly quantify over 10,000 phosphorylation sites from cell or tissue samples in a single run¹¹. Since a flood of sites have been generated, computational analysis of the phosphoproteomic data in a systematic level has also been boomed, such as the prediction of site-specific kinase-substrate relations (ssKSRs)^{14–16}, the re-construction and modeling of phosphorylation networks^{1,17}, the molecular evolutionary analysis of

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Database	Proteins	Sites	pS	pT	pY
dbPAF	54148	483001	318016	108615	56370
Phospho.ELM 9.0	10601	41070	30653	7232	3185
dbPTM 3.0	24601	147851	102310	28797	16744
PHOSIDA	15924	64118	51318	10782	2018
PhosphositePlus	37568	310779	194217	69647	46915
PhosphoPep 2.0	16234	75278	57762	13492	4024
PhosphoGRID 2.0	3121	19831	14871	4343	617
SysPTM 2.0	13867	54224	41875	9126	3223
HPRD 9	8280	51733	36052	11388	4293
UniProt ⁴	14904	50713	42189	6943	1581

Table 1. The comparison of dbPAF with other public databases. “For the UniProt database, only experimentally verified phospho-sites were considered, whereas the sites annotated with “By similarity”, “Potential” or “Probable” were not included.

phosphorylation signalings^{6,18}, and the identification of genetic variations that potentially change phosphorylation^{19,20}. Thus, we can expect that more and more experimental and computational studies will be carried out in the near future, to accumulatively enhance our knowledge on the phosphorylation.

Due to the data accumulation, the collection and integration of numerous phosphorylated substrates with their sites from different studies have emerged to be a great challenge. The first phosphorylation database, PhosphoBase, was developed in 1998 and only contained 398 experimentally identified phosphorylation sites at that time²¹. In 2004, this database was merged into a new database of Phospho.ELM^{22,23}, and its 9.0 version contained 41,070 phosphorylation sites in 10,601 substrates (Table 1). In 2006, two phosphorylation databases of dbPTM²⁴ and PHOSIDA²⁵ were released, and both databases were continuously maintained nearly ten years until now^{26,27}. To date, one of the most comprehensive and popular database is PhosphositePlus, which was initially established in 2004 and currently contained 310,779 phosphosites of 37,568 proteins^{28,29} (Table 1). In contrast to a general collection of phosphosites, several databases were mainly focused on specific species for a better annotation, because the number of known phosphorylation sites is still too limited in most of organisms. For example, PhosphoPep only maintained phosphorylation sites for 4 model organisms, including *H. sapiens*, *D. melanogaster*, *C. elegans* and *S. cerevisiae*^{30,31}, while PhosphoGRID only collected known phosphosites in *S. cerevisiae*^{32,33}. PHOSIDA was also organized in a species-specific manner, containing phosphorylation sites in nine prokaryotic or eukaryotic organisms^{25,26}. In addition, although several databases, such as SysPTM^{34,35}, HPRD³⁶ and UniProt³⁷ were constructed for a more general purpose, they also maintained a large number of phosphorylation sites.

During the past decade, we also collected and curated phosphorylation sites beyond directly using known data sets from public databases. Previously, we developed two phosphorylation databases of dbPPT⁵ and dbPSP⁴ for plants and prokaryotes, respectively. Here we reported a new database of dbPAF for the phospho-sites in human, animals and fungi. From the scientific literature and public databases, we totally collected and integrated 483,001 known phosphorylation sites in 54,148 proteins from *H. sapiens*, *M. musculus*, *R. norvegicus*, *D. melanogaster*, *C. elegans*, *S. pombe* and *S. cerevisiae*. The detailed annotations together with original references were provided for each protein entry. Using the new data set, we detected significantly over-represented sequence motifs around phospho-serine (pS), phospho-threonine (pT) and phospho-tyrosine (pY) sites for each species, separately. From the results, we observed that the most significant pT motifs are highly conserved across different organisms, while pS sites prefer to occur at intrinsically disordered regions with low-complexity. We also predicted potential PKs for phospho-sites and demonstrated that different PK groups or families play different roles in the regulation of phosphorylation. In addition, we systematically analyzed the conservation states of phosphorylation sites, and observed that the phosphorylation is similarly conserved across different species. Taken together, the dbPAF database can serve as a useful resource for further studies of protein phosphorylation in human, animals and fungi.

Results

Database construction and data statistics. In this work, we mainly focused on the collection and integration of known phosphosites identified from large-scale phosphoproteomic studies, and the procedure for the implementation of dbPAF database was shown in Fig. 1. First, we searched the PubMed with multiple keywords, including “phosphoproteomics”, “phosphoproteomic” and “phosphoproteome”. All retrieved articles were carefully curated, and we directly took the identified phosphorylated proteins, peptides and sites from the supplementary materials published together with these manuscripts if available. Because only a handful of eukaryotic species had the enough phosphorylation information, here we only reserved phosphorylation sites in *H. sapiens*, *M. musculus*, *R. norvegicus*, *D. melanogaster*, *C. elegans*, *S. pombe* and *S. cerevisiae*. Totally, we collected 294,370 non-redundant phosphorylation sites of 40,432 proteins from 115 published papers (Supplementary Table S1). For each species, we mapped corresponding phosphorylated proteins to its proteome sequences downloaded from UniProt database³⁷, and the phosphorylation sites were exactly pinpointed (Fig. 1). The detailed annotations such as protein names, gene names, keywords, functional descriptions and sequence annotations of phosphoproteins were retrieved from UniProt and further integrated, while the original references of identified phosphorylation sites were also provided in dbPAF (Fig. 1).

Besides the manual curation of the literature, we also integrated known phosphorylation sites of the seven species from several public databases, including Phospho.ELM^{22,23}, dbPTM^{24,27}, PHOSIDA^{25,26}, PhosphositePlus^{28,29},

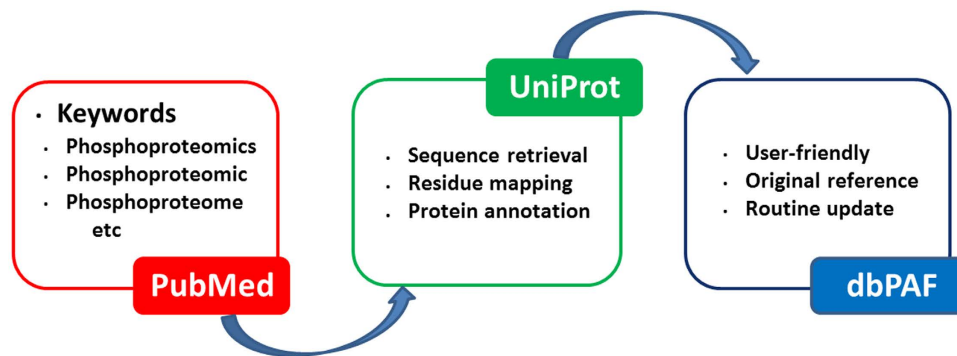


Figure 1. The procedure for the construction of dbPAF database. Also, we also integrated known phosphorylation sites from several public databases, including Phospho.ELM^{22,23}, dbPTM^{24,27}, PHOSIDA^{25,26}, PhosphositePlus^{28,29}, PhosphoPep^{30,31}, PhosphoGRID^{32,33}, SysPTM^{34,35}, HPRD³⁶ and UniProt³⁷.

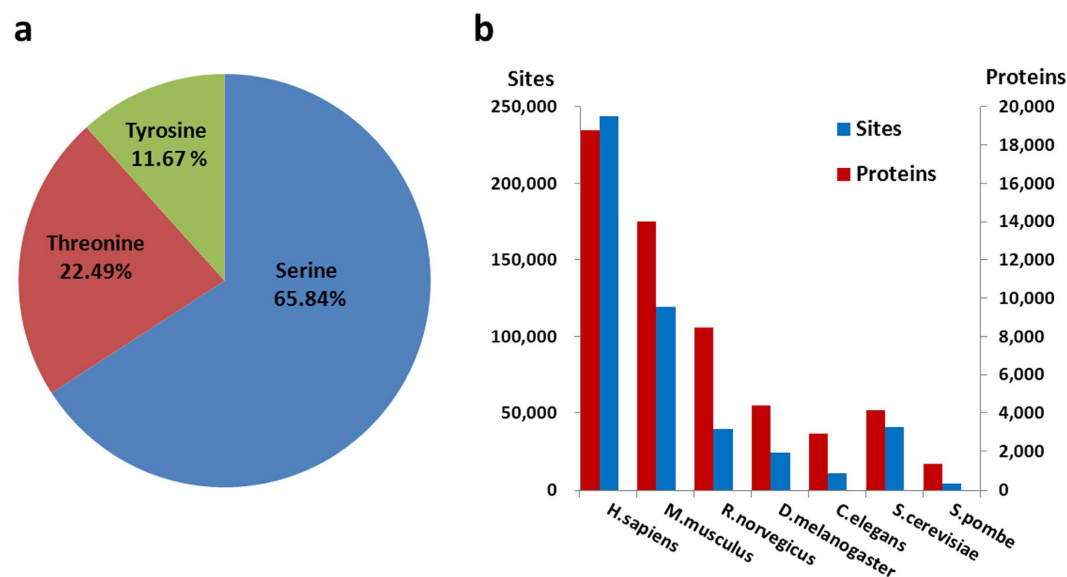


Figure 2. The data statistics of dbPAF. (a) The distribution of *p*S, *p*T and *p*Y residues. (b) The distribution of phosphorylated proteins and sites in each species.

PhosphoPep^{30,31}, PhosphoGRID^{32,33}, SysPTM^{34,35}, HPRD³⁶ and UniProt³⁷ (Table 1). For multiple entries with an identical phosphosite in the same protein, only one record was reserved. Finally, dbPAF contained 483,001 known phosphorylation sites of 54,148 proteins, as a comprehensive data resource for human, animals and fungi.

With the data set, we counted the distribution of phospho-sites for different amino acid types, and observed that the phosphorylation predominantly modifies *p*S residues (65.84%), while only 22.49% and 11.67% of phosphorylation events occur in *p*T and *p*Y sites, respectively (Fig. 2a). At the current stage, the phosphorylation events are not equally profiled in each species, whereas 50.52% (244,034 sites in 18,773 proteins) and 24.71% (119,328 sites in 14,044 proteins) of total phosphorylation sites were detected in *H. sapiens* and *M. musculus*, separately (Fig. 2b). In contrast, only 1,389 phosphorylated proteins with 3,957 sites were identified in *S. pombe*.

Usage of the dbPAF database. Our database was developed in an easy-to-use manner, and multiple options were provided to access the phosphorylation information. First, phosphorylated substrates in dbPAF can be browsed in a species-specific mode (Fig. 3a). Here we chose human peroxisomal alkylidihydroxyacetonephosphate synthase (AGPS) as an example to demonstrate the usage of “Browse by species”. After clicking on the species diagram of *H. sapiens* (Fig. 3a), all human phosphorylated proteins would be listed in a tabular format with “dbPAF ID”, “UniProt Accession”, “Species”, and “Protein Name” (Fig. 3b). A unique “dbPAF ID” was assigned to each protein for the convenient organization of dbPAF database. Then by clicking on the “dbPAF-0000003”, the detailed annotations including 18 known phospho-sites of human AGPS could be shown (Fig. 3c).

Besides the “Browse by species” option, our database provided up to four search options including “Substrate Search” (Fig. 4a), “Advanced search” (Fig. 4b), “Batch Search” (Fig. 4c), and “Blast Search” (Fig. 4d). For the “Substrate Search”, users can input one or multiple keywords, e.g. using “P53_HUMAN” and selecting “UniProt

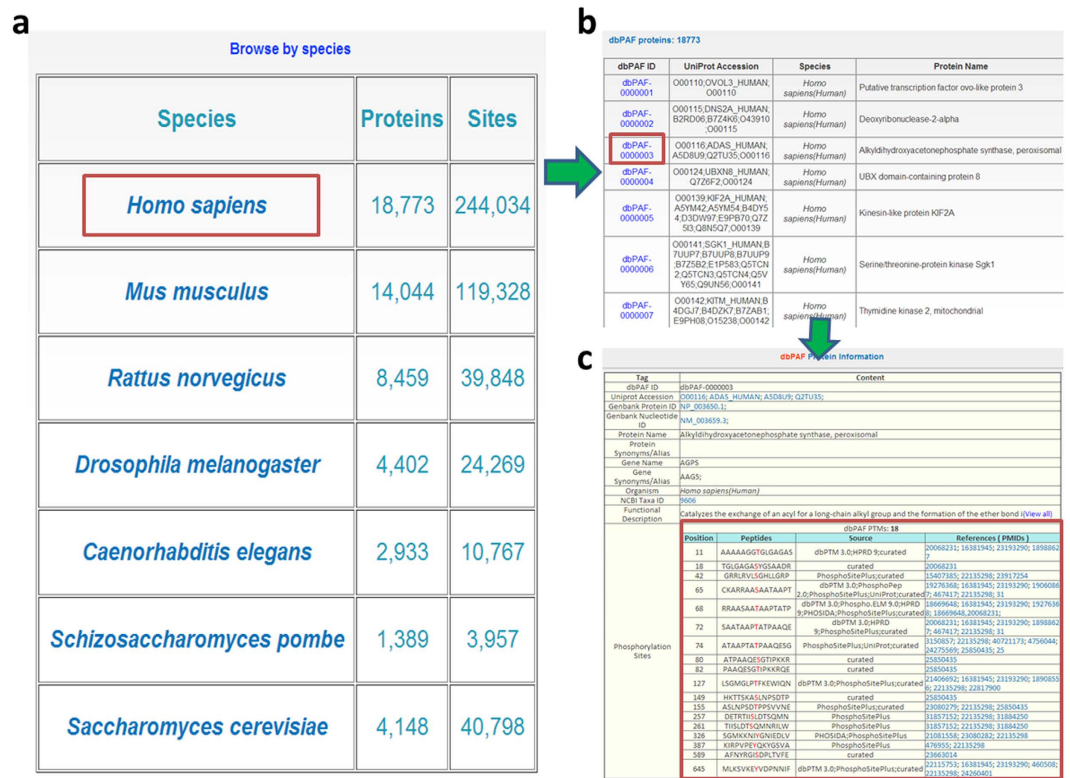


Figure 3. The “Browse by species” option. (a) Phosphorylated proteins can be browsed in a species-specific manner by clicking on the corresponding diagram. (b) The phosphorylated substrates will be listed in a tabular format. (c) The detailed annotations of human AGPS together with known phospho-sites.

Accession”, to find the phosphorylation information of human p53 (Fig. 4a). Also, users can choose the ‘Advanced Search’ using three terms together with three operators of “and”, “or” and “exclude”, to obtain a more exact hit (Fig. 4b). Moreover, “Batch Search” was present for retrieving multiple phosphoproteins with a list of keywords (Fig. 4c). Finally, “Blast Search” was implemented to find identical or homologous proteins using a protein sequence in FASTA Format. The NCBI BLAST package³⁸ was adopted for the sequence alignment (Fig. 4d).

The sequence preferences around phosphorylation sites. Previous studies demonstrated that short linear motifs around phosphorylation sites confer major specificity for the recognition, although a variety of cellular contextual factors, such as subcellular co-localization of PKs with their substrates, co-complex, or physical interaction, provide additional phosphorylation specificity *in vivo*^{1,7,15}. Using the data set in dbPAF, here we performed a motif-based analysis of sequence preferences around phosphorylation sites in *H. sapiens*, *M. musculus*, *D. melanogaster*, *C. elegans*, *S. pombe* and *S. cerevisiae* (Fig. 5). The motifs in *R. norvegicus* was not computationally detected, because human and mouse can be representative for mammals. In the results, the most significant motifs of pY sites are quite different across different organisms, although the motif pYS in *H. sapiens* is considerably similar with pYSP in *M. musculus* (Fig. 5). A simple interpretation is that tyrosine phosphorylation was evolved in metazoans³⁹, and there are no tyrosine kinases encoded in *S. pombe* and *S. cerevisiae*⁴⁰. Thus, tyrosine phosphorylation in yeasts might be mediated by dual-specificity protein kinases which can phosphorylate both serine/threonine and tyrosine sites⁴¹. Also, during the evolution, the number of tyrosine kinases was significantly increased. For example, there are only 32 tyrosine kinases in *D. melanogaster*, but human has up to 90 tyrosine kinases⁴⁰. Thus, different complexities of tyrosine phosphorylation in different species determine the distinct motif patterns. In contrast, the most significant motifs of pT sites in different species are quite similar (Fig. 5), and the results demonstrated a conserved mechanism for the threonine phosphorylation during evolution. Additionally, it was demonstrated that the phosphorylation sites preferentially occur in intrinsically disordered regions with low-complexity⁴². However, only most significant motifs of pS sites in *H. sapiens*, *M. musculus*, and *S. cerevisiae* follow the intrinsically disordered sequences, whereas pS motifs in other species are still informative (Fig. 5).

To further dissect sequence preferences of mammalian phosphorylation, we systematically predicted potential PKs for phosphorylation sites in *H. sapiens*, *M. musculus* and *R. norvegicus* (Fig. 6). A previously developed tool of GPS 2.1¹⁶ was used to predict ssKSRs in the family level (Fig. 6b) and then counted in the group level (Fig. 6a). In the kinase group level, the top four groups are AGC, CAMK, Other and CMGC kinase groups, which are responsible for the modification of about 70~75% of total phosphorylation events (Fig. 6a). In the family level, the top ten kinase families carry out about ~35% to 40% of total phosphorylation events (Fig. 6b). Thus, our results

a

UniProt Accession

dbPAF proteins: 4

dbPAF ID	UniProt Accession	Species	Protein Name
	P04637, P53, HUMAN, Q15086, Q15087, Q15088, Q15635, Q16807, Q16808, Q16809, Q16810, Q16811, Q16848, Q220098, Q3LRW4, Q3LRW5, Q8BU01, Q8BU16, Q99659, Q9BTM4, Q9HAC2, Q9HPC8, Q9HPC2, Q9NZD0, Q9UBI2, Q9UQ61.	Homo sapiens(Human)	Cellular tumor antigen p53
dbPAF-0004208			
dbPAF-0025193	Q70EX8, UBP53, HUMAN, Q8BDAS, Q8VWQ5, Q9P2J7.	Homo sapiens(Human)	Inactive ubiquitin carboxyl-terminal hydrolase 53

b

Gene Name exclude

Protein Alias exclude

Species exclude

dbPAF proteins: 1

dbPAF ID	UniProt Accession	Species	Protein Name
dbPAF-0004208	P04637, P53, HUMAN, Q15086, Q15087, Q15088, Q16635, Q16807, Q16808, Q16809, Q16810, Q16811, Q16848, Q220098, Q3LRW1, Q3LRW2, Q3LRW3, Q3LRW4, Q3LRW5, Q8BU01, Q8BU16, Q99659, Q9BTM4, Q9HAC2, Q9HPC8, Q9HPC2, Q9NZD0, Q9UBI2, Q9UQ61.	Homo sapiens(Human)	Cellular tumor antigen p53

c

Batch search:

Please input Multiple keywords line-by-line to retrieve a list of hits:

UniProt Accession

dbPAF proteins: 3

dbPAF ID	UniProt Accession	Species	Protein Name
	P04637, P53, HUMAN, Q15086, Q15087, Q15088, Q16635, Q16807, Q16808, Q16809, Q16810, Q16811, Q16848, Q220098, Q3LRW1, Q3LRW2, Q3LRW3, Q3LRW4, Q3LRW5, Q8BU01, Q8BU16, Q99659, Q9BTM4, Q9HAC2, Q9HPC8, Q9HPC2, Q9NZD0, Q9UBI2, Q9UQ61.	Homo sapiens(Human)	Cellular tumor antigen p53
dbPAF-0004208			
dbPAF-0044454	ABA223PXF8, Y066, YEAST.	Saccharomyces cerevisiae (strain ATCC 24848 / S288c)(Baker's yeast)	Putative uncharacterized membrane protein YL066W-A
dbPAF-0052008	F8VWH10, F8VWH10, MOUSE.	Mus musculus(Mouse)	Vascular endothelial growth factor A

d

BLAST search:

Please input a PROTEIN sequence in FASTA format:

dbPAF proteins: 10

dbPAF ID	Gene Name	Species	Identity	E-value	Score (bits)
dbPAF-0004208	TP53	Homo sapiens(Human)	100.00%	0.0	751
dbPAF-0005331	Tp53	Rattus norvegicus(Rat)	75.51%	1e-164	577
dbPAF-0003951	Tp53	Mus musculus(Mouse)	77.35%	9e-164	574
dbPAF-0053340	Trp53	Mus musculus(Mouse)	77.38%	9e-152	534
dbPAF-0000933	Tp73	Homo sapiens(Human)	50.94%	1e-72	271
dbPAF-0044967	Tp73	Homo sapiens(Human)	50.94%	2e-72	271
dbPAF-0035094	Tp73	Mus musculus(Mouse)	49.51%	4e-72	270
dbPAF-0037315	Tp63	Homo sapiens(Human)	50.00%	5e-72	269
dbPAF-0003169	Tp63	Mus musculus(Mouse)	50.00%	6e-72	269
dbPAF-0028613	p53	Drosophila melanogaster(Fruit fly)	24.31%	2e-08	58.5

Figure 4. The search options. (a) “Substrate Search” with one or multiple keywords. (b) The “Advanced search” permitted users to input up to three terms for query. (c) The “Batch Search” for retrieving multiple protein entries with a list of terms. (d) The database can be searched with a protein sequence in FASTA format to find identical or homologous phosphoproteins.

demonstrated that different kinase groups or families play distinct roles in mammals, and combinatorially determine the phosphorylation preferences.

The conservation distributions of phosphorylation sites. In a previous study, Minguez *et al.* developed the Residue Conservation Score (RCS) to determine the conservation status of PTM sites⁴³. Because the overall conservation states of different proteins are not equal, they further used non-modified residues as a reference data set to normalize RCS and further calculated the relative RCS (rRCS) value for each PTM site. In their analysis, they totally collected ~93,000 phosphorylation sites, including 58,501, 20,880, 1,748, 1,951, 1,337 and 9,764 sites in *H. sapiens*, *M. musculus*, *R. norvegicus*, *D. melanogaster*, *C. elegans* and *S. cerevisiae*, respectively. Based on the data set, the average rRCS values were calculated as about 55~85%, exhibiting a strong difference across eukaryotes⁴³.

Here, with the newly integrated data set in dbPAF, we adopted the same procedure and re-performed the analysis of phosphorylation conservation in seven organisms (Fig. 7a). In contrast with the previously study⁴³, our average rRCS values ranged from 73.10% (in *S. pombe*) to 84.34% (in *R. norvegicus*), and there was not a significant difference across different organisms (Fig. 7b, Table 2). For example, the average rRCS values are about 55~60% in *D. melanogaster*, *C. elegans* and *S. cerevisiae*, while our results are 76.65%, 75.50% and 75.72%, respectively (Fig. 7b). However, our average rRCS values on *H. sapiens* and *M. musculus* are similar with the previous study (Table 2). Thus, the previous results might be biased for lower species due to the data limitation⁴³. When the data set was enlarged, each species shows a similar conservation status of phosphorylation. Also, for the pY conservation, we observed that the rRCS values in mammals are much higher, while the ones in lower species are <80% (Fig. 7b). Again, this is because the complexity of tyrosine phosphorylation regulation is much higher in mammals with more tyrosine kinases^{39,40}, and thus mammalian tyrosine phospho-sites undergo a stronger functional constraint during the evolution.

Discussion

Recent progresses in the development and improvement of high-throughput phosphoproteomic techniques have facilitated a rapid increase of the number of identified phosphorylated proteins and sites in animals, fungi, plants⁵ and prokaryotes⁴. Due to the data accumulation, computational analysis of the big data has also emerged to be an intriguing topic, in contrast with conventionally experimental assays^{1,6,14–20}. All of these studies are heavily dependent on a high quality data resource of phospho-sites. Although a number of public databases were

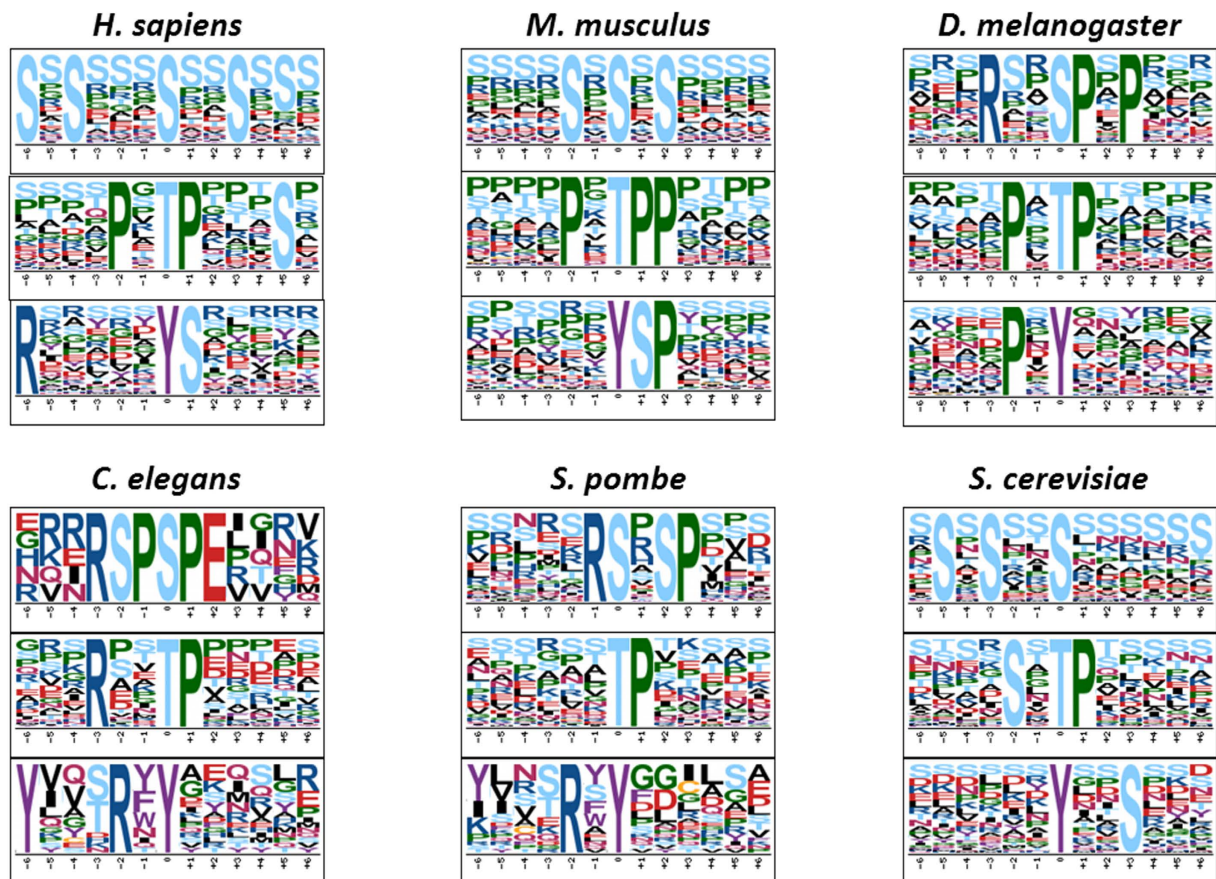


Figure 5. Motif-based analysis of sequence preferences around phosphorylation sites in *H. sapiens*, *M. musculus*, *D. melanogaster*, *C. elegans*, *S. pombe* and *S. cerevisiae*. In each species, the most significant motif was visualized for each type of phosphorylated residue. In the default threshold, we did not detect any *pY* motifs for *S. pombe*, due to the data limitation. However, when we slightly relaxed the stringency, a significant *pY* motif was detected, with a *p*-value < 0.00001.

developed^{22–37}, no one can collect and maintain all known phospho-sites. Also, several databases were developed for a more general purpose. For example, SysPTM 2.0 contained modification sites for 50 types of PTMs across 2031 species^{34,35}, whereas HPRD is one of the most useful resource for human proteins but not limited to PTMs³⁶. UniProt is a popular database for protein annotations, and PTMs are just one part of features³⁷. In addition, a considerable proportion of databases were developed for a general collection of phosphorylation sites, while only PhosphoPep^{30,31}, PhosphoGRID^{32,33} and PHOSIDA^{25,26} were constructed for specific organisms with a rich data set, since few phospho-sites were reported in most of species. Thus, the collection, integration, and annotation of phosphorylation sites in an organism-specific manner can be highly useful for further studies of phosphorylation in a specific species.

In this work, we reported a new database of dbPAF with 483,001 known phosphorylation sites in 54,148 proteins from human, animals and fungi. By comparison, our database contained more phospho-sites than other databases (Table 1). Because phosphorylated proteins, peptides and sites from different papers or databases might be differentially processed and annotated with distinct criteria, it's a great challenge to ensure the standardization of data quality. Although several pilot studies have been performed for the quality control of phosphoproteomic data sets, the experimental evidences, e.g., the use of phospho-specific antibodies to verify the existence of normalized phosphosites⁴⁴, yet remain to be provided. Because this study was mainly focused on the collection and integration of known phosphosites, any further normalization or standardization of the data set was not carried out. Based on the newly integrated data set, we carefully analyzed the sequence motifs around *pS*, *pT* and *pY* sites (Fig. 5), predicted potential PKs that are responsible for modifying the phosphorylation sites, and performed an analysis of the phosphorylation conservation status across different organisms. We anticipated that such a database can be a useful resource for further analyses. The database will be continuously updated and maintained when new phosphorylation sites are reported.

Methods

Motif-based analysis of phospho-sites. In this study, we chose Motif-X (<http://motif-x.med.harvard.edu/motif-x.html>), a widely-used online tool for detecting phosphorylation motifs from the phosphoproteomic data⁴⁵. For each species, known phosphorylated peptides in length of 13 with central characters of S/T/Y residues

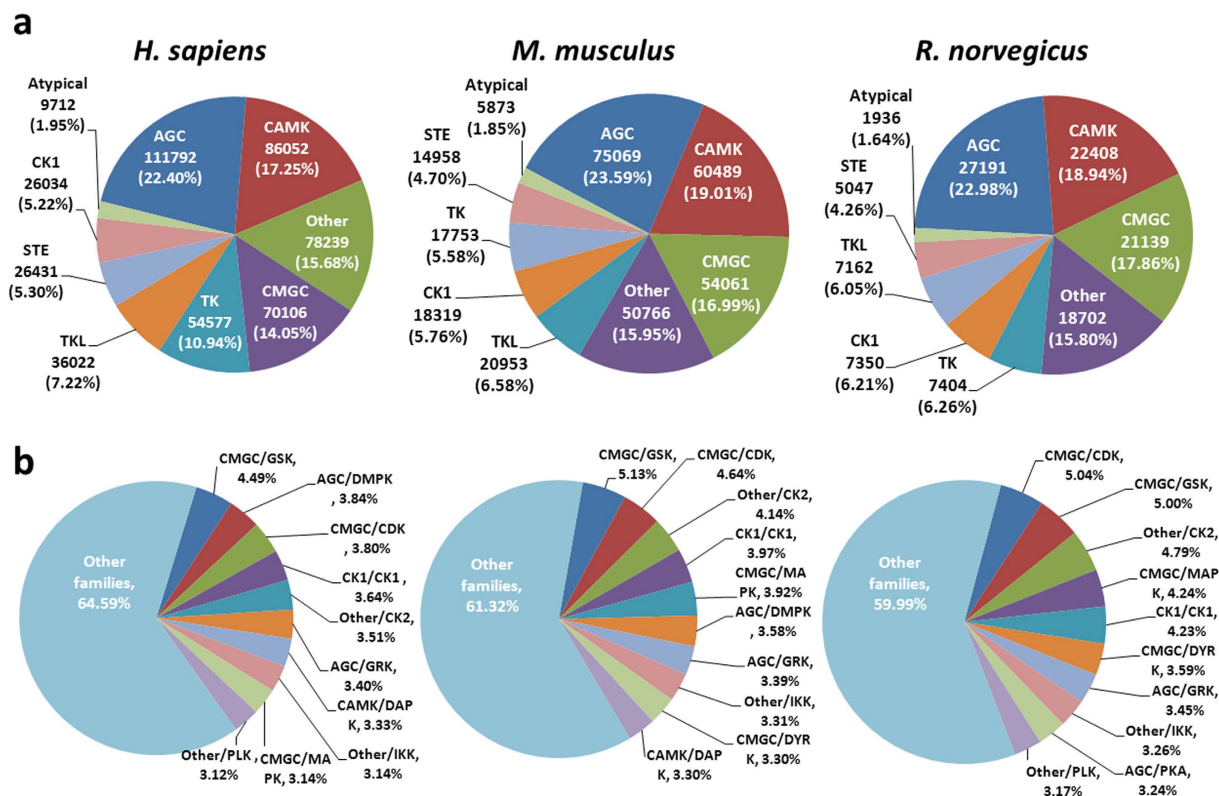


Figure 6. The distributions of PKs that were predicted to modify phospho-sites in *H. sapiens*, *M. musculus* and *R. norvegicus*, (a) in the kinase group level, and (b) in the kinase family level.

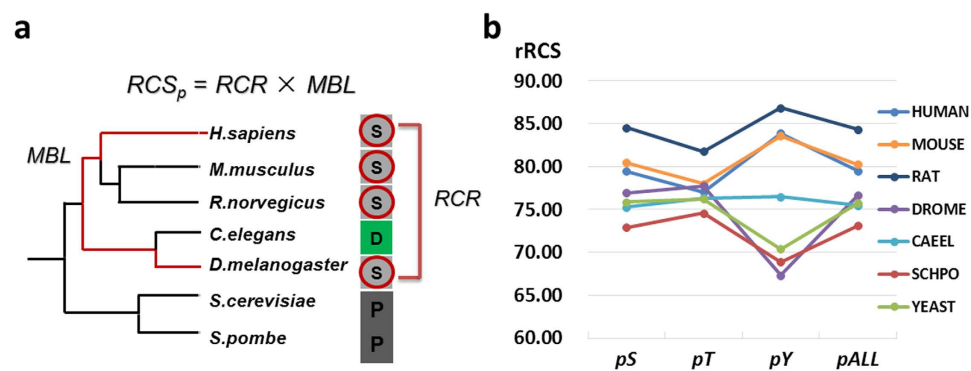


Figure 7. The evolutionary analysis of phosphorylation conservation. (a) As previously described⁴³, RCS_p was calculated as $RCR \times MBL$. (b) We further calculated rRCS values for phosphorylation sites, and the average rRCS values of pS, pT, pY and all sites were shown for each species.

Organism	pS		pT		pY		Final rRCS ^c
	RCS ^a	rRCS ^b	RCS	rRCS	RCS	rRCS	
<i>H. sapiens</i>	0.19	79.44	0.19	77.04	0.23	83.90	79.53
<i>M. musculus</i>	0.20	80.48	0.20	78.00	0.26	83.61	80.26
<i>R. norvegicus</i>	0.19	84.56	0.20	81.76	0.22	86.85	84.34
<i>D. melanogaster</i>	0.17	76.97	0.17	77.75	0.20	67.32	76.65
<i>C. elegans</i>	0.22	75.27	0.21	76.32	0.30	76.48	75.50
<i>S. pombe</i>	0.22	72.92	0.23	74.60	0.30	68.90	73.10
<i>S. cerevisiae</i>	0.19	75.91	0.19	76.23	0.24	70.38	75.72

Table 2. For each species, the conservation distribution were calculated for pS, pT and pY residues, respectively. ^aThe average RCS value; ^bThe average rRCS value; ^cThe final rRCS value.

were prepared as the foreground data set, while non-phosphorylated peptides in the same proteins were regarded as the background data set. The default parameters were adopted, with a p -value < 0.000001 . The phosphorylation motifs were calculated for pS , pT and pY sites, respectively. The most significant motifs of the three types of residues for each species were diagrammed in sequence logos (Fig. 5).

Prediction of kinase-specific phosphorylation sites in the kinase family level. Previously, we developed a software package of GPS 2.1¹⁶, in which the predictors were established based on the kinase classifications of mammals. Thus, here we only predicted potentially kinase-specific phospho-sites in *H. sapiens*, *M. musculus* and *R. norvegicus*, respectively. Also, because the prediction accuracy at the group level is limited, here we only predicted ssKSRs for known phosphorylation sites in the family level, and further counted in the group level. Totally, 42 and 20 predictors were selected from GPS 2.1 for serine/threonine kinases and tyrosine kinases, separately. The high thresholds were adopted for the prediction.

The calculation of RCS and rRCS. We analyzed the conservation distributions of phosphorylation sites in *H. sapiens*, *M. musculus*, *R. norvegicus*, *D. melanogaster*, *C. elegans*, *S. pombe* and *S. cerevisiae*, separately. First, the InParanoid 4.1 program⁴⁶ was obtained from Stockholm Bioinformatics Centre (<http://InParanoid.sbc.su.se>) for pairwise detecting orthologs among the seven species. All orthologous proteins among different species were multi-aligned by Clustal Omega (<http://www.clustal.org/omega/>)⁴⁷. As previously described⁴³, for each species, if a serine, threonine, or tyrosine residue in a multiple sequence alignment (MSA) is phosphorylated, the corresponding column was regarded as the modified position. Other columns containing serines, threonines or tyrosines were taken as the reference positions. To calculate the conservation of a phosphorylated site from a MSA, we adopted a previously defined RCS_p ⁴³, which can be calculated as below:

$$RCS_p = RCR * MBL = \frac{N_p}{N} * MBL \quad (1)$$

The N_p/N was defined as the Residue Conservation Ratio (RCR)⁴³. The N is the number of sequences with the maximum branch length (MBL), which is the maximum branch distance between any two species that contain a conserved phospho-site. The N_p is the number of phospho-sites observed in the column. Then the calculated RCS_p values of phosphorylated sites were mapped into the reference distribution to calculate the relative RCS or $rRCS$ ⁴³. The species tree of the seven organisms was taken from the Interactive Tree Of Life (iTOL, <http://itol.embl.de/>)⁴⁸.

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Author Contributions

Y. Xue and Z.L. designed and supervised experiments. S.U., S.L. and Y. Xu performed experiments and data analysis. S.L. developed the database. W.D., L.M. and Y.Z. contributed to data analysis. Y. Xue. wrote the manuscript with contributions of all authors. All authors reviewed the manuscript.

Additional Information

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