Supporting Information

UniPTM: Multiple PTM site prediction on full-length protein sequence

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Dataset construction details



Figure S1: Distribution of sequence lengths in raw data.



Figure S2: Data scale comparison in PTMseq pre- and post-CD-HIT deduplication processing.

One-hot											
PTM type	Accuracy	Precision	Recall	F1	MCC	AUROC	AUPRC				
Phosphoserine	0.9060 ± 0.0090	0.4659 ± 0.0354	0.4721 ± 0.0399	0.4671 ± 0.0155	0.4168 ± 0.0180	0.8483 ± 0.0057	0.4588 ± 0.0173				
Phosphothreonine	0.9255 ± 0.0032	0.4556 ± 0.0395	0.4509 ± 0.0194	0.4526 ± 0.0260	0.4130 ± 0.0270	0.8531 ± 0.0070	0.4475 ± 0.0410				
Phosphotyrosine	0.8994 ± 0.0297	0.6205 ± 0.1204	0.6090 ± 0.0319	0.6077 ± 0.0501	0.5548 ± 0.0660	0.8766 ± 0.0184	0.6481 ± 0.0720				
Acetyllysine	0.9209 ± 0.0221	0.5766 ± 0.1414	0.4442 ± 0.0752	0.5001 ± 0.1021	0.4634 ± 0.1161	0.8197 ± 0.0368	0.5203 ± 0.1203				
Succinyllysine	0.8238 ± 0.0358	0.3636 ± 0.1209	0.3228 ± 0.1290	0.3188 ± 0.0671	0.2348 ± 0.0783	0.7370 ± 0.0477	0.3438 ± 0.0378				
Methylarginine	0.9235 ± 0.0148	0.3805 ± 0.0645	0.5027 ± 0.1203	0.4224 ± 0.0435	0.3932 ± 0.0409	0.8897 ± 0.0131	0.3573 ± 0.0742				
Methyllysine	0.9477 ± 0.0127	0.6448 ± 0.2528	0.4288 ± 0.0748	0.5035 ± 0.1193	0.4945 ± 0.1405	0.8354 ± 0.0726	0.5108 ± 0.1801				
Trimethyllysine	0.9547 ± 0.0193	0.6657 ± 0.3799	0.4495 ± 0.3246	0.5236 ± 0.3366	0.5257 ± 0.3310	0.8041 ± 0.1611	0.5677 ± 0.3061				
Dimethyllysine	0.9462 ± 0.0199	0.7372 ± 0.0549	0.5923 ± 0.1405	0.6474 ± 0.0956	0.6281 ± 0.0867	0.8672 ± 0.0605	0.6901 ± 0.0805				
ProtBert											
PTM type	Accuracy	Precision	Recall	F1	MCC	AUROC	AUPRC				
Phosphoserine	0.9342 ± 0.0076	0.6403 ± 0.0576	0.5890 ± 0.0570	0.6095 ± 0.0165	0.5768 ± 0.0168	0.8957 ± 0.0074	0.6426 ± 0.0183				
Phosphothreonine	0.9301 ± 0.0161	0.5146 ± 0.1115	0.6017 ± 0.0530	0.5446 ± 0.0475	0.5152 ± 0.0459	0.8868 ± 0.0036	0.5679 ± 0.0376				
Phosphotyrosine	0.9290 ± 0.0121	0.7183 ± 0.0475	0.7098 ± 0.0426	0.7137 ± 0.0418	0.6734 ± 0.0476	0.9227 ± 0.0122	0.7521 ± 0.0423				
Acetyllysine	0.9356 ± 0.0090	0.6487 ± 0.0871	0.5813 ± 0.0351	0.6124 ± 0.0582	0.5789 ± 0.0641	0.8626 ± 0.0156	0.6267 ± 0.0552				
Succinyllysine	0.9071 ± 0.0322	0.7171 ± 0.0993	0.5316 ± 0.0835	0.6084 ± 0.0802	0.5664 ± 0.0989	0.8547 ± 0.0437	0.6585 ± 0.0695				
Methylarginine	0.9464 ± 0.0111	0.5359 ± 0.0732	0.5739 ± 0.0628	0.5484 ± 0.0295	0.5236 ± 0.0254	0.9235 ± 0.0086	0.4879 ± 0.0836				
Methyllysine	0.9651 ± 0.0055	0.8228 ± 0.1419	0.5312 ± 0.1091	0.6418 ± 0.1093	0.6430 ± 0.1121	0.8812 ± 0.0628	0.6177 ± 0.1522				
Trimethyllysine	0.9676 ± 0.0116	0.9031 ± 0.1216	0.6428 ± 0.1530	0.7360 ± 0.1115	0.7389 ± 0.0976	0.9021 ± 0.0585	0.7331 ± 0.1551				
Dimethyllysine	0.9634 ± 0.0219	0.8984 ± 0.1334	0.6910 ± 0.0958	0.7744 ± 0.0840	0.7661 ± 0.0933	0.9282 ± 0.0297	0.8070 ± 0.0614				
						Conti	nued on next page				

Table S1: Results of 5-fold CV on entire data set before splitting	Ta	ble S1	1: R	esults	of	5-fold	CV	on	entire	data	set	before	splitting	
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ProtT5												
PTM type	Accuracy	Precision	Recall	F1	MCC	AUROC	AUPRC					
Phosphoserine	0.9354 ± 0.0022	0.6367 ± 0.0296	0.6080 ± 0.0291	0.6212 ± 0.0146	0.5866 ± 0.0153	0.9032 ± 0.0059	0.6490 ± 0.0185					
Phosphothreonine	0.9322 ± 0.0106	0.5121 ± 0.0945	0.5738 ± 0.0229	0.5383 ± 0.0542	0.5046 ± 0.0588	0.8888 ± 0.0100	0.5358 ± 0.0612					
Phosphotyrosine	0.9206 ± 0.0111	0.6606 ± 0.0530	0.7513 ± 0.0269	0.7022 ± 0.0360	0.6590 ± 0.0391	0.9180 ± 0.0117	0.7522 ± 0.0366					
Acetyllysine	0.9387 ± 0.0058	0.6738 ± 0.0481	0.5871 ± 0.0538	0.6256 ± 0.0370	0.5951 ± 0.0365	0.8816 ± 0.0183	0.6392 ± 0.0490					
Succinyllysine	0.8971 ± 0.0243	0.6435 ± 0.0467	0.5358 ± 0.0886	0.5812 ± 0.0606	0.5284 ± 0.0691	0.8589 ± 0.0356	0.6363 ± 0.0484					
Methylarginine	0.9440 ± 0.0126	0.5138 ± 0.0531	0.5793 ± 0.0784	0.5399 ± 0.0384	0.5139 ± 0.0363	0.9308 ± 0.0061	0.5135 ± 0.0613					
Methyllysine	0.9646 ± 0.0060	0.7843 ± 0.1290	0.5544 ± 0.1264	0.6465 ± 0.1188	0.6407 ± 0.1207	0.8850 ± 0.0599	0.6272 ± 0.1686					
Trimethyllysine	0.9705 ± 0.0139	0.9091 ± 0.1107	0.6759 ± 0.1696	0.7623 ± 0.1230	0.7632 ± 0.1143	0.9039 ± 0.0719	0.7491 ± 0.1515					
Dimethyllysine	0.9611 ± 0.0239	0.8458 ± 0.0887	0.7008 ± 0.1276	0.7613 ± 0.0946	0.7472 ± 0.1019	0.9082 ± 0.0356	0.7853 ± 0.0821					
	ESM-2											
PTM type	Accuracy	Precision	Recall	F1	MCC	AUROC	AUPRC					
Phosphoserine	0.9372 ± 0.0031	0.6513 ± 0.0301	0.6046 ± 0.0248	0.6265 ± 0.0170	0.5931 ± 0.0183	0.9019 ± 0.0061	0.6634 ± 0.0203					
Phosphothreonine	0.9264 ± 0.0070	0.4727 ± 0.0174	0.6338 ± 0.0279	0.5411 ± 0.0127	0.5086 ± 0.0155	0.8944 ± 0.0101	0.5592 ± 0.0191					
Phosphotyrosine	0.9321 ± 0.0062	0.7287 ± 0.0524	0.7249 ± 0.0396	0.7258 ± 0.0352	0.6877 ± 0.0376	0.9229 ± 0.0160	0.7694 ± 0.0457					
Acetyllysine	0.9400 ± 0.0054	0.6814 ± 0.0308	0.5837 ± 0.0703	0.6277 ± 0.0532	0.5980 ± 0.0519	0.8709 ± 0.0200	0.6295 ± 0.0628					
Succinyllysine	0.9087 ± 0.0223	0.7087 ± 0.0484	0.5567 ± 0.0942	0.6188 ± 0.0635	0.5766 ± 0.0665	0.8783 ± 0.0327	0.6764 ± 0.0499					
Methylarginine	0.9417 ± 0.0145	0.5156 ± 0.1095	0.5824 ± 0.1109	0.5310 ± 0.0297	0.5107 ± 0.0267	0.9239 ± 0.0059	0.4831 ± 0.0686					
Methyllysine	0.9639 ± 0.0099	0.7696 ± 0.1810	0.5577 ± 0.1305	0.6443 ± 0.1435	0.6362 ± 0.1510	0.8769 ± 0.0660	0.6186 ± 0.1817					
Trimethyllysine	0.9670 ± 0.0118	0.8711 ± 0.1143	0.6619 ± 0.1645	0.7360 ± 0.1129	0.7348 ± 0.0982	0.9073 ± 0.0715	0.7437 ± 0.1358					
Dimethyllysine	0.9606 ± 0.0246	0.8793 ± 0.1351	0.6862 ± 0.1296	0.7611 ± 0.0970	0.7521 ± 0.1056	0.9015 ± 0.0715	0.7781 ± 0.0882					

Table S1 – Continued from previous page

Transformer model and multi-head attention

In this research, we apply the encoder component of the transformer model as the foundation for the UniPTM framework. Transformers are a category of deep learning models that have made significant advancements in natural language processing (NLP)^{1,2} and have recently been utilized to model protein sequences.³ The encoder component learns the latent representation of the input sequence through the self-attention mechanism. The self-attention mechanism has been used in conjunction with sequential models, such as recurrent neural networks (RNN)⁴ and LSTM,⁵ to address declines in model performance when processing long sequences. The transformer model discards the recurrent architecture found in earlier sequential models and depends entirely on the self-attention mechanism to learn input sequence's representations. The benefits of this model architecture are various. Self-attention reduces the computational complexity per layer, allowing for faster training as the model relies solely on self-attention, which primarily involves matrix multiplications. Additionally, the self-attention mechanism effectively facilitates learning long-range dependencies within the input sequence.

The weights used for linearly projecting the entire input sequence are represented by the matrices W_q , W_k , and W_v . In this research, we applied the scaled dot-product attention:

Attention
$$(Q, K, V) = \operatorname{softmax}\left(\frac{QK^{\top}}{\sqrt{d_k}}\right)V$$
 (1)

Multi-head attention involves using several linear projection matrices simultaneously to compute attention. In multi-head attention, the matrices W_q , W_k , and W_v are initialized differently for each head, allowing the model to capture various representations of the input sequence from different subspaces. If we assume that there are h attention heads, the multihead attention can be computed using the following formula:

$$Multi-Head(Q, K, V) = Concat(head_1, head_2, \dots, head_h)W^O$$
(2)

where each $head_i$ is computed as:

$$head_i = Attention(QW_i^Q, KW_i^K, VW_i^V)$$
(3)

Here, W_i^Q , W_i^K , and W_i^V are the parameter matrices for the *i*-th head; and W^O is the output projection matrix that combines the outputs of all heads. In this study, $d_{\text{emb}} = 256$ and the number of head h = 8. The dimension of the W_i^Q , W_i^K , and W_v^i linear projection matrices is $W_i^Q \in \mathbb{R}^{d_{emb} \times d_k}$, $W_i^Q \in \mathbb{R}^{d_{emb} \times d_k}$, and $W_i^V \in \mathbb{R}^{d_{emb} \times d_k}$, where $d_{\text{emb}} = 256$ and $d_v = d_k = d_{emb}/h = 32$. Moreover, $W^O \in \mathbb{R}^{d_{emb} \times d_{emb}}$.

Training details

In this section, we provide detailed descriptions of the hyperparameters (Table S1) used for training the UniPTM model, along with the equations (C.4-C.8) for the five evaluation criteria.

Table S2: Hyperparameters for the UniPTM model training

Batch Size	lr	Optimizer	Epochs	Emb size	$\mathrm{Train}/\mathrm{Val}$	Weight Decay	dropout rate	pos_weight
32	5e-5	Adam	200	1024/1280	0.9/0.1	1e-5	0.5	3

Accuracy =
$$\frac{TP + TN}{TP + TN + FP + FN}$$
 (4)

$$Precision = \frac{TP}{TP + FP}$$
(5)

$$\text{Recall} = \frac{TP}{TP + FN} \tag{6}$$

F1 score
$$= \frac{2 \times \text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}}$$
 (7)

$$MCC = \frac{TP \cdot TN - FP \cdot FN}{\sqrt{(TP + FP)(TP + FN)(TN + FP)(TN + FN)}}$$
(8)

Model evaluation results

One-hot											
PTM type	Accuracy	Precision	Recall	F1	MCC	AUROC	AUPRC				
Phosphoserine	0.9076 ± 0.0081	0.4712 ± 0.0396	0.4168 ± 0.0233	0.4419 ± 0.0279	0.3928 ± 0.0328	0.8367 ± 0.0078	0.4367 ± 0.0353				
Phosphothreonine	0.9176 ± 0.0123	0.3953 ± 0.0686	0.3889 ± 0.0583	0.3916 ± 0.0618	0.3477 ± 0.0675	0.8298 ± 0.0172	0.3751 ± 0.0764				
Phosphotyrosine	0.8838 ± 0.0190	0.5341 ± 0.0744	0.5735 ± 0.0521	0.5513 ± 0.0536	0.4863 ± 0.0629	0.8482 ± 0.0200	0.5749 ± 0.0695				
Acetyllysine	0.9039 ± 0.0103	0.4367 ± 0.0488	0.3863 ± 0.0581	0.4085 ± 0.0470	0.3581 ± 0.0521	0.7912 ± 0.0229	0.4088 ± 0.0601				
Succinyllysine	0.8431 ± 0.0381	0.1963 ± 0.1795	0.1789 ± 0.1755	0.1846 ± 0.1735	0.1271 ± 0.1182	0.6601 ± 0.0838	0.2644 ± 0.0711				
Methylarginine	0.9276 ± 0.0167	0.3918 ± 0.0750	0.4383 ± 0.0653	0.4074 ± 0.0467	0.3736 ± 0.0472	0.8702 ± 0.0200	0.3519 ± 0.0909				
Methyllysine	0.9443 ± 0.0161	0.5517 ± 0.3704	0.3241 ± 0.1880	0.4007 ± 0.2431	0.3982 ± 0.2509	0.8119 ± 0.0529	0.4908 ± 0.1251				
Trimethyllysine	0.9516 ± 0.0236	0.6795 ± 0.3815	0.4260 ± 0.2419	0.5219 ± 0.2926	0.5214 ± 0.2921	0.8309 ± 0.1145	0.6348 ± 0.1680				
Dimethyllysine	0.9348 ± 0.0254	0.7324 ± 0.4237	0.2807 ± 0.2282	0.3776 ± 0.2714	0.4154 ± 0.2614	0.7980 ± 0.1060	0.5633 ± 0.1335				
ProtBert											
PTM type	Accuracy	Precision	Recall	F1	MCC	AUROC	AUPRC				
Phosphoserine	0.9331 ± 0.0039	0.6416 ± 0.0306	0.5420 ± 0.0262	0.5867 ± 0.0090	0.5534 ± 0.0100	0.8837 ± 0.0043	0.6126 ± 0.0141				
Phosphothreonine	0.9331 ± 0.0069	0.5088 ± 0.0471	0.5132 ± 0.0426	0.5101 ± 0.0370	0.4747 ± 0.0402	0.8744 ± 0.0045	0.5048 ± 0.0494				
Phosphotyrosine	0.9108 ± 0.0052	0.6250 ± 0.0436	0.6994 ± 0.0219	0.6595 ± 0.0292	0.6101 ± 0.0307	0.9097 ± 0.0160	0.7076 ± 0.0286				
Acetyllysine	0.9278 ± 0.0081	0.5985 ± 0.0547	0.5059 ± 0.0605	0.5461 ± 0.0472	0.5108 ± 0.0493	0.8348 ± 0.0234	0.5595 ± 0.0491				
Succinyllysine	0.8961 ± 0.0286	0.7613 ± 0.1427	0.3465 ± 0.1177	0.4609 ± 0.1217	0.4582 ± 0.0978	0.8154 ± 0.0371	0.5647 ± 0.0645				
Methylarginine	0.9456 ± 0.0125	0.5239 ± 0.0623	0.4813 ± 0.1122	0.4969 ± 0.0784	0.4714 ± 0.0788	0.9028 ± 0.0235	0.4331 ± 0.0941				
Methyllysine	0.9575 ± 0.0108	0.7642 ± 0.1417	0.4957 ± 0.0836	0.6006 ± 0.1020	0.5948 ± 0.1116	0.8683 ± 0.0507	0.5923 ± 0.1395				
Trimethyllysine	$0.9\overline{678} \pm 0.0163$	0.8800 ± 0.1460	0.6613 ± 0.1582	0.7443 ± 0.1138	0.7418 ± 0.1182	0.8865 ± 0.0725	0.7551 ± 0.1448				
Dimethyllysine	0.9553 ± 0.0200	0.8556 ± 0.1089	0.6013 ± 0.1882	0.6819 ± 0.1241	0.6831 ± 0.0954	0.8566 ± 0.0927	0.7122 ± 0.1416				
						Conti	nued on next page				

Table S3:	Results	of 5-fold	CV	on	training	data.

$\operatorname{ProtT5}$													
PTM type	Accuracy	Precision	Recall	F1	MCC	AUROC	AUPRC						
Phosphoserine	0.9329 ± 0.0063	0.6403 ± 0.0577	0.5619 ± 0.0491	0.5948 ± 0.0035	0.5621 ± 0.0068	0.8923 ± 0.0043	0.6231 ± 0.0068						
Phosphothreonine	0.9327 ± 0.0101	0.5118 ± 0.0756	0.5944 ± 0.0306	0.5466 ± 0.0383	0.5143 ± 0.0406	0.8893 ± 0.0033	0.5489 ± 0.0487						
Phosphotyrosine	0.9127 ± 0.0078	0.6298 ± 0.0318	0.7089 ± 0.0584	0.6665 ± 0.0401	0.6182 ± 0.0444	0.9073 ± 0.0127	0.7001 ± 0.0479						
Acetyllysine	0.9210 ± 0.0045	0.5476 ± 0.0451	0.5247 ± 0.0546	0.5325 ± 0.0211	0.4917 ± 0.0202	0.8483 ± 0.0188	0.5364 ± 0.0318						
Succinyllysine	0.8838 ± 0.0265	0.6379 ± 0.1527	0.4082 ± 0.1498	0.4674 ± 0.1127	0.4335 ± 0.0704	0.8218 ± 0.0415	0.5560 ± 0.0692						
Methylarginine	0.9483 ± 0.0084	0.5475 ± 0.0666	0.4581 ± 0.0781	0.4976 ± 0.0712	0.4734 ± 0.0724	0.9109 ± 0.0275	0.4877 ± 0.0895						
Methyllysine	0.9559 ± 0.0097	0.7644 ± 0.1451	0.4689 ± 0.0647	0.5791 ± 0.0844	0.5768 ± 0.0957	0.8551 ± 0.0446	0.5865 ± 0.1310						
Trimethyllysine	0.9676 ± 0.0139	0.8942 ± 0.0769	0.6381 ± 0.1504	0.7341 ± 0.0915	0.7349 ± 0.0847	0.8994 ± 0.0791	0.7504 ± 0.1378						
Dimethyllysine	0.9418 ± 0.0232	0.8548 ± 0.1068	0.4266 ± 0.2142	0.5293 ± 0.1838	0.5551 ± 0.1295	0.8327 ± 0.1040	0.6357 ± 0.1412						
	ESM-2												
PTM type	Accuracy	Precision	Recall	F1	MCC	AUROC	AUPRC						
Phosphoserine	0.9338 ± 0.0055	0.6472 ± 0.0578	0.5597 ± 0.0486	0.5968 ± 0.0109	0.5647 ± 0.0117	0.8894 ± 0.0050	0.6291 ± 0.0094						
Phosphothreonine	0.9310 ± 0.0090	0.4981 ± 0.0524	0.5884 ± 0.0360	0.5375 ± 0.0293	0.5038 ± 0.0320	0.8846 ± 0.0060	0.5373 ± 0.0419						
Phosphotyrosine	0.9198 ± 0.0074	0.6736 ± 0.0436	0.6943 ± 0.0615	0.6812 ± 0.0209	0.6372 ± 0.0225	0.9150 ± 0.0181	0.7340 ± 0.0367						
Acetyllysine	0.9320 ± 0.0070	0.6452 ± 0.0248	0.4793 ± 0.0711	0.5467 ± 0.0426	0.5193 ± 0.0356	0.8491 ± 0.0184	0.5580 ± 0.0365						
Succinyllysine	0.8989 ± 0.0264	0.6973 ± 0.1027	0.4304 ± 0.0749	0.5289 ± 0.0688	0.4950 ± 0.0803	0.8449 ± 0.0323	0.5950 ± 0.0755						
Methylarginine	0.9459 ± 0.0074	0.5199 ± 0.0640	0.5089 ± 0.0929	0.5110 ± 0.0664	0.4844 ± 0.0683	0.9016 ± 0.0286	0.4566 ± 0.1121						
Methyllysine	0.9578 ± 0.0091	0.7661 ± 0.1249	0.4991 ± 0.0733	0.6035 ± 0.0880	0.5976 ± 0.0961	0.8521 ± 0.0572	0.5776 ± 0.1582						
Trimethyllysine	0.9687 ± 0.0174	0.9206 ± 0.0865	0.6571 ± 0.1528	0.7535 ± 0.0733	0.7565 ± 0.0689	0.8999 ± 0.0724	0.7600 ± 0.1390						
Dimethyllysine	0.9536 ± 0.0176	0.8350 ± 0.1073	0.5936 ± 0.1680	0.6745 ± 0.0974	0.6724 ± 0.0816	0.8468 ± 0.0814	0.6701 ± 0.1064						

Table S3 – Continued from previous page

PTM type	Accuracy	Precision	Recall	F1	MCC	AUROC	AUPRC
Phosphoserine	0.9389 ± 0.0015	0.6691 ± 0.0157	0.5910 ± 0.0197	0.6275 ± 0.0141	0.5958 ± 0.0143	0.9028 ± 0.0062	0.6592 ± 0.0176
MusiteDeep (S,T)	0.7528	0.2272	0.7915	0.3531	0.3305	0.8399	0.4278
Phosphothreonine	0.9385 ± 0.0048	0.5522 ± 0.0536	0.5827 ± 0.0461	0.5640 ± 0.0240	0.5330 ± 0.0234	0.8908 ± 0.0044	0.5736 ± 0.0310
MusiteDeep (S,T)	0.8901	0.3394	0.5884	0.4305	0.3919	0.8594	0.4378
Phosphotyrosine	0.9299 ± 0.0097	0.7127 ± 0.0596	0.7361 ± 0.0369	0.7230 ± 0.0383	0.6838 ± 0.0418	0.9241 ± 0.0178	0.7712 ± 0.0397
MusiteDeep (Y)	0.7843	0.3591	0.8772	0.5096	0.4668	0.9030	0.6487
Acetyllysine	0.9358 ± 0.0109	0.6500 ± 0.1010	0.5902 ± 0.0402	0.6170 ± 0.0658	0.5839 ± 0.0726	0.8739 ± 0.0194	0.6216 ± 0.0720
DeepAcet	0.5040	0.1081	0.7111	0.1877	0.1074	0.6249	0.1214
Succinyllysine	0.9019 ± 0.0242	0.6774 ± 0.0590	0.5328 ± 0.1071	0.5902 ± 0.0719	0.5447 ± 0.0756	0.8651 ± 0.0274	0.6395 ± 0.0541
LMSuccSite	0.4365	0.1854	0.8918	0.3070	0.1880	0.7102	0.2700
Methylarginine	0.9336 ± 0.0196	0.4645 ± 0.0880	0.6083 ± 0.1307	0.5111 ± 0.0250	0.4909 ± 0.0213	0.9261 ± 0.0116	0.4929 ± 0.0540
DeepRMethylSite	0.9191	0.2917	0.5283	0.3758	0.3534	0.9015	0.3108
Methyllysine	0.9603 ± 0.0117	0.7326 ± 0.2043	0.5467 ± 0.1134	0.6217 ± 0.1409	0.6108 ± 0.1514	0.8811 ± 0.0579	0.6290 ± 0.1491
DeepKme	0.6030	0.1380	0.7910	0.2160	0.2240	0.8100	0.2310
Trimethyllysine	0.9681 ± 0.0096	0.8771 ± 0.0931	0.6729 ± 0.1321	0.7501 ± 0.0758	0.7468 ± 0.0657	0.9072 ± 0.0699	0.7516 ± 0.1394
Dimethyllysine	0.9586 ± 0.0247	0.8465 ± 0.1223	0.6896 ± 0.1144	0.7518 ± 0.0842	0.7387 ± 0.0941	0.9091 ± 0.0428	0.7721 ± 0.0761

Table S4: Performance comparison of state-of-the-art models and UniPTM on independent testing set.

UniPTM mechanism visualization



Figure S3: Visualization of abstract features extracted by UniPTM and original site features by pre-trained ESM-2 model (Part I: phosphoserine, phosphothreosine, and phosphotyrosine). Colored dots represent positive site samples, which are the PTM residues in full-length protein sequences, and gray dots represent negative site samples, which are non-PTM residues.



Figure S4: Visualization of abstract features extracted by UniPTM and original site features by pre-trained ESM-2 model (Part II: acetyllysine, succinyllysine, and methylarginine). Colored dots represent positive site samples, which are the PTM residues in full-length protein sequences, and gray dots represent negative site samples, which are non-PTM residues.



Figure S5: Visualization of abstract features extracted by UniPTM and original site features by pre-trained ESM-2 model (Part III: methyllysine, trimethyllysine, and dimethyllysine). Colored dots represent positive site samples, which are the PTM residues in full-length protein sequences, and gray dots represent negative site samples, which are non-PTM residues.

References

- Vaswani, A.; Shazeer, N.; Parmar, N.; Uszkoreit, J.; Jones, L.; Gomez, A. N.; Kaiser, L.; Polosukhin, I. Attention is all you need. *Advances in neural information processing systems* 2017, 30.
- (2) Brown, T.; Mann, B.; Ryder, N.; Subbiah, M.; Kaplan, J. D.; Dhariwal, P.; Neelakantan, A.; Shyam, P.; Sastry, G.; Askell, A.; others Language models are few-shot learners. *Advances in neural information processing systems* **2020**, *33*, 1877–1901.
- (3) Chowdhury, R.; Bouatta, N.; Biswas, S.; Floristean, C.; Kharkar, A.; Roy, K.; Rochereau, C.; Ahdritz, G.; Zhang, J.; Church, G. M.; others Single-sequence protein structure prediction using a language model and deep learning. *Nature Biotechnology* 2022, 40, 1617–1623.
- (4) Bahdanau, D.; Cho, K.; Bengio, Y. Neural machine translation by jointly learning to align and translate. arXiv preprint arXiv:1409.0473 2014,
- (5) Hochreiter, S.; Schmidhuber, J. Long short-term memory. Neural computation 1997, 9, 1735–1780.