Shengli Zhang

List of Publications by Year in descending order

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SHENCU ZHANC

#	Article	IF	CITATIONS
1	Prediction of protein subcellular localization with oversampling approach and Chou's general PseAAC. Journal of Theoretical Biology, 2018, 437, 239-250.	1.7	81
2	A novel protein structural classes prediction method based on predicted secondary structure. Biochimie, 2012, 94, 1166-1171.	2.6	57
3	High-accuracy prediction of protein structural class for low-similarity sequences based on predicted secondary structure. Biochimie, 2011, 93, 710-714.	2.6	56
4	Using principal component analysis and support vector machine to predict protein structural class for low-similarity sequences via PSSM. Journal of Biomolecular Structure and Dynamics, 2012, 29, 1138-1146.	3.5	55
5	Predicting apoptosis protein subcellular localization by integrating auto-cross correlation and PSSM into Chou's PseAAC. Journal of Theoretical Biology, 2018, 457, 163-169.	1.7	53
6	Accurate prediction of protein structural classes by incorporating PSSS and PSSM into Chou's general PseAAC. Chemometrics and Intelligent Laboratory Systems, 2015, 142, 28-35.	3.5	45
7	Identifying DNase I hypersensitive sites using multi-features fusion and F-score features selection via Chou's 5-steps rule. Biophysical Chemistry, 2019, 253, 106227.	2.8	40
8	Identify Gram-negative bacterial secreted protein types by incorporating different modes of PSSM into Chou's general PseAAC via Kullback–Leibler divergence. Journal of Theoretical Biology, 2018, 454, 22-29.	1.7	36
9	Predict protein structural class by incorporating two different modes of evolutionary information into Chou's general pseudo amino acid composition. Journal of Molecular Graphics and Modelling, 2017, 78, 110-117.	2.4	33
10	iRSpot-DTS: Predict recombination spots by incorporating the dinucleotide-based spare-cross covariance information into Chou's pseudo components. Genomics, 2019, 111, 1760-1770.	2.9	27
11	Use of information discrepancy measure to compare protein secondary structures. Computational and Theoretical Chemistry, 2009, 909, 102-106.	1.5	21
12	Identifying <scp>DNA</scp> â€binding proteins based on multiâ€features and <scp>LASSO</scp> feature selection. Biopolymers, 2021, 112, e23419.	2.4	21
13	Improving the prediction accuracy of protein structural class: Approached with alternating word frequency and normalized Lempel–Ziv complexity. Journal of Theoretical Biology, 2014, 341, 71-77.	1.7	20
14	Use Chou's 5-steps rule to identify DNase I hypersensitive sites via dinucleotide property matrix and extreme gradient boosting. Molecular Genetics and Genomics, 2020, 295, 1431-1442.	2.1	19
15	Prediction of Protein Structural Classes for Low-Similarity Sequences Based on Consensus Sequence and Segmented PSSM. Computational and Mathematical Methods in Medicine, 2015, 2015, 1-9.	1.3	16
16	Prediction of apoptosis protein subcellular localization via heterogeneous features and hierarchical extreme learning machine. SAR and QSAR in Environmental Research, 2019, 30, 209-228.	2.2	16
17	KD-KLNMF: Identification of IncRNAs subcellular localization with multiple features and nonnegative matrix factorization. Analytical Biochemistry, 2020, 610, 113995.	2.4	16
18	A Complexity-based Method to Compare RNA Secondary Structures and its Application. Journal of Biomolecular Structure and Dynamics, 2010, 28, 247-258.	3.5	14

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19	PA-PseU: An incremental passive-aggressive based method for identifying RNA pseudouridine sites via Chou's 5-steps rule. Chemometrics and Intelligent Laboratory Systems, 2021, 210, 104250.	3.5	13
20	iDHS-DSAMS: Identifying DNase I hypersensitive sites based on the dinucleotide property matrix and ensemble bagged tree. Genomics, 2020, 112, 1282-1289.	2.9	12
21	UMAP-DBP: An Improved DNA-Binding Proteins Prediction Method Based on Uniform Manifold Approximation and Projection. Protein Journal, 2021, 40, 562-575.	1.6	12
22	Pep-CNN: An improved convolutional neural network for predicting therapeutic peptides. Chemometrics and Intelligent Laboratory Systems, 2022, 221, 104490.	3.5	12
23	Prediction of Apoptosis Protein's Subcellular Localization by Fusing Two Different Descriptors Based on Evolutionary Information. Acta Biotheoretica, 2018, 66, 61-78.	1.5	11
24	iPromoter-ET: Identifying promoters and their strength by extremely randomized trees-based feature selection. Analytical Biochemistry, 2021, 630, 114335.	2.4	11
25	i6mA-VC: A Multi-Classifier Voting Method for the Computational Identification of DNA N6-methyladenine Sites. Interdisciplinary Sciences, Computational Life Sciences, 2021, 13, 413-425.	3.6	10
26	Identification of amyloidogenic peptides via optimized integrated features space based on physicochemical properties and PSSM. Analytical Biochemistry, 2019, 583, 113362.	2.4	9
27	iEnhancer-MFGBDT: Identifying enhancers and their strength by fusing multiple features and gradient boosting decision tree. Mathematical Biosciences and Engineering, 2021, 18, 8797-8814.	1.9	9
28	Feature analysis of protein structure by using discrete Fourier transform and continuous wavelet transform. Journal of Mathematical Chemistry, 2009, 46, 562-568.	1.5	8
29	Accurate Prediction of Anti-hypertensive Peptides Based on Convolutional Neural Network and Gated Recurrent unit. Interdisciplinary Sciences, Computational Life Sciences, 2022, 14, 879-894.	3.6	8
30	Detrended cross-correlation coefficient: Application to predict apoptosis protein subcellular localization. Mathematical Biosciences, 2016, 282, 61-67.	1.9	7
31	iDHS-DMCAC: identifying DNase I hypersensitive sites with balanced dinucleotide-based detrending moving-average cross-correlation coefficient. SAR and QSAR in Environmental Research, 2019, 30, 429-445.	2.2	7
32	iORI-ENST: identifying origin of replication sites based on elastic net and stacking learning. SAR and QSAR in Environmental Research, 2021, 32, 317-331.	2.2	7
33	iPro-GAN: A novel model based on generative adversarial learning for identifying promoters and their strength. Computer Methods and Programs in Biomedicine, 2022, 215, 106625.	4.7	7
34	iDHS-DASTS: identifying DNase I hypersensitive sites based on LASSO and stacking learning. Molecular Omics, 2021, 17, 130-141.	2.8	6
35	iR5hmcSC: Identifying RNA 5-hydroxymethylcytosine with multiple features based on stacking learning. Computational Biology and Chemistry, 2021, 95, 107583.	2.3	6
36	M6A-GSMS: Computational identification of N ⁶ -methyladenosine sites with GBDT and stacking learning in multiple species. Journal of Biomolecular Structure and Dynamics, 2022, 40, 12380-12391.	3.5	5

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#	Article	IF	CITATIONS
37	Accurate prediction of Gram-negative bacterial secreted protein types by fusing multiple statistical features from PSI-BLAST profile. SAR and QSAR in Environmental Research, 2018, 29, 469-481.	2.2	4
38	Application of Machine Learning Techniques in Drug-target Interactions Prediction. Current Pharmaceutical Design, 2021, 27, 2076-2087.	1.9	4
39	Integrating Second-order Moving Average and Over-sampling Algorithm to Predict Apoptosis Protein Subcellular Localization. Current Bioinformatics, 2020, 15, 517-527.	1.5	3
40	Identification of DNA N4-methylcytosine sites based on multi-source features and gradient boosting decision tree. Analytical Biochemistry, 2022, 652, 114746.	2.4	3
41	A Gram-Negative Bacterial Secreted Protein Types Prediction Method Based on PSI-BLAST Profile. BioMed Research International, 2016, 2016, 1-5.	1.9	1
42	Prediction of Protein Subcellular Localization by Using ?-Order Factor and Principal Component Analysis. Letters in Organic Chemistry, 2017, 14, .	0.5	1
43	Application of Machine Learning Techniques to Predict Protein Phosphorylation Sites. Letters in Organic Chemistry, 2019, 16, 247-257.	0.5	1
44	Integrating LASSO Feature Selection and Soft Voting Classifier to Identify Origins of Replication Sites. Current Genomics, 2022, 23, 83-93.	1.6	1
45	Pathogenic Genes Selection Model of Genetic Disease based on Network Motifs Slicing Feedback. Current Proteomics, 2019, 16, 392-401.	0.3	0