



# 基于人工智能与多源信息融合 的蛋白质功能预测研究

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南京农业大学 人工智能学院

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2023年10月31日

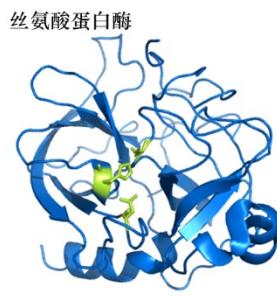
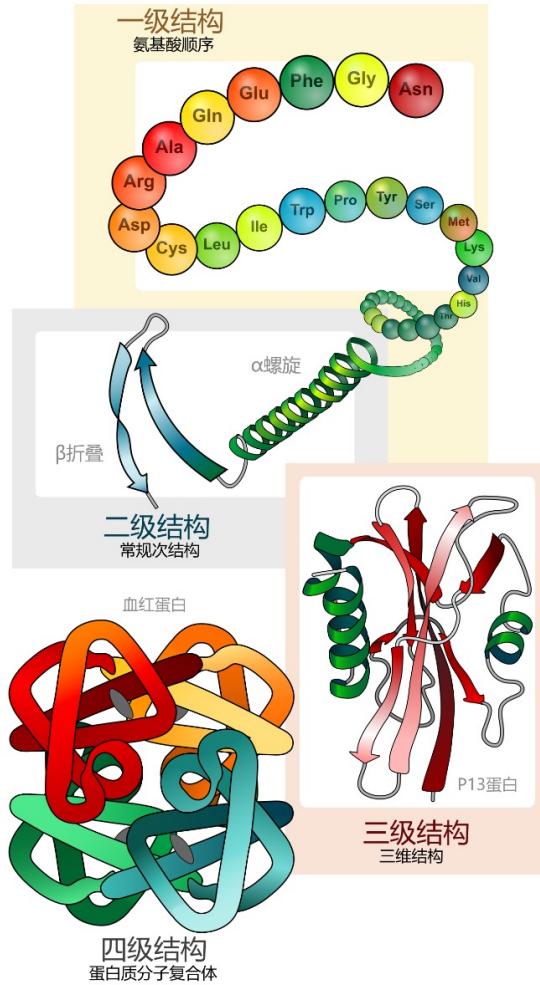
# 目 录

-  **研究背景**
-  **研究内容**
-  **未来展望**

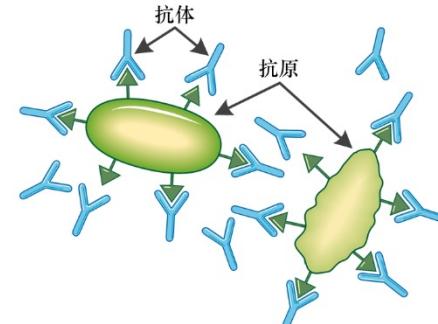
01      Part one

# 研究背景

# 01 蛋白质的生物功能



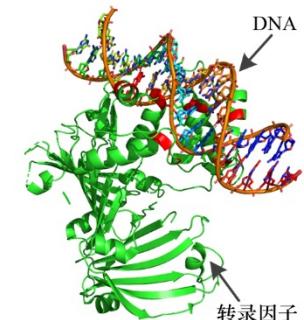
(a) 催化反应



(b) 免疫保护



(c) 运输载体

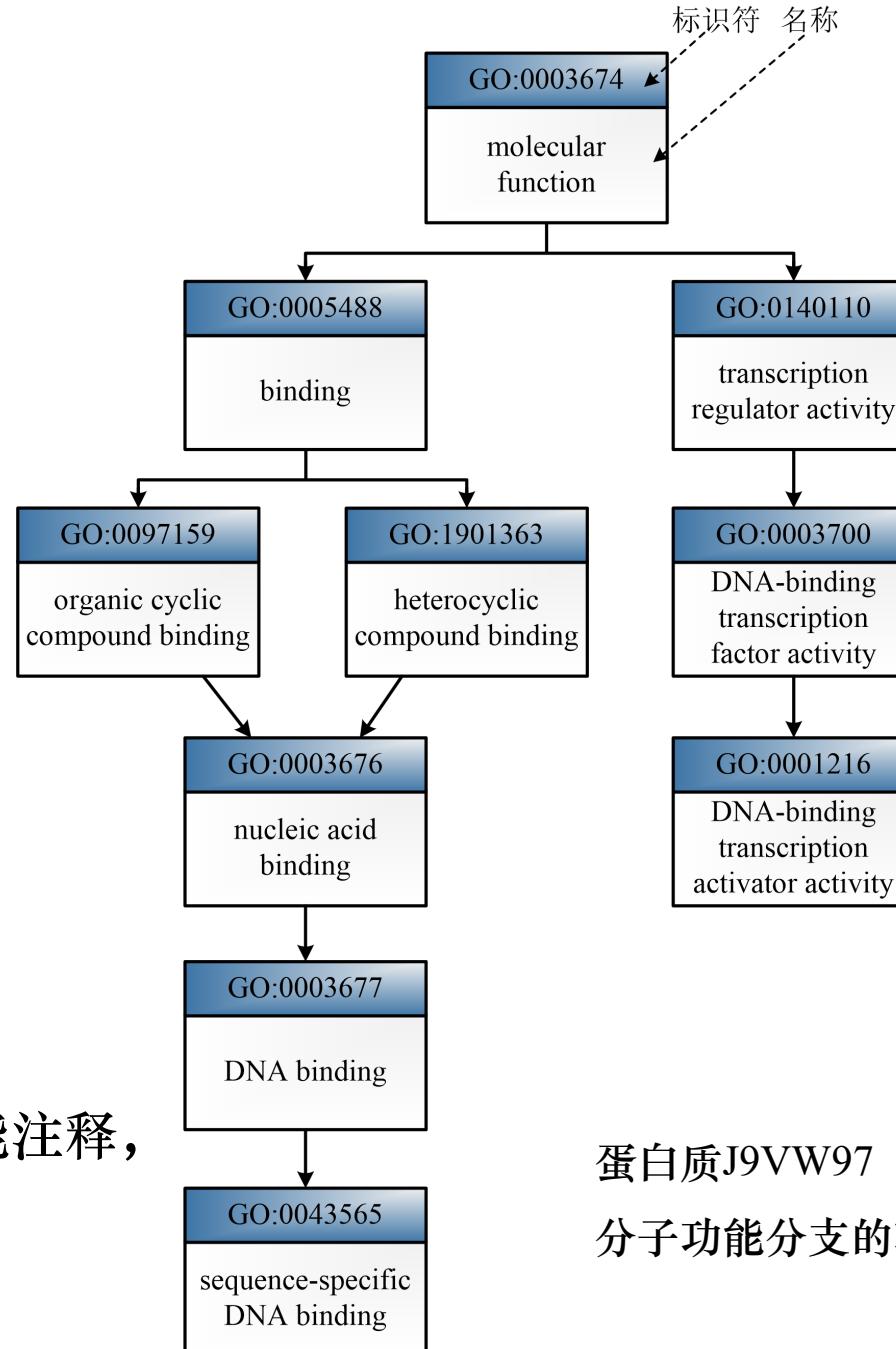


(d) 基因调控

- 识别和分析蛋白质的功能有助于解释各种生命活动现象，并阐明相关疾病的发病机理，进而指导相应的药物设计，以期推动智能医疗的发展。
- 蛋白质功能注释是后基因时代的首要任务之一。

## 02 蛋白质的功能注释方法

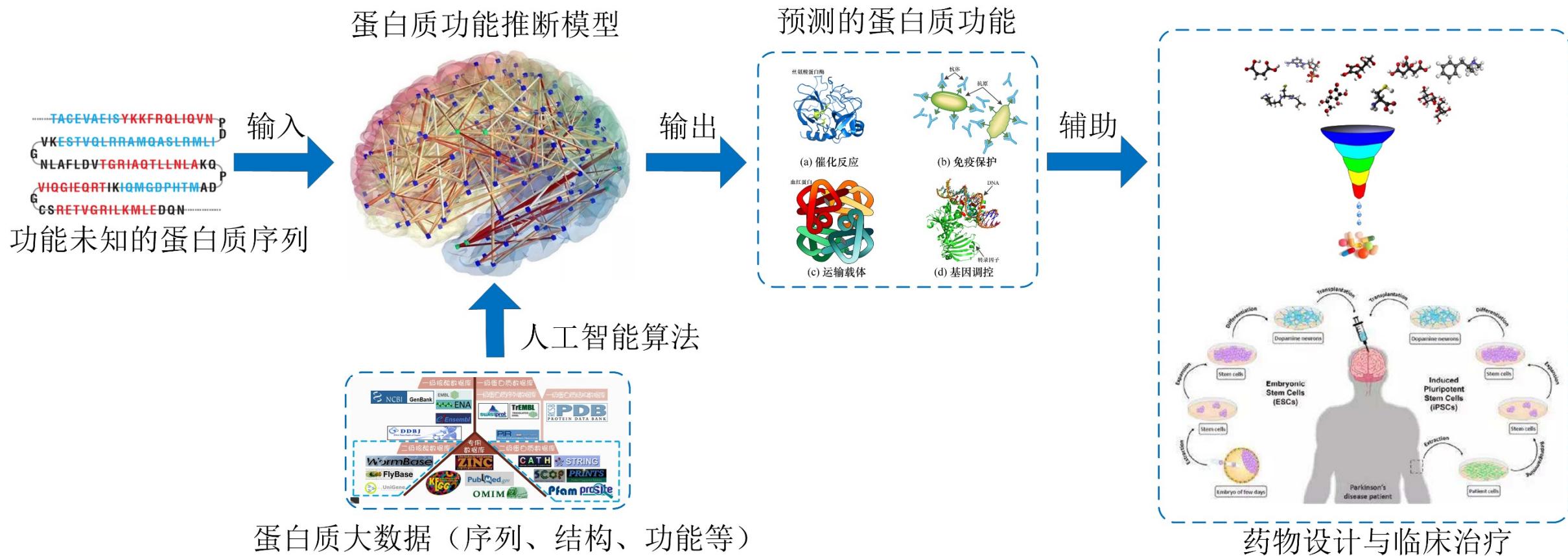
- 基因本体论 (Gene Ontology, GO)
  - 分子功能 (Molecular Function, MF)
  - 生物过程 (Biological Process, BP)
  - 细胞组件 (Cellular Component, CC)



蛋白质J9VW97 (UniProt ID) 在分子功能分支的功能注释图

## 03 人工智能算法预测蛋白质功能

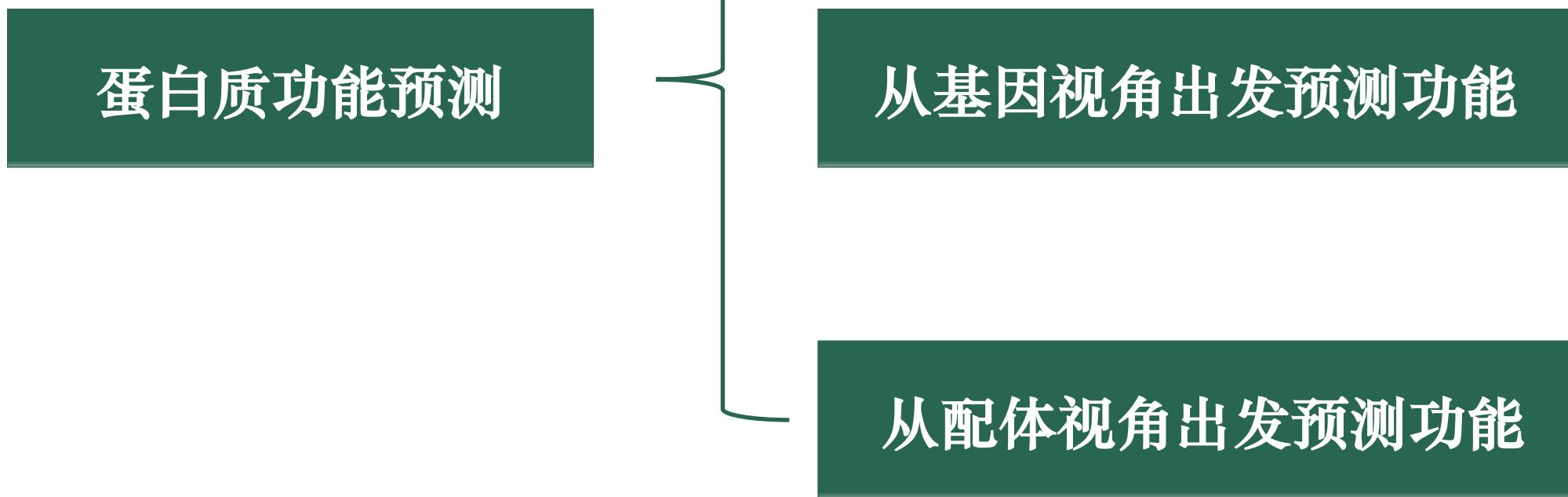
- 蛋白质功能注释最可靠的途径是生物实验，但它存在周期长、成本高等缺陷。
- 研发高效的人工智能算法来预测蛋白质功能已迫在眉睫（多标签预测任务）。



02 Part two

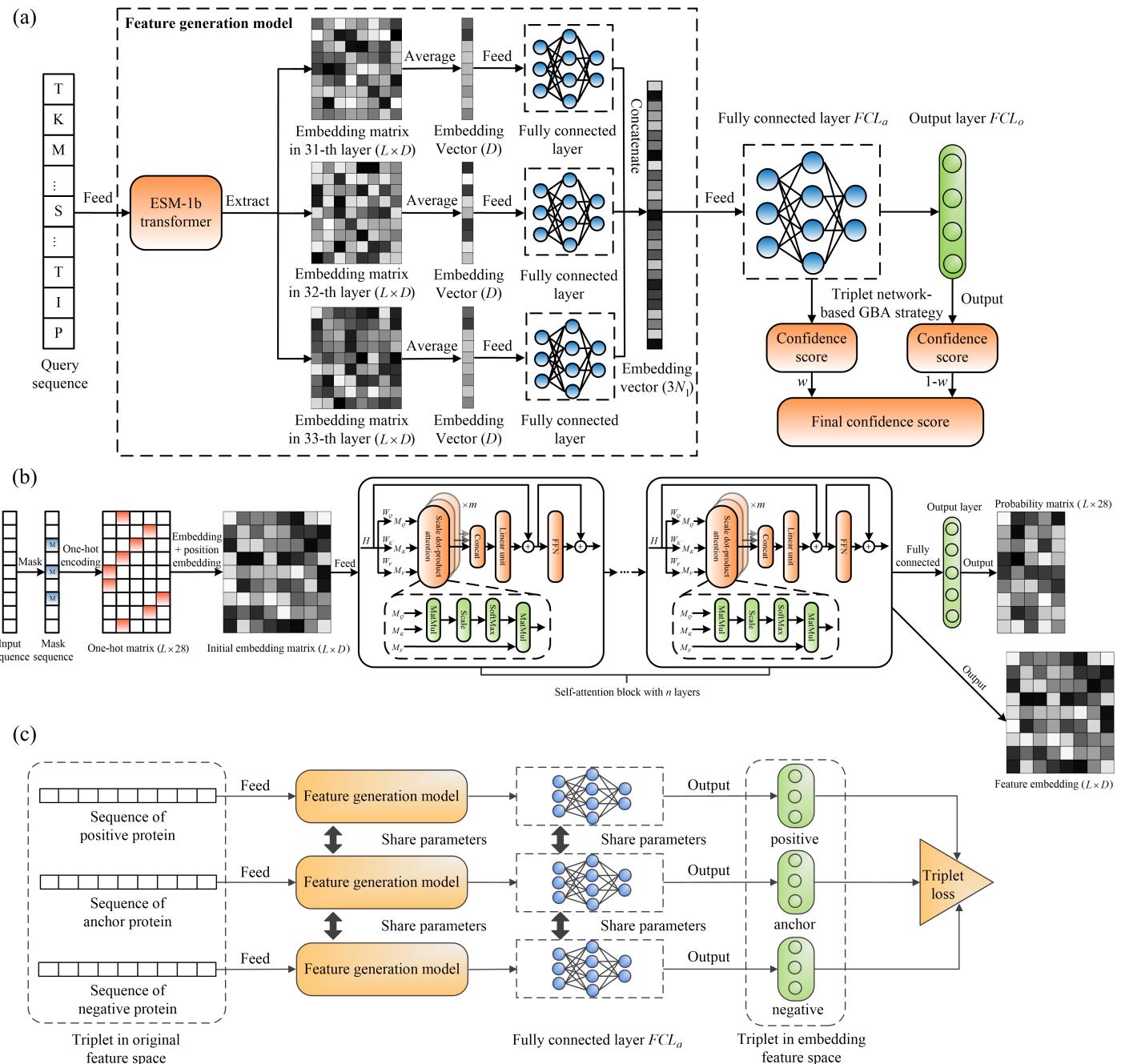
## 研究內容

## 04 研究内容



# 从蛋白质视角出发预测功能

- 基于注意力机制与三元组神经网络的预测方法 ATGO
- 主要贡献：首次将计算机视觉领域的无监督语言模型迁移到蛋白质功能预测领域



Yi-Heng Zhu, Chengxin Zhang, Dong-Jun Yu, Yang Zhang. Integrating Unsupervised Language Model with Triplet Neural Networks for Protein Gene Ontology Prediction. **PLOS Computational Biology**. 2022, 18(12): e1010793.



## Online Services

- I-TASSER
- QUARK
- LOMETS
- COACH
- COFACTOR
- MetaGO
- MUSTER
- CEthreader
- SEGMER
- FG-MD
- ModRefiner
- REMO
- DEMO
- SPRING
- COTH
- BSpred
- ANGLOR
- EDock
- BSP-SLIM
- SAXSTER
- FUpred
- ThreaDom
- ThreaDomEx
- EvoDesign
- GPCR-I-TASSER
- MAGELLAN
- BindProf
- BindProfX
- SSIPe
- ResQ
- IonCom
- STRUM
- DAMpred
- TM-score



ATGO is a deep learning-based algorithm for high accuracy protein Gene Ontology (GO) prediction. Starting from a query sequence, it first extracts three layers of feature embeddings from a pre-trained protein language model (ESM-1b). Next, a fully connected neural network is used to fuse the feature embeddings, which are then fed into a supervised triplet network for GO function prediction. Large-scale benchmark tests demonstrated significant advantage of ATGO on protein function annotations due to the integration of discriminative feature embeddings from attention transformer models. ([view an example of ATGO prediction](#))

## ATGO On-line Server

Input Sequence (Optional, [30,10000] residues in [FASTA format](#))

Copy and paste your protein sequence file here ([Sample input](#))

```
>Q9HG13
MAYFRLAVILLAVASSVVAVKVNPLPAPRHISWGHSKPPLSDVSLRTERDTDDSIITNAWNRAWEVIVSLEWVPAGIEA
PIPEFDEFPTSTPSASAAATRSKRANVPQFVDVVEDWDADLQHGVDESYTLDAKAGSDAIDTAKTVWGAFTTLQ
QLVISDGNGGLILEQPVHKDAPLYPYRGLMVDTGRNFISVRKLHEQLDMALSKLNVLHWHLDDTQSWPVHIDAYPEM
TIDKAYSERTYSHDDLRNVYARARGIRVPEIDMPAHASAGWQQVDPDIVACANSWWNSNDNWPLHTAVQPNPGQL
DIINPKTYEVQDVYEELSSIFTDDWFHVGGDEIOPNCYNFSTYTEWFQEDPSRTYNDLMQHWDVKAPVIFRSVSDSR
RLVMWEDVNLTEHADYPTDVMQSWNNNGLENKLTERGYDVIVSSADFMYLDCRGGGYVTNDDRYNEQTNPDDP
TPSFNYGGGGSWCGPYKTWQRIYNYDFTLNLTNAQAKHVIGATAPLWSEQVDDVNISNLFWPRAALAELEVWSGNRD
AKGNKRTTFLTQRILNFREYLLANGMAATVVPKYCLQHPHACDLNYDQTVLH
```

Or upload the sequence file from your local computer

未选择文件

Email: (mandatory, where results will be sent to)

Job ID: (optional, your given name to your job)

## ATGO Download

- [Download the standalone package.](#)
- [Download prediction models.](#)
- [Download benchmark datasets.](#)

## References:

- Yi-Heng Zhu, Chengxin Zhang, Dong-Jun Yu, Yang Zhang. Integrating unsupervised language model with triplet neural networks for protein gene ontology prediction. PLOS Computational Biology, 2022, 18 (12): e1010793.

<https://zhanggroup.org/ATGO/>

## ATGO result for protein E7CIP7

[Download [result.zip](#) for all prediction results]

## User Input

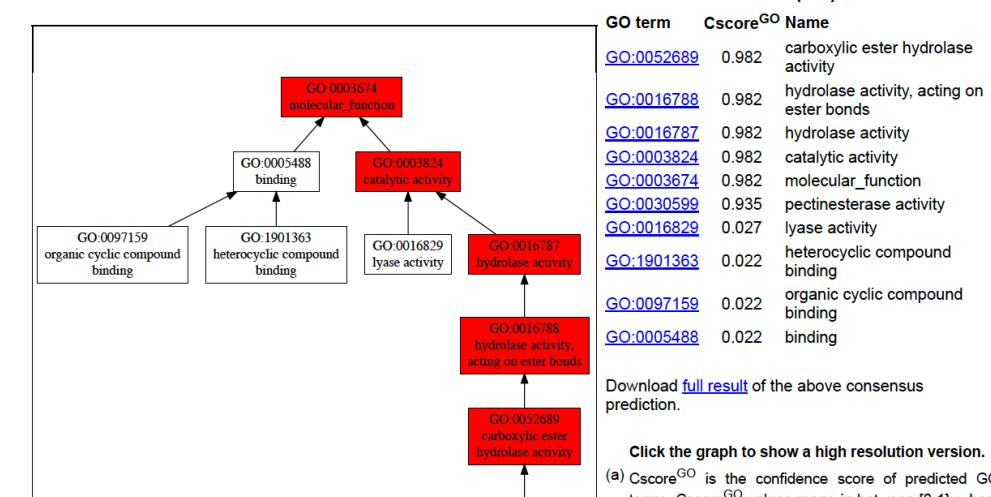
>E7CIP7 (382 residues)

```
MKIIVLILLAVVLASADQTAPGTASRPILTASESNYFTTATYLOQWSPPSISTSKADYTV
GNGYNTIQAQAVNAINTGGTRKYIKINACTYQEVVYIPNTKVPLTIYGGSSPSDTLIT
LNMPAQTPSAYKSLSVGSLFNADPAYSMYNCSASKSGTIGTSCSTVFWKAPAVQINV
SIENSANKNTGDDQAVALQTNSDQIQIHNARLLGHQDTLYAGSGSSSVERSYYNTYIEGD
IDFVFGGSAIFESCTFYVKADRRSDTAVVFAPDTDPMKMYFVYKSTITGDSAWSSK
KAYLGRRAWDGSVSSSAVPGTSPNGQLIKESTIDGIINTSGPWTATSGRTYSGNNAN
SRDLNNNDNYNRFWEYNNNGNGA
```

Download query [sequence](#)

## Predicted Gene Ontology (GO) Terms

## Molecular Function (MF)



Download [full result](#) of the above consensus prediction.

Click the graph to show a high resolution version.

(a)  $Cscore^{GO}$  is the confidence score of predicted GO terms.  $Cscore^{GO}$  values range in between [0-1]; where a higher value indicates a better confidence in predicting the function using the template.

(b) The graph shows the predicted terms within the Gene Ontology hierarchy for Molecular Function. Confidently predicted terms are color coded by  $Cscore^{GO}$ :

[0.40,0.5) [0.5,0.6) [0.6,0.7) [0.7,0.8) [0.8,0.9) [0.9,1.0]

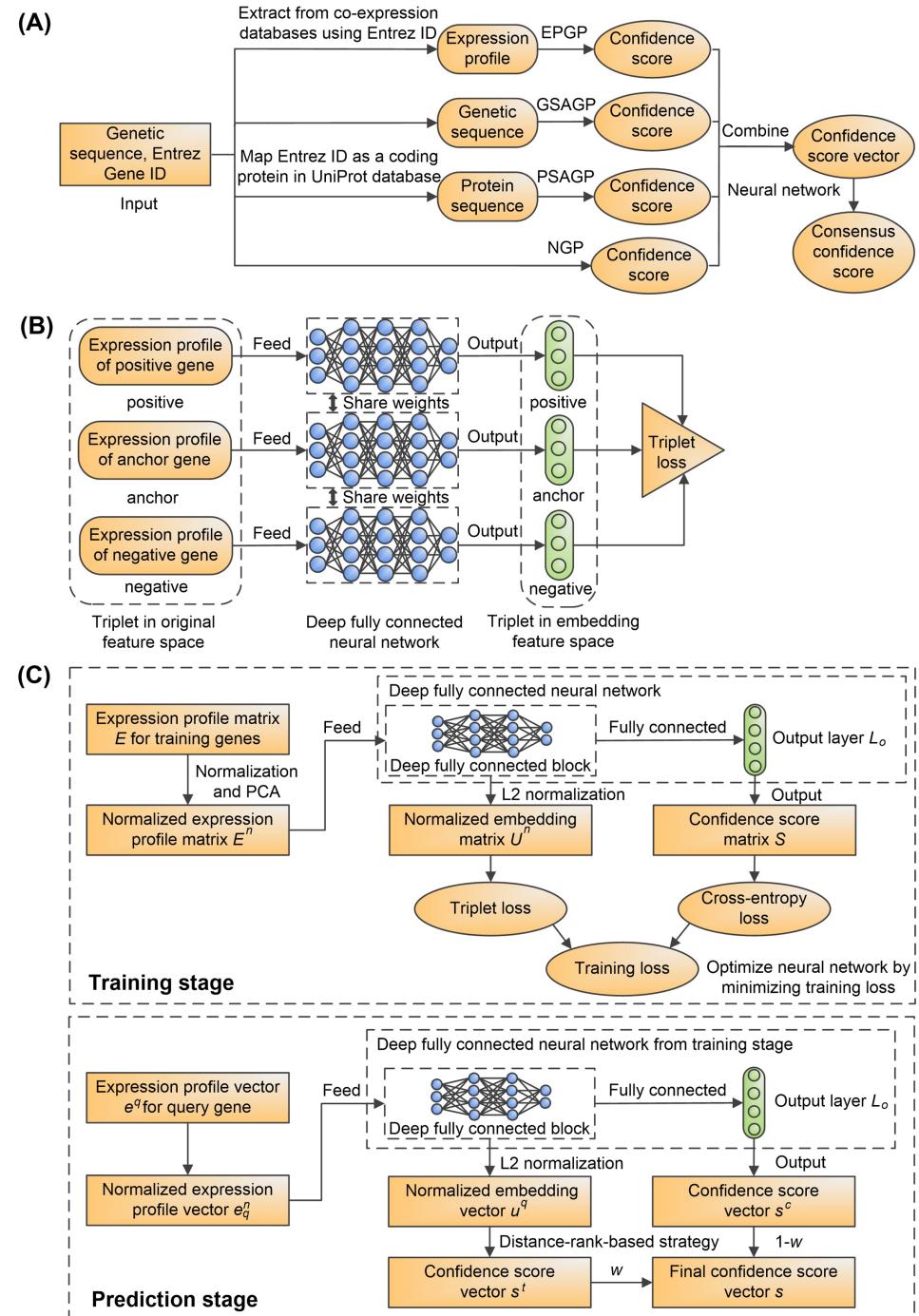
## Biological Process (BP)

GO term	Cscore <sup>GO</sup>	Name
<a href="#">GO:0008150</a>	0.751	biological_process
<a href="#">GO:0071704</a>	0.727	organic substance metabolic process
<a href="#">GO:0044238</a>	0.727	primary metabolic process
<a href="#">GO:0008152</a>	0.727	metabolic process

# 从基因视角出发预测功能

- 基于度量学习与多源信息融合的功能预测方法 TripletGO
- 主要贡献：首次将基于基因视角的方法和基于蛋白质视角的相结合，为后续的研究开辟了新的思路

Yi-Heng Zhu, Chengxin Zhang, Yan Liu, Gilbert Omenn, Peter Freddolino, Dong-Jun Yu, Yang Zhang. TripletGO: Integrating Transcript Expression Profiles with Protein Homology Inferences for Gene Function Prediction. *Genomics, Proteomics & Bioinformatics*. 2022, 20(5): 1013-1027.





## Online Services

- [I-TASSER](#)
- [I-TASSER-MTD](#)
- [C-I-TASSER](#)
- [CR-I-TASSER](#)
- [QUARK](#)
- [C-QUARK](#)
- [LOMETS](#)
- [MUSTER](#)
- [CEthreader](#)
- [SEGMER](#)
- [DeepFold](#)
- [DeepFoldRNA](#)
- [FoldDesign](#)
- [COFACTOR](#)
- [COACH](#)
- [MetaGO](#)
- [TripletGO](#)
- [IonCom](#)
- [FG-MD](#)
- [ModRefiner](#)
- [REMO](#)
- [DEMO](#)
- [DEMO-EM](#)
- [SPRING](#)
- [COTH](#)
- [Threpp](#)
- [PEPPI](#)
- [BSpred](#)
- [ANGLOR](#)
- [EDock](#)
- [BSP-SLIM](#)
- [SAXSTER](#)
- [FUpred](#)
- [ThreaDom](#)
- [ThreaDomEx](#)
- [EvoDesign](#)
- [BindProf](#)
- [BindProfX](#)
- [SSIPe](#)
- [GPCR-I-TASSER](#)



TripletGO is an algorithm for predicting Gene Ontology (GO) of genes. It consists of four pipelines to detect GO terms through (1) expression profile similarity based on triplet network, (2) genetic sequence alignment, (3) protein sequence alignment, and (4) naive probability. The final function insights are a combination of the four pipelines through neural network. ([view an example of TripletGO prediction](#))

## Triplet On-line Server

Sequence of Query Gene (Optional, [30,10000] residues in FASTA format)

Copy and paste your genetic sequence file here ([Sample input](#))

We would suggest you provide Entrez ID for query gene, which helps to find its expression profile and coding proteins.

Entrez ID provides unique integer identifiers for genes in National Center for Biotechnology Information.

```
>839799
GGGCTTATTGGGCTGGACCTAGCCCATTGTGAGGTGTTAAACGATGCGTTGGCATTCAAGTAGG
GCTTCAATTTCATTCGTTTGCTGAGATAAAAGTAGAGAGAAATCTAAATTCGAGAGGAGA
AGTTTAATTTCGAGTTAGATTCAATGGAAGAGATCACGGAAGGGAGTTAACACATG
AACCTGGCTGGTGTGATACCCAGAAAGAAGATGGATTCAAGTTCCAACACTAAAGAACCA
TTGTTCTCTACGTCAATCTGCCAAGAGGTACATGCAGCAGTACACTGATGTCGAATTG
TCTGCACTAGGAATGGCTATTGCCACTGTTGTTACGGTGCCTGAGATATTGAAGAACAAAT
GGCTTGCTGTTGAAAAGAAGATCATGACATCAGCTGTTGAGATATCAAGGATGATTCAAGG
GGTCTCTGTGAGAAAGCTAAAGATGAGATCAGCTTGCCTAAGCTGAGAGTTGAT
GAACATGGCTGAGCTAATGAAGAGAAGAGCTGCAAGAGCCAAAGAGCAAACACTAG
ATTGTTCAAGTTTCTGTTCAACGATCTATTCTGTTCACTTTTATTTCACCTTGGAT
TAATTTAAAGACACTTCTGTTAATTTCGTTACTTTTATTTCACCTTGGAT
TGTGCTCTGTGACCTCTGAGCATTTTAAAGATGCTAGGAAGTATAAAAAGATG
GCTTCGTTGCATAAA
```

Or upload the sequence file from your local computer

未选择文件

Email: (mandatory, where results will be sent to)

E-value e1 (optional, default 0.1)

The e-value for Blastn software in genetic sequence alignment

E-value e2 (optional, default 0.1)

The e-value for Blasp software in protein sequence alignment

Cut-off value t1 (optional, 0.0-1.0, default 1.0)

The templates which have more than t1 sequence identity with the query are removed in genetic sequence alignment

Cut-off value t2 (optional, 0.0-1.0, default 1.0)

The templates which have more than t2 sequence identity with the query are removed in protein sequence alignment

Job ID: (optional, your given name to your job)

<https://zhanggroup.org/TripletGO/>

## TripletGO result for Gene 839799

[Download [result.zip](#) for all prediction results]

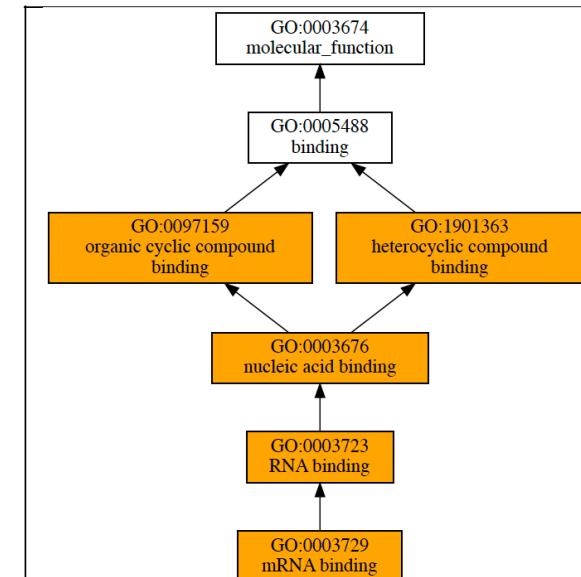
## User Input

>839799 (795 residues)

```
GGGCTTATTGGGCTGGACCTAGCCCATTGTGAGGTGTTAAACGATGCGTTGGCATTCAAGTAGG
GCTTCAATTTCATTCGTTTGCTGAGATAAAAGTAGAGAGAAATCTAAATTCGAGAGGAGA
AGTTTAATTTCGAGTTAGATTCAATGGAAGAGATCACGGAAGGGAGTTAACACATG
AACCTGGCTGGTGTGATACCCAGAAAGAAGATGGATTCAAGTTCCAACACTAAAGAACCA
TTGTTCTCTACGTCAATCTGCCAAGAGGTACATGCAGCAGTACACTGATGTCGAATTG
TCTGCACTAGGAATGGCTATTGCCACTGTTGTTACGGTGCCTGAGATATTGAAGAACAAAT
GGCTTGCTGTTGAAAAGAAGATCATGACATCAGCTTGCCTAAGCTGAGAGTTGAT
GAACATGGCTGAGCTAATGAAGAGAAGAGCTGCAAGAGCCAAAGAGCAAACACTAG
ATTGTTCAAGTTTCTGTTCAACGATCTATTCTGTTCACTTTTATTTCACCTTGGAT
TAATTTAAAGACACTTCTGTTAATTTCGTTACTTTTATTTCACCTTGGAT
TGTGCTCTGTGACCTCTGAGCATTTTAAAGATGCTAGGAAGTATAAAAAGATG
GCTTCGTTGCATAAA
```

[Download query sequence](#)

## Predicted Gene Ontology (GO) Terms



## Molecular Function (MF)

GO term	Cscore <sup>GO</sup>	GO Name
<a href="#">GO:1901363</a>	0.866	heterocyclic compound binding
<a href="#">GO:0097159</a>	0.866	organic cyclic compound binding
<a href="#">GO:0003676</a>	0.884	nucleic acid binding
<a href="#">GO:0003723</a>	0.877	RNA binding
<a href="#">GO:0003729</a>	0.874	mRNA binding

[Download full result](#) of the above consensus prediction.

Click the graph to show a high resolution version.

(a) Cscore<sup>GO</sup> is the confidence score of predicted GO terms. Cscore<sup>GO</sup> values range in between [0-1]; where a higher value indicates a better confidence in predicting the function using the template.

(b) The graph shows the predicted terms within the Gene Ontology hierarchy for Molecular Function. Confidently predicted terms are color coded by Cscore<sup>GO</sup>:

[0.13,0.5] [0.5,0.6] [0.6,0.7] [0.7,0.8] [0.8,0.9] [0.9,1.0]

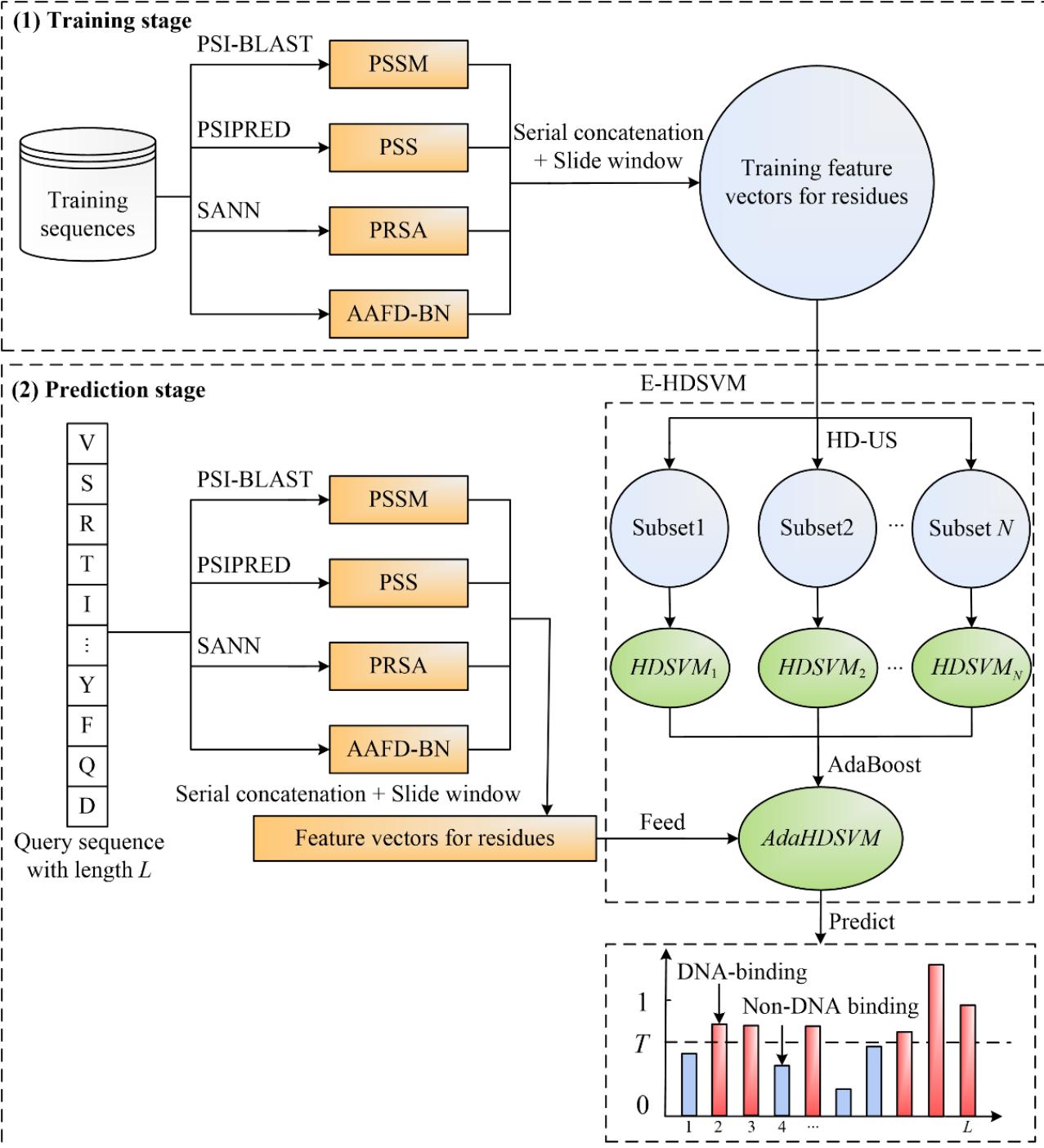
## Biological Process (BP)

GO term	Cscore <sup>GO</sup>	GO Name
<a href="#">GO:0009987</a>	0.443	cellular process
<a href="#">GO:0008152</a>	0.231	metabolic process
<a href="#">GO:0071704</a>	0.221	organic substance metabolic

# 从配体视角出发预测功能

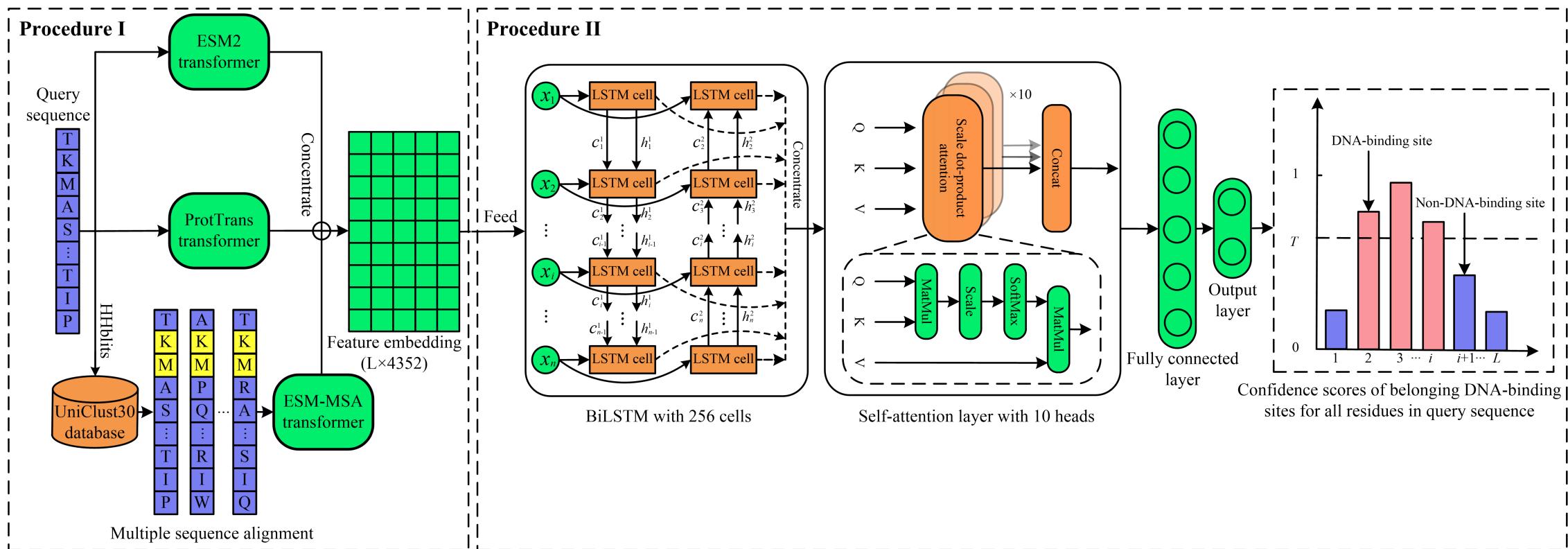
- 基于多粒度支持向量机集成与序列特征的蛋白质-DNA结合定位点预测方法 DNAPred
- 主要贡献：提出了新的类不平衡学习算法 E-HDSVM，显著地提升了蛋白质-DNA结合定位点预测精度。

Yi-Heng Zhu, Jun Hu, Xiao-Ning Song, Dong-Jun Yu. DNAPred: Accurate Identification of DNA-binding Sites from Protein Sequence by Ensembled Hyperplane-Distance-Based Support Vector Machines. **Journal of Chemical Information and Modeling.** 2019, 59:3057-3071.



# 从配体视角出发预测功能

- 基于无监督语言模型与多源信息融合的蛋白质-DNA绑定点预测方法 ULDNA
- 主要贡献：融合多种无监督蛋白质语言模型，显著地提升了蛋白质-DNA绑定点预测精度。



# DNAPred: Identifying DNA-Binding Sites from Protein Sequence by Ensemble Hyperplane-Distance-Based Support Vector Machine

| [Read Me](#) | [Dataset](#) | [Citation](#) | [Large-Scale Test](#) |

Input query protein sequence(s) in FASTA format:

```
>2XTNA
MDQNEHSHWGPHAKGQCASRSELRIILVGKTGTGKSAAGNSILRKQAFESKLGS
QTLTKTCSKSQGSWGNREIVIIDTPDMFSWKDHCEALYKEVQRCYLLSAPGPHV
LLLVTQLGRYTSQDQQAAQRVKEIFGEDAMGHTIVLFTHKEDLNGSLMDYMH
DSDNKALSKLVAACGGRICAFNNRAEGSNQDDQVKELMDCIEDLLMEKNGDHY
TNGLYSLIQRSKCGPVGSD
```

[Example](#)

[Reset Sequence\(s\)](#)

## Choose a prediction model

Model constructed on PDNA-543

Model constructed on PDNA-335

## Choose a threshold

Threshold 1 (*Max MCC*)

Threshold 2 (*FPR*≈5%)

Threshold 3 (*Sen*≈*Spe*)

**Email Address (For receiving your prediction results)\***

[Submit](#)

[Clear All](#)

## Reference:

Yi-Heng Zhu, Jun Hu, Xiao-Ning Song and Dong-Jun Yu \*. DNAPred: Identifying DNA-Binding Sites from Protein Sequence by Ensemble Hyperplane-Distance-Based Support Vector Machine. Journal of Chemical Information and Modeling, 2019.

## RESULTS PAGE

Predicting Protein-DNA Binding Sites

### Protein Name

2XTNA

### Model constructed on Dataset

PDNA-543

### Threshold

0.265 (*Max MCC*)

### Prediction Summary

Number of predicted DNA-binding residues in protein **2XTNA**: 4

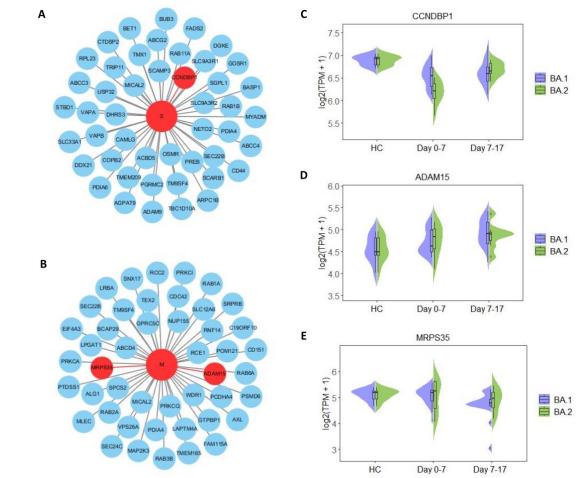
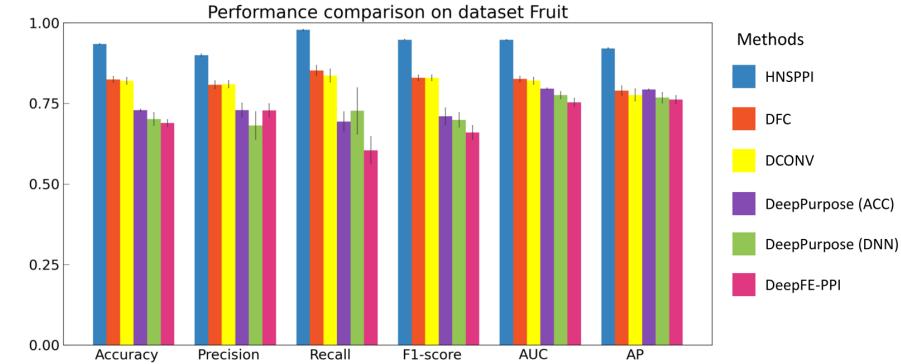
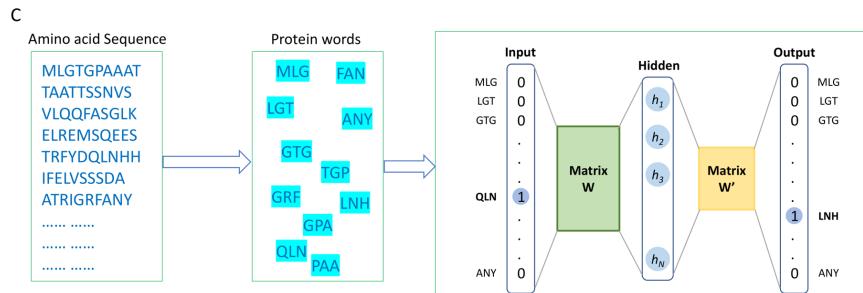
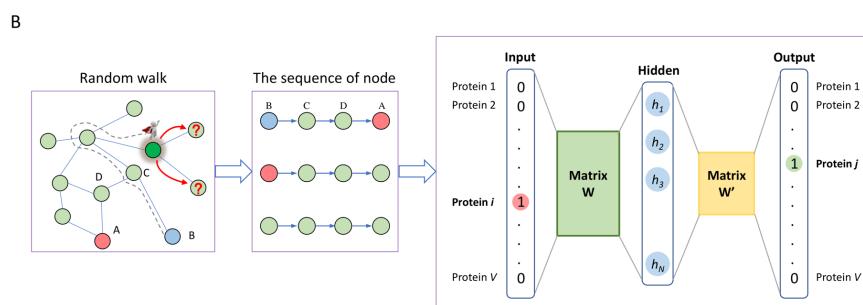
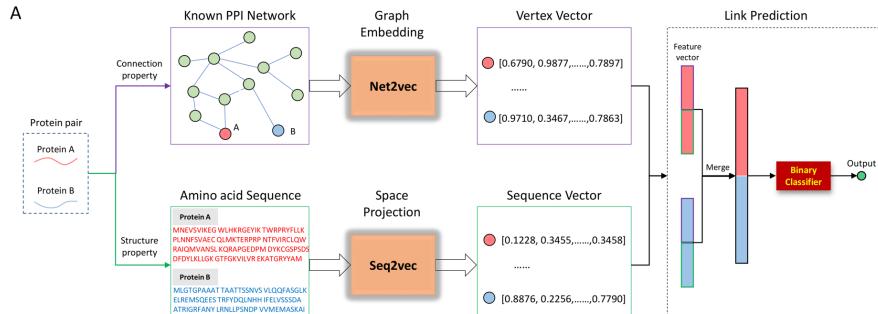
Specific position: **58 T 117 R 119 T 147 H**

### Predicted Results

Residue #	Amino Acid Type	Probability	Binding Residue
0001	M	0.058	N
0002	D	0.028	N
0003	Q	0.024	N
0004	N	0.049	N
0005	E	0.013	N
0006	H	0.063	N
0007	S	0.008	N
0008	H	0.037	N
0009	W	0.095	N
0010	G	0.009	N
0011	P	0.019	N
0012	H	0.081	N
0013	A	0.017	N
0014	K	0.080	N
0015	G	0.006	N
0016	Q	0.013	N

# 蛋白质-蛋白质相互作用预测

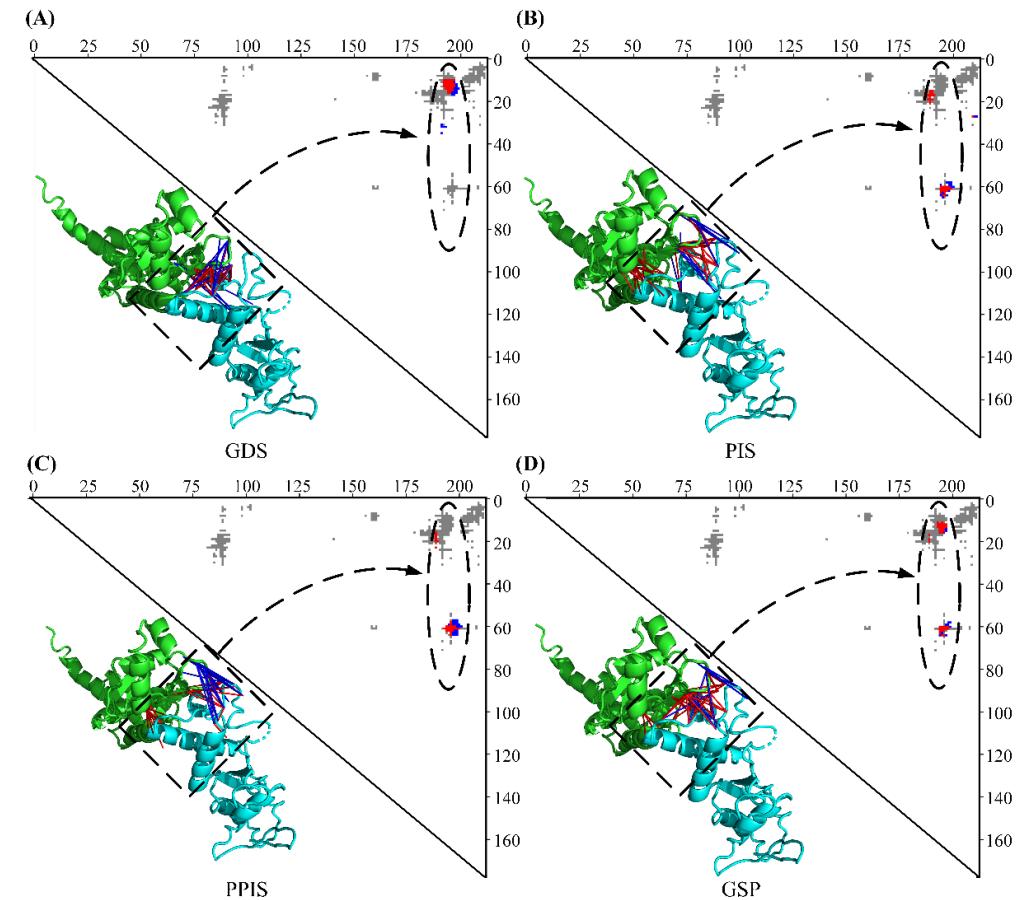
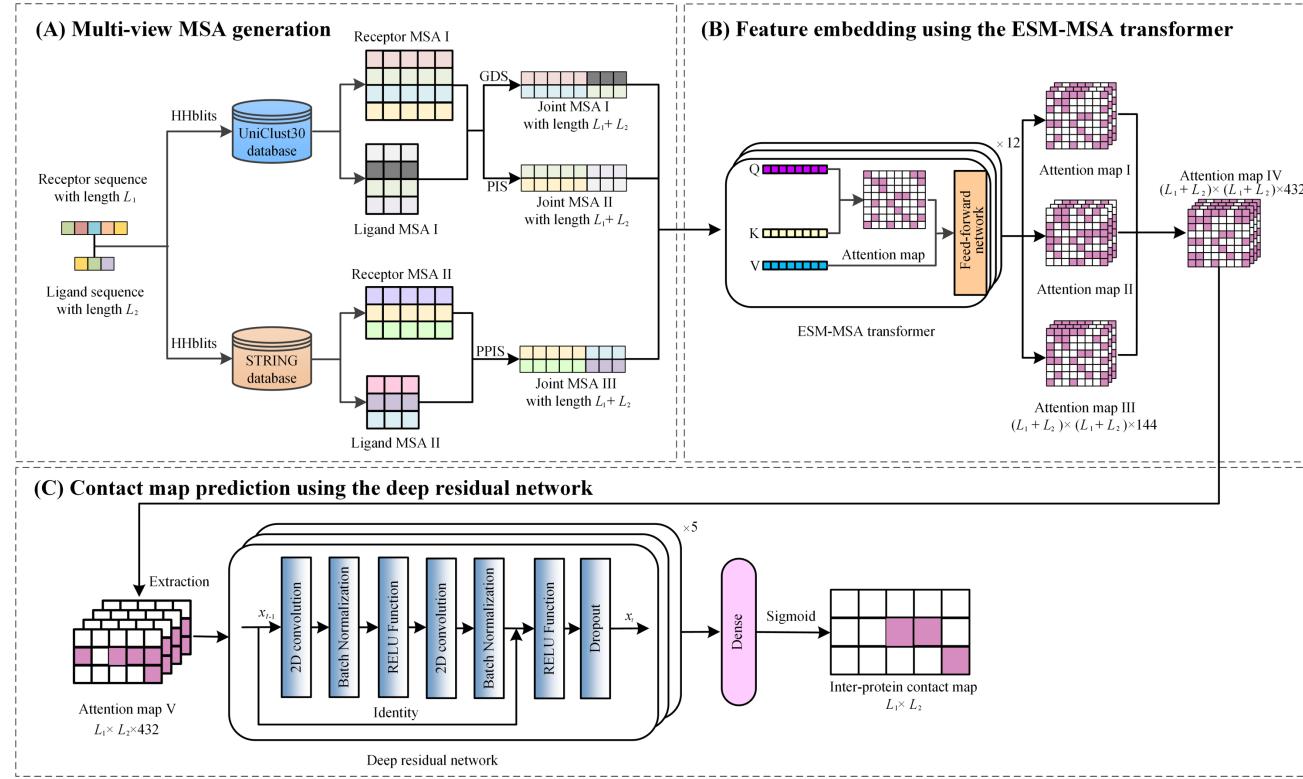
## ➤ 基于无监督语言模型与多网络融合的蛋白质相互作用预测方法HNSPPI



Shijie Xie, Xiaojun Xie, Xin Zhao, Fei Liu, Yiming Wang, Jihui Ping, Zhiwei Ji. HNSPPI: A hybrid computational model combining network and sequence information for predicting protein–protein interaction. *Briefings in Bioinformatics*, 2023, 24(5): bbad261.

# 蛋白质-蛋白质相互作用预测

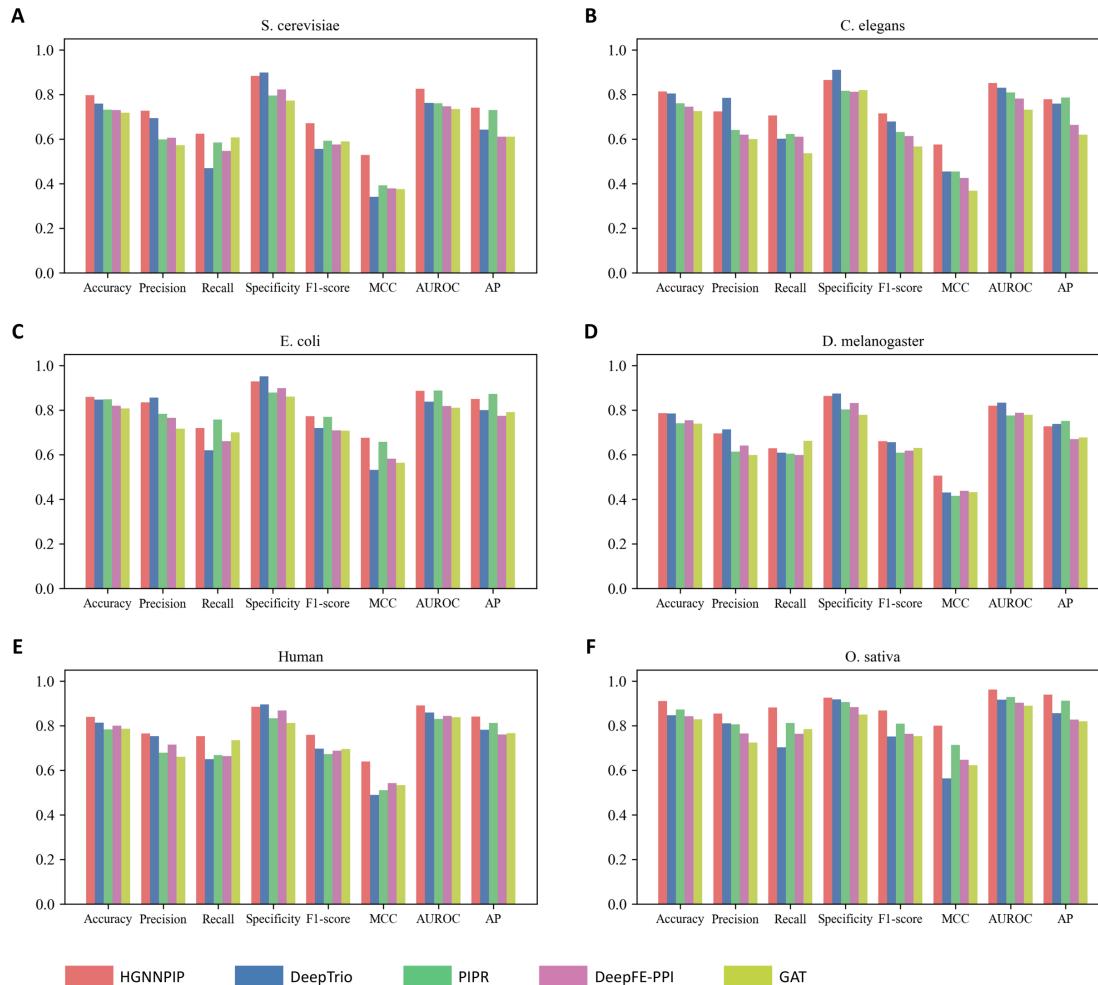
➤ 基于无监督语言模型与多视角多序列联配的蛋白质-蛋白质相互作用预测方法ICCPred



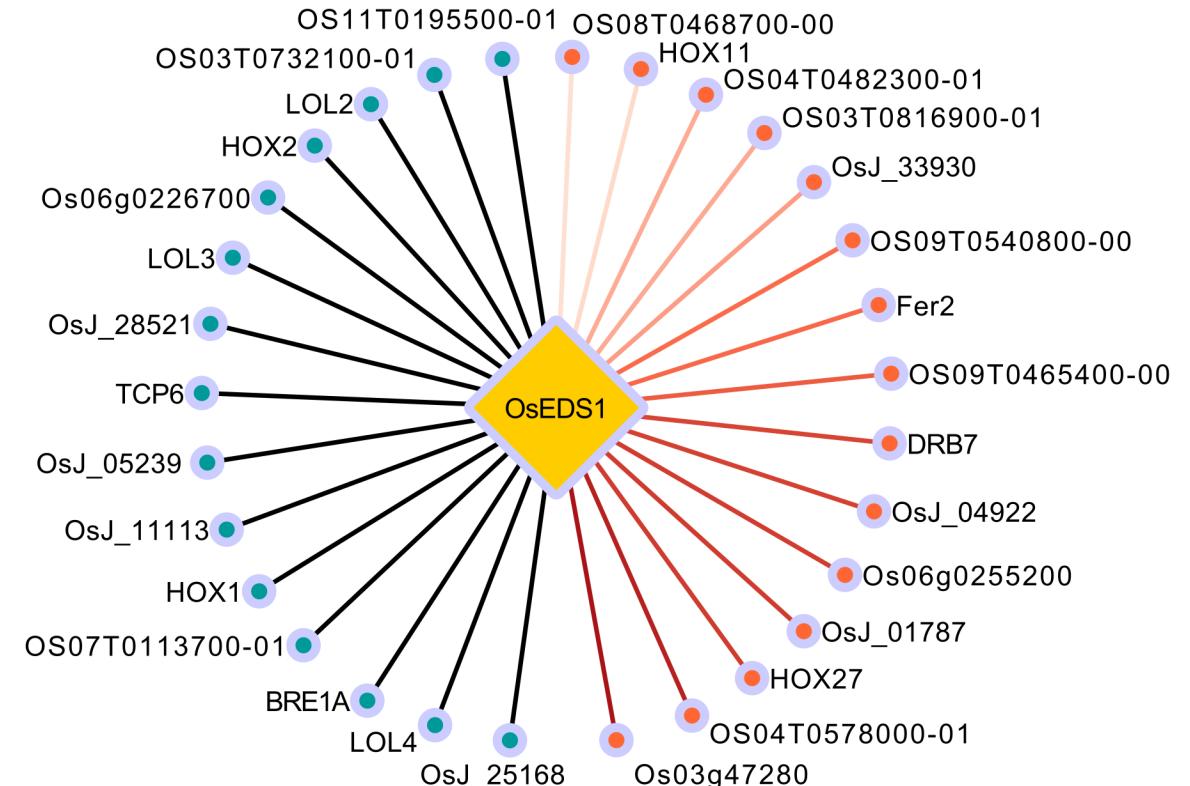
# 蛋白质-蛋白质相互作用预测

## ➤ Prediction of *U.virens*-Rice Interactions with Graph Convolutional networks

HGNNPIP模型与4种SOTA方法的性能比较



HGNNPIP模型预测OsEDS1的作用蛋白



03

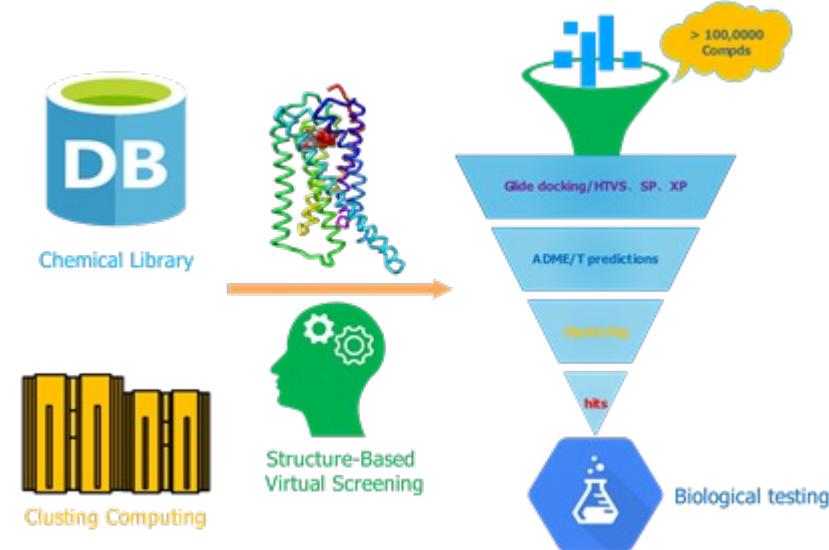
Part three

# 未来展望

# 01 蛋白质功能预测研究的应用前景

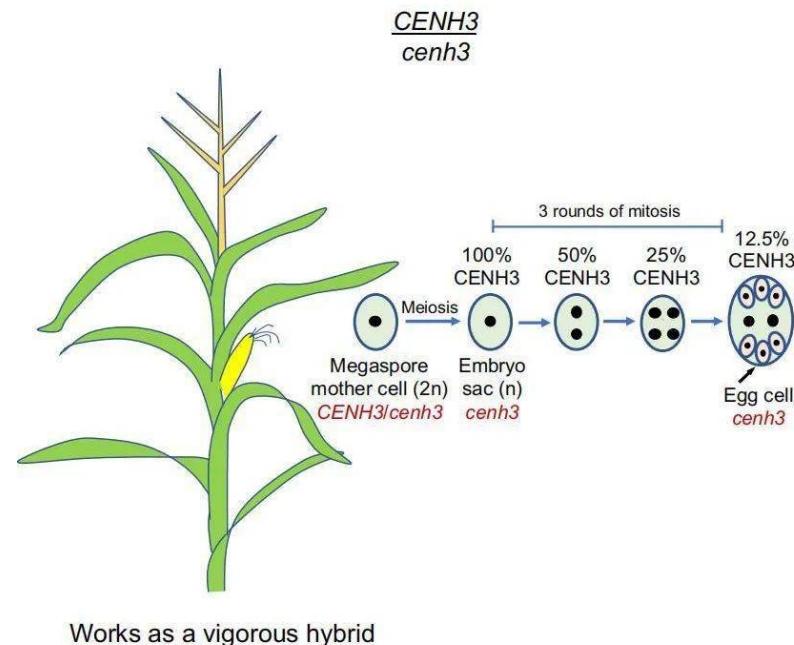
蛋白质功能预测研究有助于推动智能医疗的发展

- (1) 辅助疾病分析和诊断 (推断关键致病蛋白质)
- (2) 辅助药物设计 (药物分子筛选)



蛋白质功能预测在农业领域的应用前景

- (1) 植物遗传育种
- (2) 植物与微生物的相互作用
- (3) 植物蛋白组学



## 02 科研团队信息

### 计智伟教授团队

- ◆ 教授1人
- ◆ 副教授1人
- ◆ 助理教授2人
- ◆ 博士研究生3人
- ◆ 硕士研究生9人



### 研究方向

#### 人工智能与模式识别

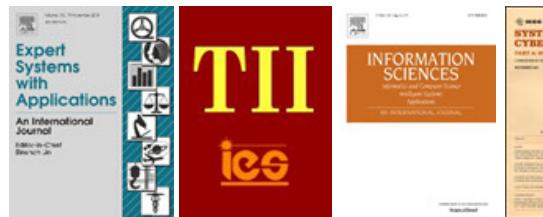
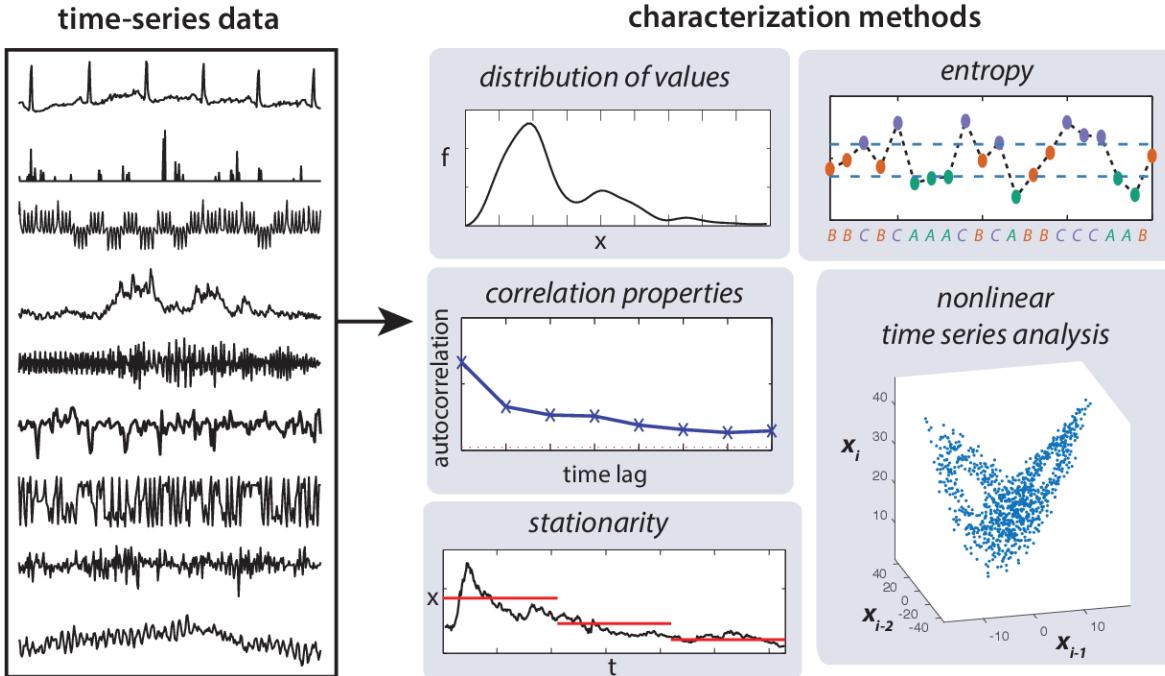
人工智能的理论及应用  
大数据计算与模式识别

#### 生物信息与系统生物学

多组学数据整合分析与计算  
复杂生物系统的数学建模与预测

# 主要成果1：人工智能与模式识别领域

## ◆ 时间序列数据挖掘与异常模式发现



Z Ji\*, Y Wang, X Xie, et al., *Expert Systems with Applications*, 2022.

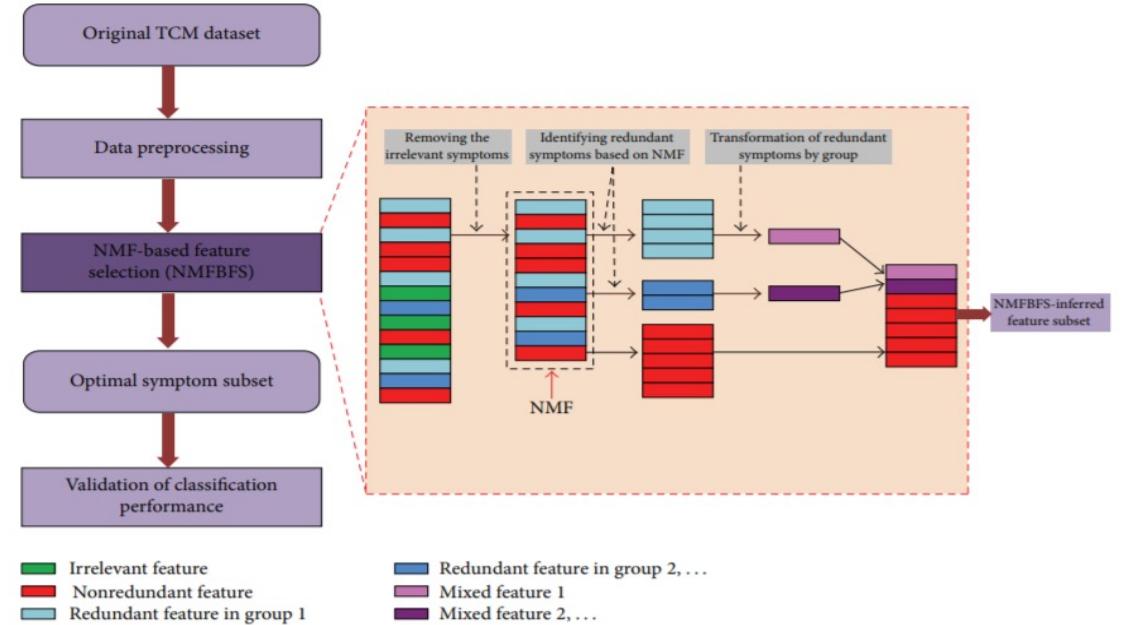
N Jin, Y Zeng, K Yan, Z Ji, *IEEE Transactions on Industrial Informatics*, 2021.

M Hu, X Feng, Z Ji\*, et al., *Information Sciences*, 2019.

K Yan#, Z Ji#, et al., *IEEE Transactions on Systems, Man, and Cybernetics: Systems*, 2019.

K Yan, Z Ji\*, et al., *Neurocomputing*, 2017.

## ◆ 高维复杂数据的维度约简和模型优化



X Xie, F Xia, K Yan, H Xu, Z Ji\*, *Plant Phenomics*, 2023.

F Xia, X Xie, S Jin, K Yan, Z Ji\*, *Frontiers in Plant Science*, 2021.

K Yan, ..., Z Ji\*, et al., *IEEE/ACM Trans on Computational Biology and Bioinformatics*, 2021.

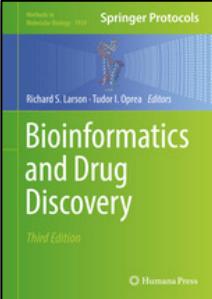
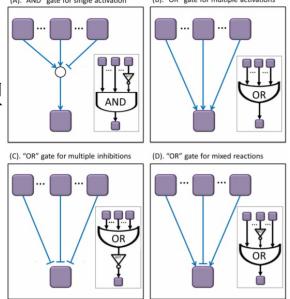
X Xie, X Gu, Y Li, Z Ji\*, *Knowledge-based Systems*, 2021.

Z Ji, ..., B Wang\*, *Computational and mathematical methods in medicine*, 2015.

# 主要成果2：生物信息与系统生物学领域

- ◆ 创立了一套独特的**生物分子网络建模方法**
- ◆ 创立了**分子-细胞-组织的3D多尺度建模方法**
- ◆ 建立了**生物组学大数据挖掘**的计算框架

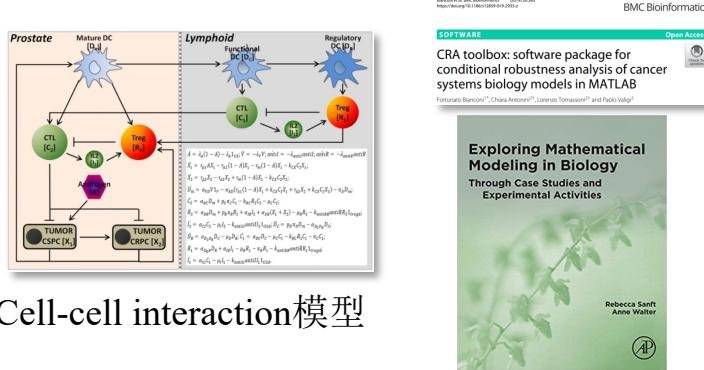
基于线性规划的离散时间建模方法BLP,DILP, TILP, MIP



BLP模型

✓ BLP作为**经典模型**被写入了Springer教材 Methods in Molecular Biology 丛书之一《**Bioinformatics and Drug Discovery**》(第16章第287页)

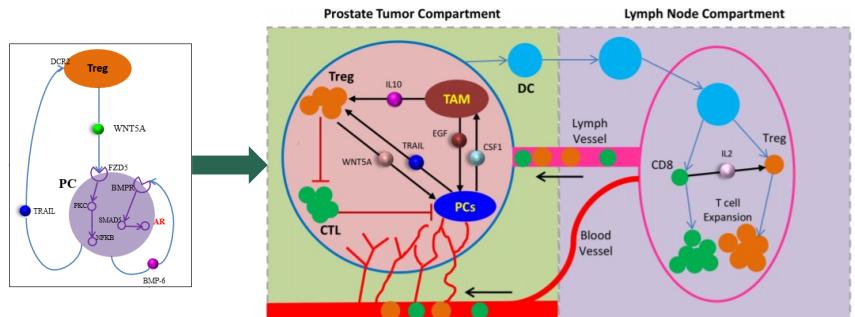
基于微分方程的连续时间建模方法



Cell-cell interaction模型

✓ 该模型被Matlab工具包CRA收录，并在BMC Bioinformatics进行长篇报道  
✓ 作为**经典模型**被写入Elsevier教材(2020年)：《Exploring Mathematical Modeling in Biology》(第2章第54页)

发现了**WNT5A**调控CRPC (前列腺癌) 进展的新机制  
构建了面向**分子-细胞-组织的3D多尺度建模方法HABM**



**“WNT5A在免疫调节中发挥重要作用”**



Development and validation of a prognostic immune-associated gene signature in clear cell renal cell carcinoma

Chengquan Shen<sup>a,1</sup>, Jing Liu<sup>b,1</sup>, Jirong Wang<sup>a</sup>, Xulong Zhong<sup>a</sup>, Dahai Dong<sup>a</sup>, Xiaokun Younghua Wang<sup>a,\*</sup>

Results: A total of 681 differentially expressed lAGs were identified and seven genes (IF20, WNT5A, IRBP, ACER, PLAUR, TIX, RDI) were finally selected in a lAGs signature. Survival analysis revealed that high lAGs risk scores were associated with poor overall survival and disease-free survival in breast cancer patients [13]. WNT5A belongs to the large WNT family of cysteine-rich secreted glycoproteins, which is involved in multiple signaling pathways that regulate a variety of cellular processes [14]. Ji et al. revealed that WNT5A could mediate the activation of Treg and TAM cells, which induced the immunosuppression during castration-resistant prostate cancer progression [15]. Chen

✓ **HABM是十几年来首个对肿瘤生长-免疫反应-血管生成进行3D时空建模的数学模型**

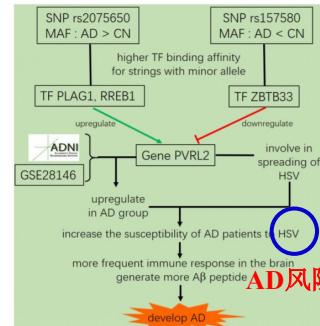


Digital Pathology Analysis Quantifies Spatial Heterogeneity of CD3, CD4, CD8, CD20, and FoxP3 Immune Markers in Triple-Negative Breast Cancer

Haoyang Mi<sup>a</sup>, Chang Gong<sup>a</sup>, Jeremias Sulman<sup>a</sup>, Elena J. Fertig<sup>a,1</sup>, Alexander S. Szalay<sup>a,1</sup>, Elizabeth M. Jaffee<sup>a,1</sup>, Vered Stearns<sup>a</sup>, Leisha A. Emens<sup>a</sup>, Ashley M. Cimino-Mathews<sup>a,2</sup> and Aleksander S. Popel<sup>a,3</sup>

of disease trajectories in response to intervention. On tissue-cellular scale, ABMs have been employed and used for spatially explicit simulations to investigate emergent behavior arising from interactions between cancer and immune cells, such as spatial and spatio-temporal variations in tumor morphology and immuno-architecture (Kim et al., 2009; Shi et al., 2014; Wells et al., 2015; Gong et al., 2017; Norton et al., 2017, 2019; Pourhasanzade et al., 2017; Hoehme et al., 2018; Ji et al., 2019).

组学大数据挖掘发现AD潜在风险因子 (TREM2, HSV-1)



Neuron Article

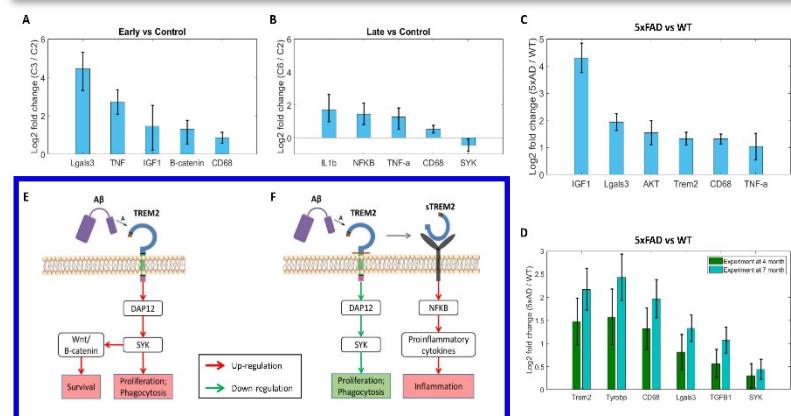
Human Herpesvirus 6 Detection in Alzheimer's Disease Cases and Controls across Multiple Cohorts

Mary Alice Allinott,<sup>1</sup> Kory Johnson,<sup>2</sup> David A. Bennett,<sup>2</sup> Sarah M. Connor,<sup>3</sup> Juan C. Troncoso,<sup>3</sup> Marilyn S. Albert,<sup>2</sup> Susan M. Resnick,<sup>2</sup> Sonja W. Scholtz,<sup>2,4</sup> Philip L. De Jager,<sup>2</sup> and Steven J.

是关于“**HSV-1是AD风险因子**”的较早报道之一

or plasma (Lövheim et al., 2018). In addition, several groups have identified overlap between AD genetic risk factors and genes affected by viral infection, such as a receptor involved in spreading HSV-1 (Liu et al., 2018) and a human leukocyte antigen (HLA) subtype associated with increased susceptibility to HHV-6A infection (Rizzo et al., 2019).

✓ **时序组学大数据挖掘，首次解析了TREM2调控Microglia表型转换的分子机制**



谢谢各位专家观看  
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