生物信息学

Bioinformatics

第八章 结构分析

生物信息学分析

- 根据研究对象及目的分类:
 - 1. macromolecular sequences;
 - 2. macromolecular structures;
 - 3. expression profiles; (microarrays;RNA-seq;2D-PAGE)
 - 4. biochemical network; (Interactions and reactions)
 - 5. evolution history.

生物信息学分析

章节	源数据	结果知识	种类	
	DNA序列	基因等特征序列	Soa	
四、序列分析*	蛋白质序列	特征域、特性	Seq.	
	EST	表达基因(mRNA)	Expr.	
五、系统发育分析	DNA/蛋白质序列	进化历史	Evol.	
六、基因组分析	基因组序列	基因位置、功能、 物种进化历史	Seq. Evol.	
(转录组分析)	Microarray	表达基因(mRNA)	Expr.	
	RNA-seq	表述基因(IIIKINA)	Lxpi.	
七、蛋白质组分析	2D-Page	表达基因(蛋白质)	Expr.	
	Y2-hybrid	蛋白质相互作用	Net.	
八、结构分析	蛋白质序列	蛋白质结构		
	RNA序列	RNA结构	Struct.	

蛋白质结构分析

- Basic notes
- Structural alignment of proteins
- Structure-based protein classification
- Structure prediction of proteins (Comparative Modeling*)

序列通过折叠为结构而最终 形成功能是使生命得以实现的 重要自然原理。

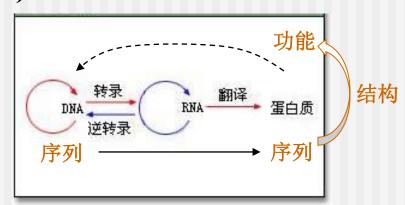
生命的信息内涵

 $Log_2 = 1 bit$

- 信息的产生: 多种可能状态下进行选择。
 - -生物信息的产生源于自然选择。\ ATCGC=? bits

■信息的"行为方式"——传递:一种选择引起 另一种选择:功能实现。

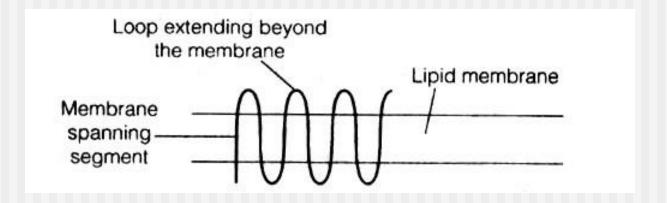
中心法则:



生物学研究的就是生物信息的产生和传递方式。

Structural types

- Globular proteins 核心疏水,表面亲水。
- Integral membrane proteins 膜内疏水,膜表面亲水。

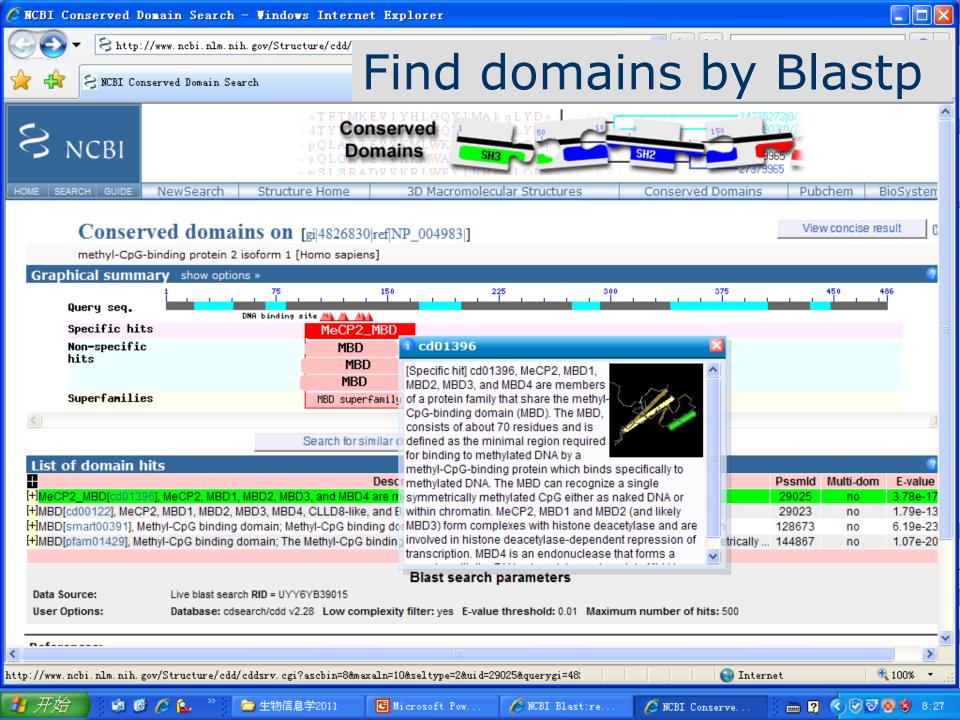


结构的局部特征 —— Domain

■序列特征域:

A domain (or module) is a protein region that adopt a particular three-dimensional structure.

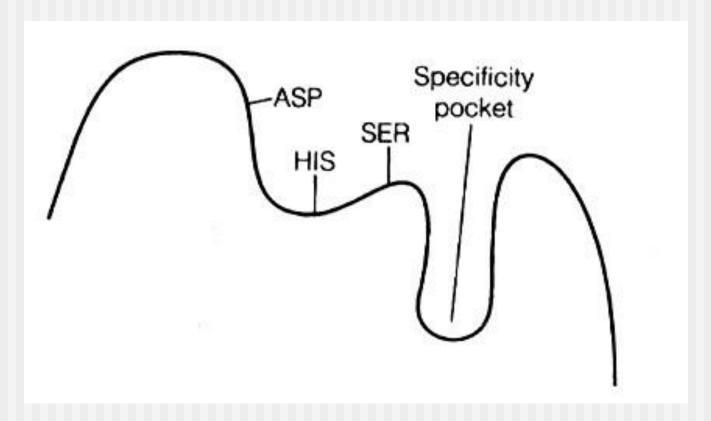
A motif (or fingerprint, or pattern) is a short conserved protein region, typically 10~20 aa.



What is a domain?

- 序列: A common subsequence observed to occur in many different proteins;
- 结构: A subsequence folding independently; a geometrically distinct substructure.
- 功能: A subsequence or substructure with a recognized function;

从结构到功能:



催化功能——形状+特定位置的特定aa



蛋白质结构源数据库---PDB



res RCSB Protein Data Bank - RCSB PDB -...

A MEMBER OF THE PDB An Information Portal to Biological Macromolecular Structures

As of Tuesday May 29, 2012 at 5 PM PDT there are 81957 Structures | PDB Statistics | 📓 🗟 🚱 🚔

Summary

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Structure of Herring Type II Antifreeze Protein

DOI:10.2210/pdb2py2/pdb

Primary Citation

Structure and evolutionary origin of Ca(2+)-dependent herring type II antifreeze protein.

Liu, Y.A. Li, Z.A. Lin, O.A. Kosinski, J.A. Seetharaman, J.A. Bujnicki, J.M.P., Sivaraman, J.P., Hew, C.L.P.,

Journal: (2007) PLoS ONE 2: e548-e548

PubMed: 17579720 2

PubMedCentral: PMC1891086 2

DOI: 10.1371/journal.pone.0000548 Search Related Articles in PubMed 🔎

PubMed Abstract:

In order to survive under extremely cold environments, many organisms produce antifreeze proteins (AFPs). AFPs inhibit the growth of ice crystals and protect organisms from freezing damage. Fish AFPs can be classified into five distinct types based on their structures.... [Read More & Search PubMed Abstracts]

‡ Molecular Description

Classification: Antifreeze Protein @

Structure Weight: 92991.48



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Simple Viewer Protein Workshop Kiosk

Hide

Biological assembly 1 assigned by authors



SITE	2 AC5	6 ASP E	114 HO	H E 932			<i>F</i> 1 1	t stat time
SITE	1 AC6	6 GLN F	92 AS	P F 94		R	公主 /	汀数排
SITE	2 AC6	6 ASP F	114 HO	H F 907	וטו		> 1	可从加口
CRYST1	31.27	9 146.41	5 192.4	06 90. O	0 90.00	90.00	P 21 21 21	24
ORIGX1	1.	000000 C	.000000	0.00000	0	0.00000		
ORIGX2	0.	000000 1	.000000	0.00000	0	0.00000		
ORIGX3	0.	000000 C	.000000	1.00000	0	0.00000		
SCALE1	0.	031970 C	.000000	0.00000	0	0.00000		
SCALE2	0.	000000 C	.006830	0.00000	0	0.00000		
SCALE3	0.	000000 0	.000000	0.00519	7	0.00000		
ATOM	1 N	CYS A	4	41.984	34.341	17.654	1.00 37.88	N
ATOM	2 C	A CYS A	4	41.522	34.320	19.073	1.00 38.27	С
ATOM	3 C	CYS A	4	40.007	34.287	19.172	1.00 38.11	С
ATOM	4 0	CYS A	4	39.306	34.731	18.264	1.00 38.43	0
ATOM	5 C	B CYS A	4	42.020	35.559	19.832	1.00 38.18	С
ATOM	6 S	G CYS A	4	43.805	35.582	20.183	1.00 40.36	S
ATOM	7 N	PRO A	5	39.480	33.755	20.284	1.00 37.81	N
ATOM	8 C	A PRO A	5	38.029	33.697	20.460	1.00 37.63	С
ATOM	9 C	PRO A	5	37.464	35.112	20.566	1.00 37.95	С
ATOM	10 0	PRO A	5	38.137	36.030	21.036	1.00 36.62	0
ATOM	11 C	B PRO A	5	37.868	32.890	21.749	1.00 37.89	С
ATOM	12 C	G PRO A	5	39.137	33.172	22.495	1.00 37.38	С
ATOM	13 C	D PRO A	5	40.172	33.094	21.404	1.00 37.61	С
ATOM	14 N	THR A	6	36.225	35.267	20.117	1.00 38.06	N
ATOM	15 C	A THR A	6	35.519	36.541	20.104	1.00 38.79	C
ATOM	16 C	THR A	6	36.154	37.767	20.777	1.00 38.89	C
ATOM	17 0	THR A	6	36.983	38.447	20.178	1.00 41.01	0

Summary

Sequence



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rom RCSB PDB - Jmol Viewer for 2PY2

PDB --- 结构视图

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Structure of Herring Type II Antifreeze Protein

Seq. Similarit

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*Right-click Jmol to view additional options. Drag the bottom-right corner to resize.







蛋白质结构分析

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结构直接支持功能

■ 结构的保守性可能超出序列比较的可识别 性 ——

Protein structures tend to be conserved even when evolution has changed the sequence beyond recognition.

序列相似性分析的意义

■进化关系推测

序列相似一般是由于进化同源, 也有例外。

- ■结构推测
- ■功能推测

序列及结构相似性比对的意义

■序列比对: 源序列与目标序列之间按残基位置相对排列。使序列之间的相似程度最大。

进化(同源) <==寻找序列相似物==>结构-----功能。

■结构比对:源结构和目标结构之间按残基位置相对排列,使结构之间相似性程度最大——相应残基的α碳原子空间位置最接近。

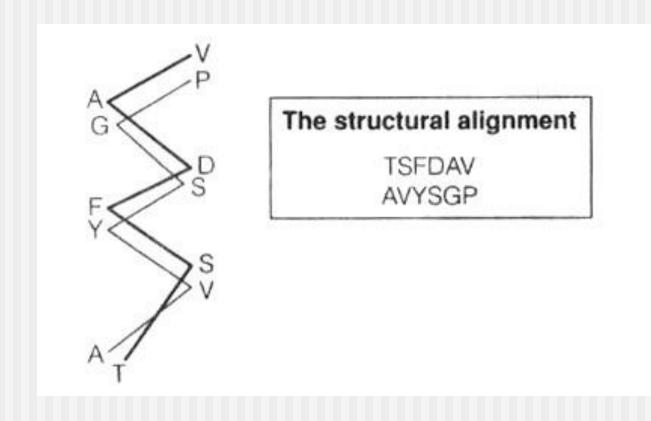
生物信息学分析

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六、基因组分析	基因组序列	基因位置、功能、 物种进化历史	Seq. Evol.	
(转录组分析)	Microarray	表达基因(mRNA)	Expr.	
	RNA-seq	表述基因(IIIKINA)	Lxpi.	
七、蛋白质组分析	2D-Page	表达基因(蛋白质)	Expr.	
	Y2-hybrid	蛋白质相互作用	Net.	
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	RNA序列	RNA结构	Struct.	

生物信息学分析

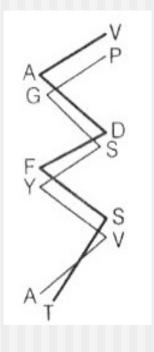
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	EST	表达基因(mRNA)	Expr.	
五、系统发育分析	DNA/RNA/	进化历史	Evol.	
	蛋白质序列		LVOI.	
	蛋白质结构	进化历史		
	蛋白质结构	蛋白质功能		
八、结构分析	蛋白质序列	蛋白质结构	Struct.	
	RNA序列	RNA结构		

A conceptual view of structural alignment



Root Mean Square Deviation (RMSD)

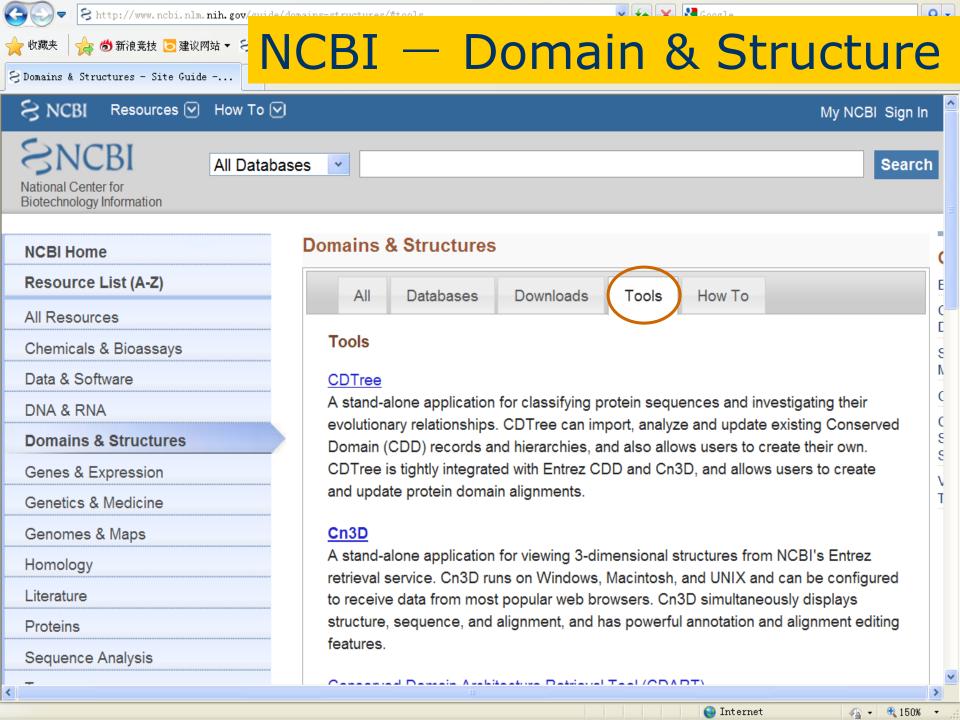
■相应残基α碳原子平均空间距离的度量。

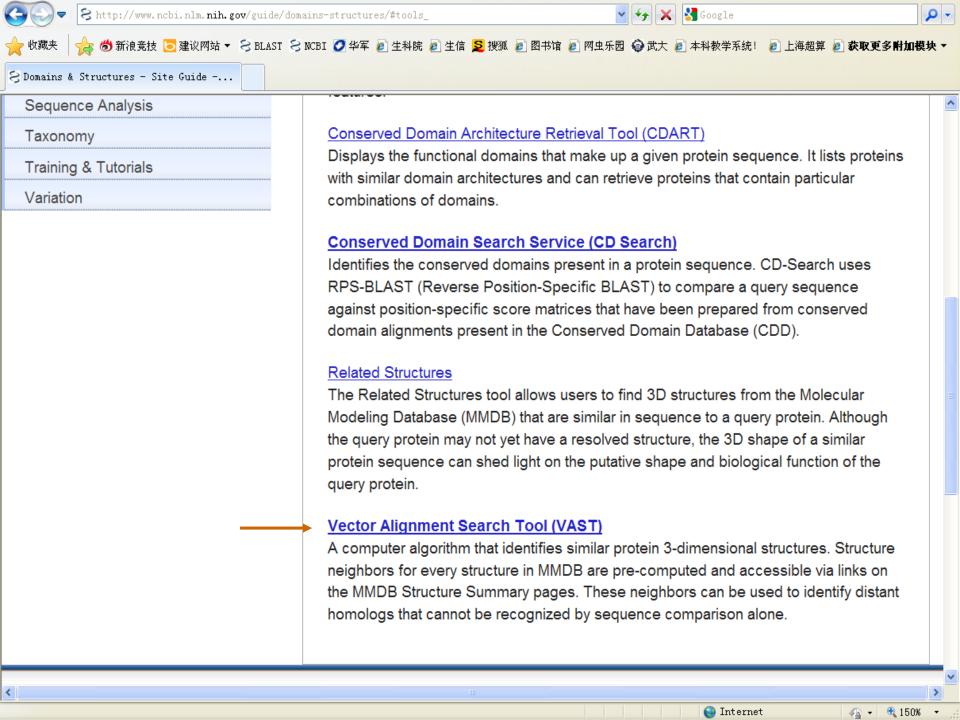


$$RMSD = \sqrt{\frac{1}{N} \sum_{i} d_{i}^{2}}$$

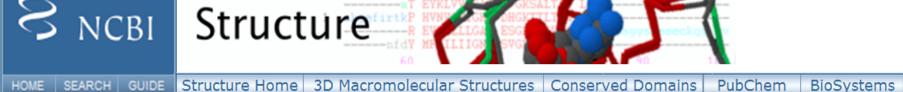
结构相似性衡量的重要指标

RMSD<1.5 埃 —— 非常相似









VAST: Vector Alignment Search Tool



ABOUT SEARCH HELP

About VAST

VAST, short for **V**ector **A**lignment **S**earch **T**ool, is a computer algorithm developed at NCBI and used to identify similar protein 3-dimensional structures by purely geometric criteria, and to identify distant homologs that cannot be recognized by sequence comparison.

VAST is applied on every protein in the Molecular Modeling Database (MMDB) during MMDB data processing in order to identify similar 3D structures. The pre-computed results are accessible from a structure's summary page; to retrieve them, you can either:

- 1. view the "show annotation" graphic for any protein molecule of interest on a structure summary page, then click on the bar graphic for the overall protein molecule or for any 3D domain it contains in order to view a list of structures that are similar in shape to the molecule or 3D domain you selected. The VAST Help document provides additional details and illustrated examples.
- 2. follow the link for "Similar Structures: VAST" in the upper right corner of a structure summary page to open a tabular list of the protein molecules and 3D domains in the structure. Then select the protein or 3D domain of interest to view a list of structures that are similar in shape to the region you selected.

Show "Similar Structures" for PDB ID or MMDB ID: 2TRX

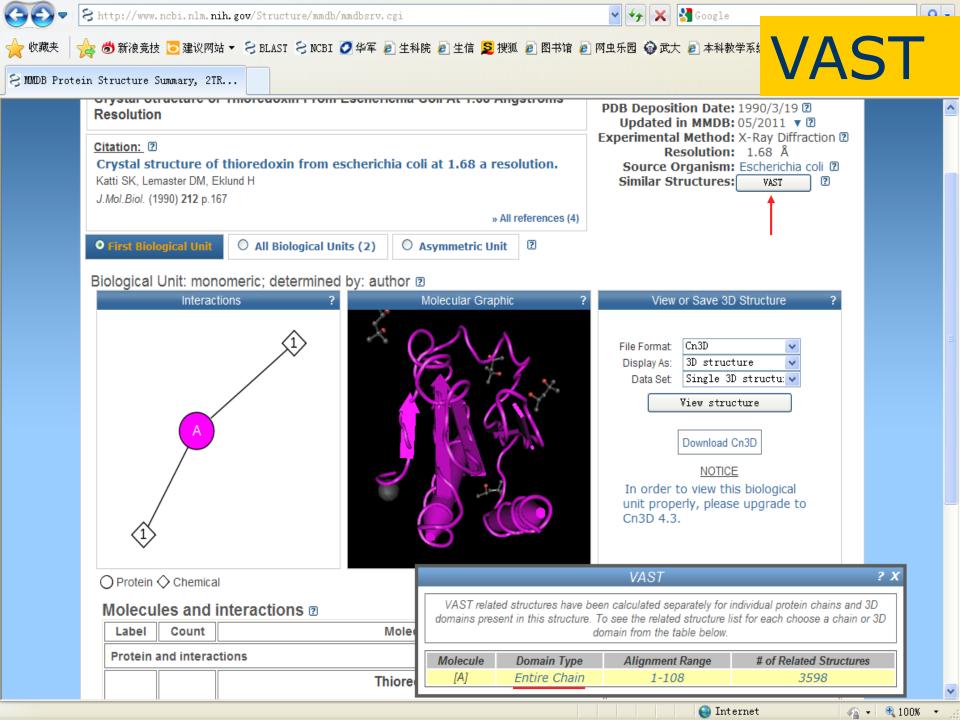


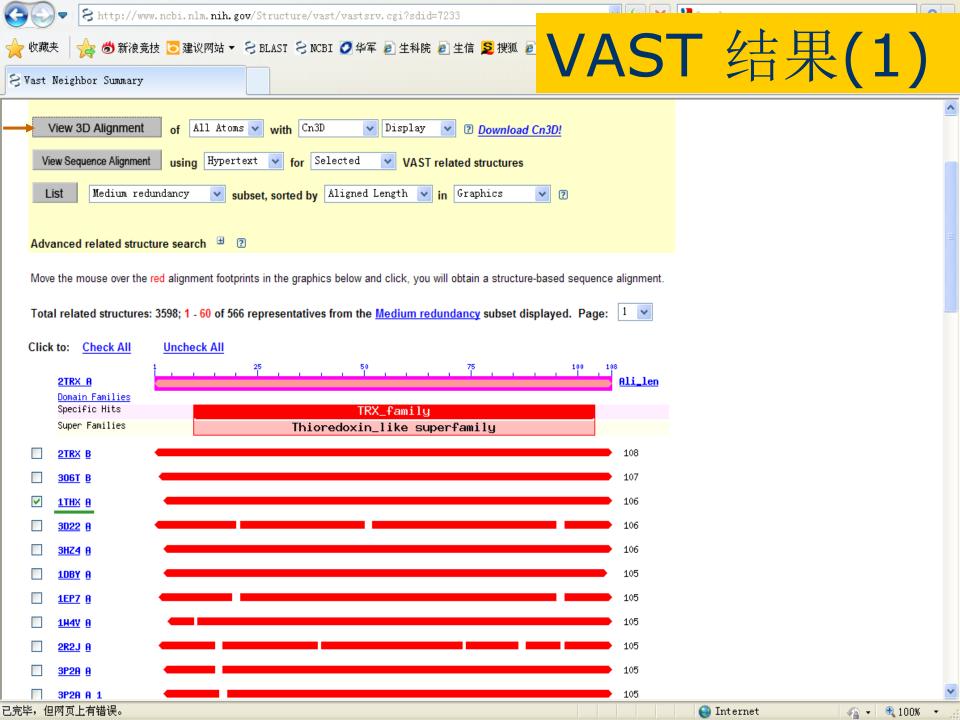


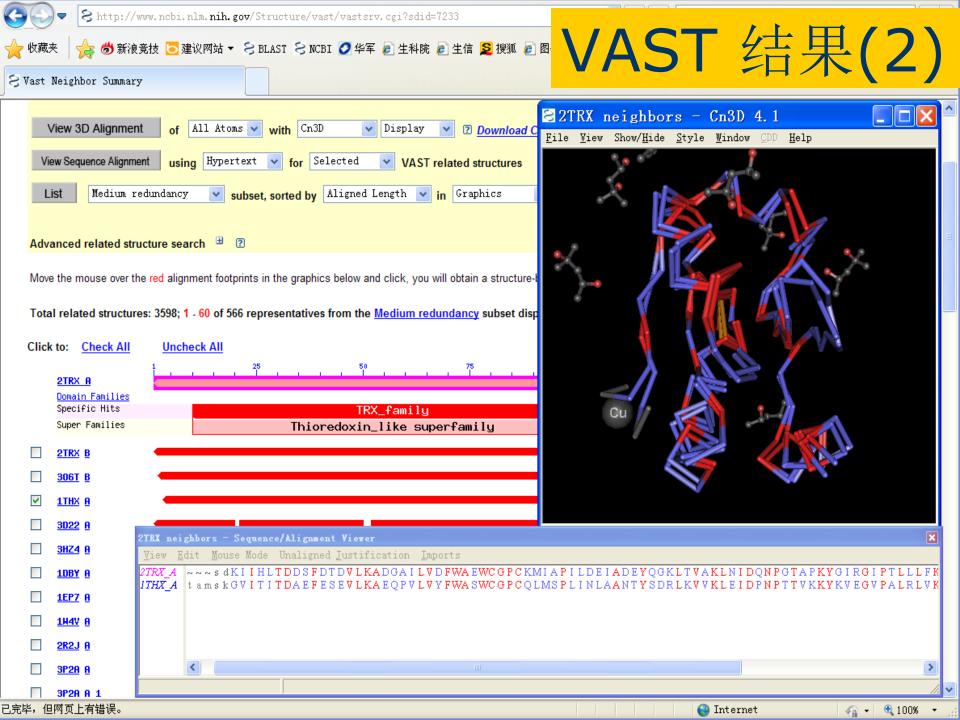
If you have a newly determined protein structure that is not yet in MMDB, then you can use the VAST Search service

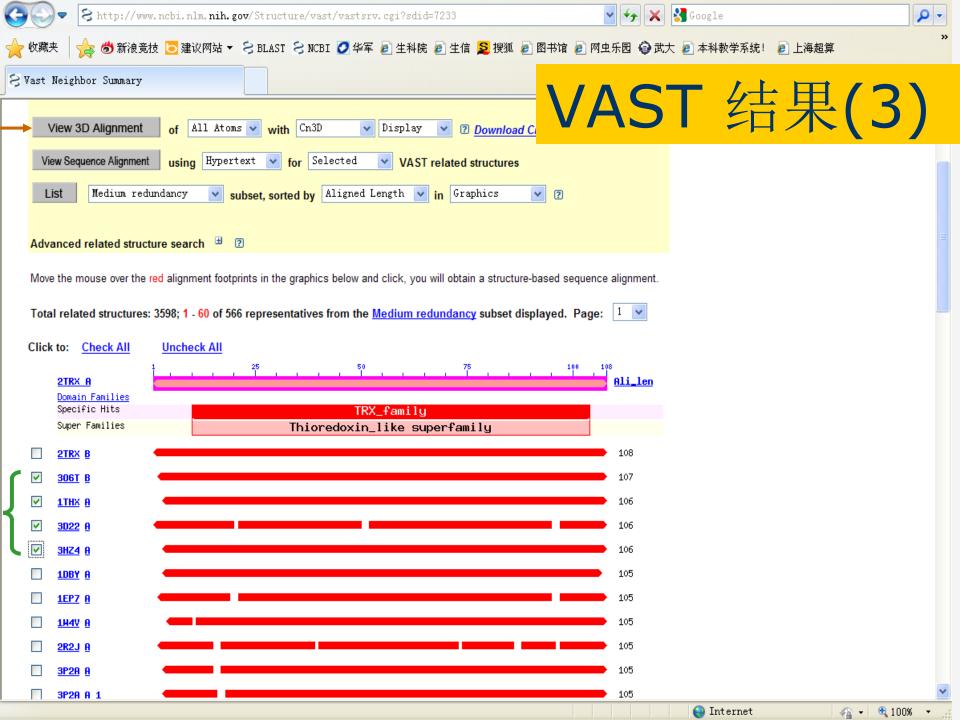
Internet

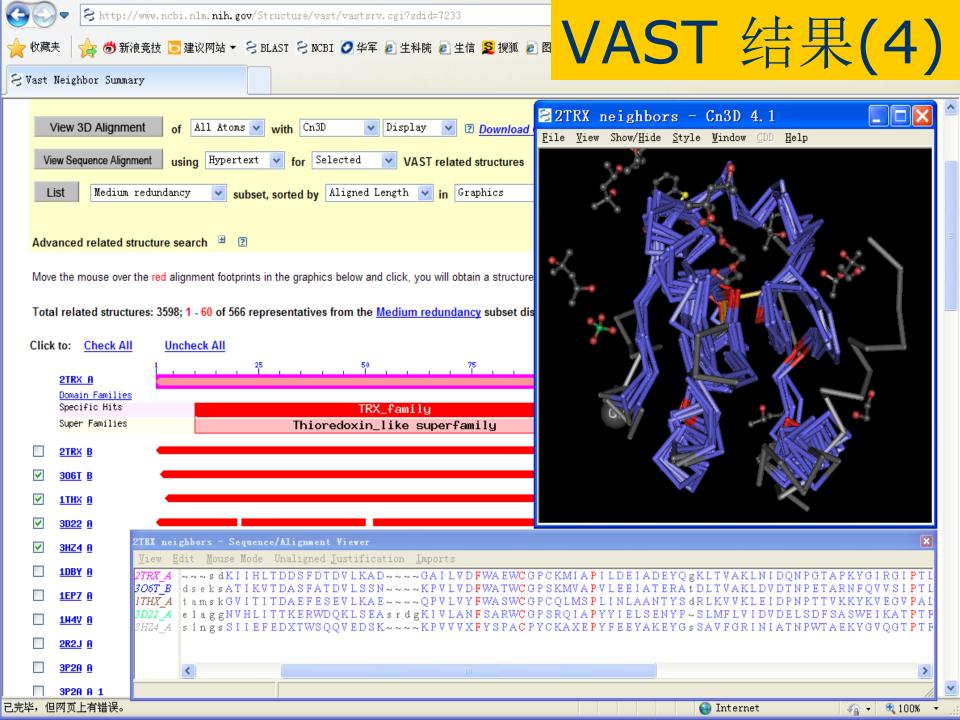
150%











蛋白质结构分析

- Basic notes
- Structural alignment of proteins
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Structure-based protein classification

---有些结构相似蛋白质的序列几乎看不出序列相似性。

Structure is more conserved, so structure-based classification is more powerful.

■ 分类等级*:

```
Families —— 近缘同源(序列相似性较明显);
Super-families —— 远缘同源;
Super-folds —— 有一定相似但不一定同源.
```

蛋白质结构分类数据库

- SCOP (Structure Classification Of Proteins) 依结构相似性进行分类,等级细化顺序为——Class,Folds、 Gene superfamily、Gene family。
- CATH (classification by Class, Architecture, Topology, and Homology)
 依结构相似性进行分类,等级细化顺序为——Class、Architecture、Topology、Homology superfamiliy、Sequence family。
- FSSP (Fold classification based on Structure-Structure alignment of Proteins) or (Families of Structurally Similar Proteins)
- SARF (Spatial ARrangement of backbone Fragments)

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The legacy SCOP websites can be accessed at SCOP 1.75 and SCOP2 prototype

SCOP 2

SCOP: Structural Classification of Proteins

Nearly all proteins have structural similarities with other proteins and, in some of these cases, share a SCOP database, created by manual inspection and abetted by a battery of automated methods, aims comprehensive description of the structural and evolutionary relationships between all proteins whose provides a broad survey of all known protein folds, detailed information about the close relatives of ar framework for future research and classification.

Latest update on 2022-06-29 includes 72,544 non-redundant domains representing 861,631 protein sand families statistics here.

Keyword and ID search

Sequence search









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151 million protein domains classified into 5,841 superfa

Search by keywords, PDB code, GO term, etc

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Core classification files for the latest version of CATH-Plus (v4.3) are now available to download. Daily updates of o



3D Structure

Find out what 3D structure your protein adopts



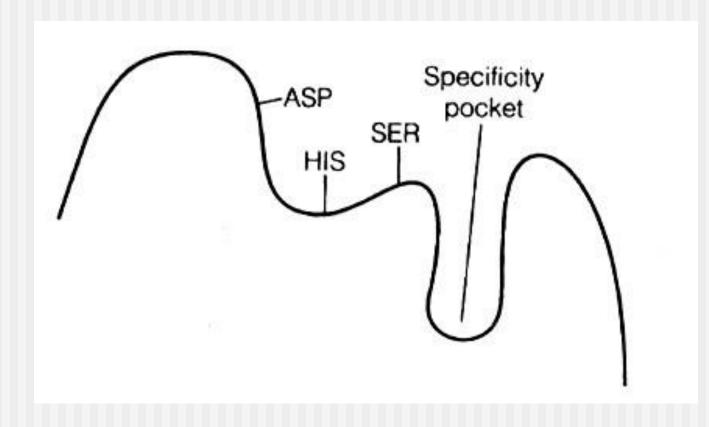
Protein Evolution

Learn about a particular protein family and how it evolved

蛋白质结构分析

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结构决定功能:

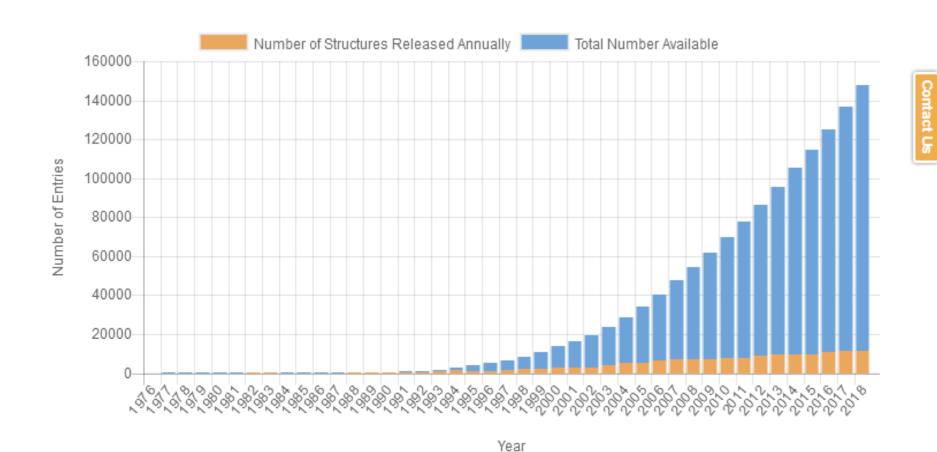


Show

entries

Other Statistics •

PDB Statistics: Overall Growth of Released Structures Per Year



Protein structure prediction

```
序列数据 / 结构数据
June 2000, 86500 in SwissProt / 12500 in PDB;
Feb 2004, 144731 in SwissProt / 24358 in PDB;
Mar 2006, 208005 in SwissProt / 35343 in PDB.
______
Jan 2017, 553231 in SwissProt / 116509 in PDB.
大量预测出的蛋白质序列.....
```

■ 根据序列预测结构——序列决定结构。

History

- Many of the first bioinformatics programs were written in order to "solve the protein folding problem".
- Even though the field is more than 40 years old, protein structure prediction continues to be one of the most active areas in all of bioinformatics research.

CASP (Critical Assessment of Structure Prediction) competitions



Protein Structure Prediction Center

Menu Home **PC Login PC Registration ▼CASP Experiments** CASP15 (2022) CASP14 (2020) CASP13 (2018) CASP12 (2016) CASP11 (2014) CASP10 (2012) CASP9 (2010) CASP8 (2008) CASP7 (2006) CASP6 (2004) CASP5 (2002) CASP4 (2000) CASP3 (1998) CASP2 (1996) CASP1 (1994) **▶** Initiatives Data Archive **Proceedings**

CASP Measures

Assessors
People

Success Stories From Recent CASPs

assembly modeling

modeling

template based template-based modeling ab initio modeling contact prediction help structur biologists

refinement

data-assi: modelii

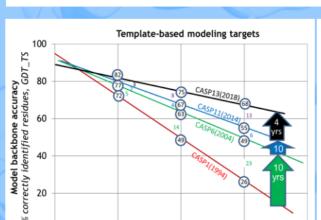
Models based on templates identified by sequence similarity remain the most accurate. Over the course of there have been enormous improvements in this area. However, the overall accuracy improvements that v 10 years of CASP remained unmatched until CASP12 (2016), when a new burst of progress happened [Kround In two years from 2014 to 2016, the backbone accuracy of the submitted models improved more than in the next CASP continued the trend [Croll et al, 2019], and the 2014-2018 model accuracy improvement of

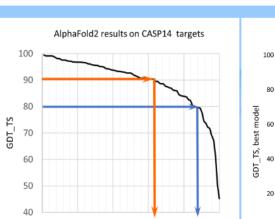
The next CASP continued the trend [Croll et al, 2019], and the 2014-2018 model accuracy improvement of 2014 (see left plot). Several factors contributed to this, including more accurate alignment of the target se available templates, combining multiple templates, improved accuracy of regions not covered by template of models, and better selection of models from decoy sets due to improved methods for estimation of models.

CASP14 marked an extraordinary increase in the accuracy of the computed three-dimensional protein stru

emergence of the advanced deep learning method AlphaFold2. Models built with this method proved to be

experimental accuracy (GDT_TS>90) for ~2/3 of the targets and of high accuracy (GDT_TS>80) for almost (middle plot). The accuracy of CASP14 models for TBM targets significally superseeded accuracy of models simple transcription of information from templates, and reached the level of GDT_TS=92 on average, which than the corresponding averages in previous two CASPs (right plot).





80 63.2,

Prediction methods

Ab initio:

Ab initio prediction

Knowledge-based:

Comparative modeling (Homology modeling) * Fold recognition (Threading)

Knowledge-trained:

Secondary structure prediction

Ab initio prediction

- Method: it proceed from fundamental physical principles, involving <u>quantum mechanics</u> and <u>statistical thermodynamics</u> --- minimizing free energy.
- Difficult: proteins plus solvent molecules --- too large the system scale for calculation; adopting approximation to capture the essentials of the folding problem.
- Interesting: from an <u>intellectual</u> viewpoint; would be <u>a huge scientific achievement</u>; a challenge for bioinformatics.

Prediction methods

Ab initio:

Ab initio prediction

Knowledge-based:

Comparative modeling (Homology modeling) Fold recognition (Threading)

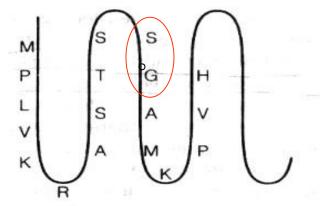
Knowledge-trained:

Secondary structure prediction

Comparative modeling (Homology modeling)

■ **Theoretical basis:** Sequences with more than 25% identity over an alignment of 80 residues or more adopt the same basic structure.

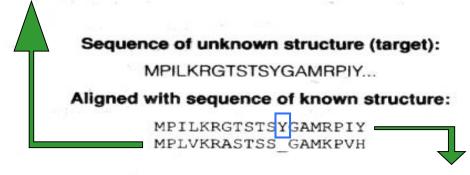
序列比对:进化(同源) <==寻找序列相似物==>结构—功能 结构比对:进化(同源) <----- 寻找结构相似物==>功能



Sequence of known structure:

MPLVKRASTSSGAMKPVH...

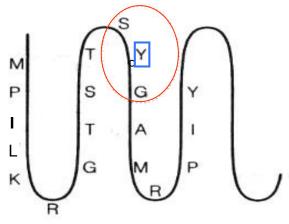
Schematic of known 3D structure (template)



Predict the 3D structure of the target sequence by replacing the old residues on the known structure with those from the target sequence made equivalent by the alignment.

Need to position new side chain atoms.

Small structural changes where gaps are found in the alignment (loop modelling, dotted circle).



Predicted 3D structure

Fig. 1. Comparative modeling.

Preparing work

Finding template structures:

BLAST in the database of sequences whose structures are already known by experimental means. (--- PDB)

(Multiple alignment of the structuretemplate sequences with the target sequence.)

Modeling

Backbone determination (a carbon)
 Placement of corresponding residues;
 ---- according to structure templates
 Loop modeling for insertions or deletions;
 ---- spare parts algorithm; from a special loop library

Side-chain positioning

```
---- conserved: according to structure templates, varied: using other sophisticated algorithms.
```

Model refinement.

```
---- energy minimization.....
```

Modeling

Backbone determination (a-carbon)
 Placement of corresponding residues;
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 ---- spare parts algorithm; from a special loop library

Side-chain positioning

```
---- conserved: according to structure templates, varied: using other sophisticated algorithms.
```

Model refinement.

```
---- energy minimization.....
```

Accuracy

- ■序列相似性低于40% --- 手工介入变得十分重要,否则预测 "can fail very badly"。
- 一般来说,30%以下的的序列相似性,用 这种方法很不可靠。

Theoretical basis: Sequences with more than 25% identity over an alignment of 80 residues or more adopt the same basic structure.

Prediction methods

Ab initio:

Ab initio prediction

Knowledge-based:

Comparative modeling (Homology modeling) Fold recognition (Threading)

Knowledge-trained:

Secondary structure prediction

Fold recognition (Threading):

- A query sequence is tried to thread through a known structure to see how well it might fit.
- For an arrangement, observe how they match up with respect to properties that affect proteinfolding such as whether it is hydrophobic or hydrophilic at particular points.
- The process is repeated for every other known structure in the database (fold-library) and finally, results are compared to determine which one is the most likely structure of the query protein.
- -- 和同源模建类似,也限于有已知的蛋白质结构,但原则上并不要求序列十分相似,如可低于25%(同源模建的底限)。

Prediction methods

Ab initio:

Ab initio prediction

Knowledge-based:

Comparative modeling (Homology modeling) Fold recognition (Threading)

Knowledge-trained:

Secondary structure prediction

Secondary structure prediction

- Predicting the conformational state of each residue in three categories, <u>helical</u>, <u>strand</u>, <u>and</u> <u>coil</u>, usually based on ideas reflecting the <u>preference</u> of a residue for a particular secondary structure.
- Accuracy: (early) 60%, (+conserved domains) 66%, (+structural data + sophisticated algorithms) >70%.
- Not for integral membrane proteins (需要专门的算法).

TABLE 10-4	Some Physical	Properties o	f Protein
-------------------	---------------	--------------	-----------

Property	Classical Method	Example		
Amino acid motifs		PDZ domain (e.g., nitric oxide synthase), coiled-coil domain (e.g., hemagglutinin, syntaxin, SNAP-25, myosin)		
Isoelectric point (pI)	Derived from isoelectric focusing	tro mograture partition reconstant		
Molecular weight	Derived from Stokes radius and sedimentation coefficient	extension deviation I applier also		
Posttranslational modifications: phosphorylation	Enzymatic analyses	Synapsin		
Posttranslational modifications: glycosylation	Enzymatic analyses	Nerve growth factor, neural cell adhesion molecule		
Posttranslational modifications: isoprenylation	Biochemical analyses	Lamin B, G protein γ subunits, $rab3A$		
Posttranslational modifications: palmitoylation	Biochemical analyses	β-Adrenergic receptor, GAP-43, insulin receptor, rhodopsin, nAChR		
Posttranslational modifications: myristoylation	Biochemical analyses	PKA, $G_{i\alpha}$ -subunit, MARCKS protein, calcineurin		
Posttranslational modifications: GPI- anchored proteins	Enzymatic analyses	Alkaline phosphatase, thy-1, prion protein, 5'-nucleotidase, uromodulin		
Sedimentation coefficient	Derived from sucrose density gradients	基于蛋白质序列的相关特		
Stokes radius	Derived from gel filtration	性分析对后续其结构和功		
Transmembrane domain	Derived from subcellular fractionation	能分析可能十分有用。		

Abbreviations: G protein, guanosine triphosphate-binding protein; GAP-43, growth-associated protein of 43 kDa; MARCKS, myristoylated alanine-rich C-kinase substrate; nAChR, nicotinic acetylcholine receptor; PDZ domain, post-synaptic density protein PSD-95, the *Drosophila* tumor suppressor discs-large, tight-junction protein ZO-1; PKA, protein kinase A; SNAP-25, synaptosomal-associated protein of 25 kDa; Rab3A, rat brain GTP-binding protein 3A; thy-1, thymocyte-1.

(Physical properties)

Strategy

Preliminary sequence analysis

Prediction methods

- Ab initio:
- 3 --- Ab initio prediction
 - Knowledge-based:
- 1 Comparative modeling (Homology modeling)
- 2 Fold recognition (Threading)
 - Knowledge-trained:
- 2 Secondary structure prediction

AlphaFold: Using AI for scientific discovery

Today we're excited to share DeepMind's first significant milestone in demonstrating how artificial intelligence research can drive and accelerate new scientific discoveries. With a strongly interdisciplinary approach to our work, DeepMind has brought together experts from the fields of structural biology, physics, and machine learning to apply cutting-edge techniques to predict the 3D structure of a protein based solely on its genetic sequence.

Our system, AlphaFold, which we have been working on for the past two years, builds on

years of prior research in using vast genomic data to predict protein structure. The 3D Deep Mind.com uses cookies to help give you the best possible user experience and to allow us to see models of proteins that Alphar Gid generates are far more accurate than any that have how the site is used. By using this site, you agree that we can set and use these cookies. For more come before—making significant progress on one of the core challenges in biology.



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rental techniques like diyo-electron inicroscopy, nuclear magnetic

y crystallography, but each method depends on a lot of trial and error, ars and cost tens of thousands of dollars per structure. This is why

biologists are turning to AI methods as an alternative to this long and laborious process for difficult proteins.

Fortunately, the field of genomics is quite rich in data thanks to the rapid reduction in the

We're proud to be part of what the CASP organisers have called "unprecedented progress in the ability of computational methods to predict protein structure," placing **first** in rankings among the teams that entered (our entry is A7D).

Our team focused specifically on the hard problem of modelling target shapes from scratch, without using previously solved proteins as templates. We achieved a high degree of accuracy when predicting the physical properties of a protein structure, and then used two distinct methods to construct predictions of full protein structures.

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CASP (Critical Assessment of Structure Prediction) competitions



Protein Structure Prediction Center



Menu Home PC Login PC Registration **CASP Experiments** CASP14 (2020)

CASP Commons (COVID-19, 2020)

CASP13 (2018)

CASP12 (2016)

CASP11 (2014)

CASP10 (2012)

CASP9 (2010)

CASP8 (2008)

CASP7 (2006)

CASP6 (2004)

CASP5 (2002)

CASP4 (2000)

CASP3 (1998)

CASP2 (1996)

CASP1 (1994)

Initiatives

Data Archive

Deccodings

Success Stories From Recent CASPs

contact prediction

[Schaarschmidt et al. 2018]

The most notable progress in recent CASPs (2014,

2016) resulted from sustained improvement in methods

for predicting three-dimensional contacts between pairs

of residues in structures. Average precision of the best

CASP12 contact predictor almost doubled compared to

that of the best CASP11 predictor (from 27% to 47% -

see the plot). Advances in the field as a whole are not

any less impressive: 26 methods in CASP12 showed better results than the best method in CASP11.

structural refinement

dataassisted ||

> not able to make happen this tim around ...

Dear CASPers, Again, many

thanks for your participation in th

to date)! One thing that we were

CASP14 meeting (and making it the most attended CASP meeting

Message Board

CASP14 Job Fair

CASP14 Program and updates

Dear CASPers, As you might have noticed, we had posted the CASP14 conference program online vesterday. It is available from the http://predictioncenter.org/casp1 web page. The CASP14 results w

Theoretical advance in contact prediction lead to improved accuracy of 3D models, especially for the hardest template-free modeling cases (see models for CASP12 target T0915 below).

CASP13 (2018) registered yet another leap in accuracy of contact prediction, with the average precision of the best contact prediction group increasing by 23% (compared to CASP12) and reaching 70%.

Early registration deadline and instructions for poster uploads

Dear CASPers, If you plan to attend the CASP14 virtual meeting, please register. We will have limited capacity to process registrations during the conference and the access to sessions will not be ...

prediction contact

C A S P 2 0 8

#	GR code	GR name	Domains Count	SUM Zscore (>-2.0)	Rank SUM Zscore (>-2.0)	AVG (>-2.
1	043	A7D ←	103	119.8429	1	1.1635
2	322	Zhang	103	106.7674	2	1.0366
3	089	MULTICOM	103	98.6232	3	0.9575
4	145	QUARK	103	90.1429	4	0.8752
5	261	Zhang-Server	103	88.1913	5	0.8562
6	460	McGuffin	103	80.7632	6	0.7841
7	354	wfAll-Cheng	103	76.8791	7	0.7464
8	135	SBROD	101	71.5034	9	0.7476
9	324	RaptorX-DeepModeller	103	75.4289	8	0.7323
10	197	MESHI	103	70.2323	10	0.6819
11	274	MUFold	103	66.9621	13	0.6501
12	222	Seok-refine	103	63.5832	15	0.6173
13	055	VoroMQA-select	103	67.2288	12	0.6527
14	196	Grudinin	103	68.8019	11	0.6680
15	192	Elofsson	102	62.8775	16	0.6361
16	086	BAKER	101	57.8510	18	0.6124
17	224	Destini	103	64.7773	14	0.6289
18	418	Seder3nc	103	40.3575	28	0.3918
19	344	Kiharalab	103	60.8535	17	0.5908
20	208	KIAS-Gdansk	99	49.0663	22	0.5764
21	406	Seder3mm	103	49.6085	21	0.4816
22	221	RaptorX-TBM	103	55.1204	19	0.5351

CASP13--2018

CASP3 (1998) CASP2 (1996) CASP1 (1994)

▶ Initiatives

Data Archive

Proceedings

CASP Measures

Feedback

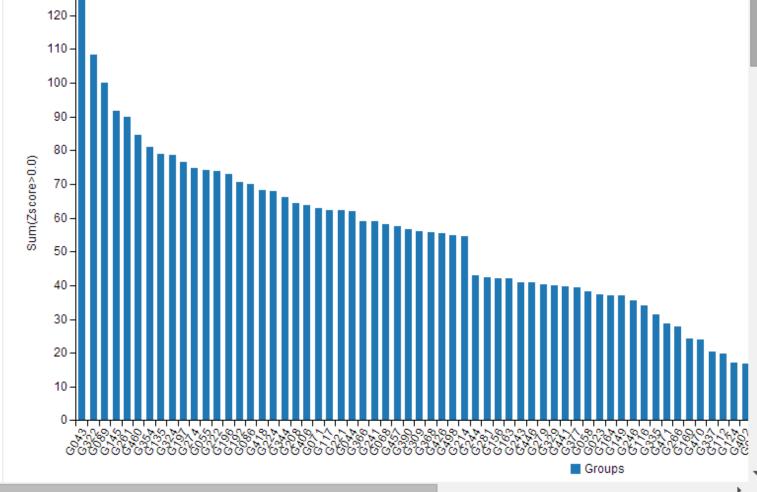
Assessors

People

Community Resources

Job Fair





CASP (Critical Assessment of Structure Prediction) competitions



Protein Structure Prediction Center

Menu Home **PC Login PC Registration ▼CASP Experiments** CASP15 (2022) CASP14 (2020) CASP13 (2018) CASP12 (2016) CASP11 (2014) CASP10 (2012) CASP9 (2010) CASP8 (2008) CASP7 (2006) CASP6 (2004) CASP5 (2002) CASP4 (2000) CASP3 (1998)

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Assessors **People**

Success Stories From Recent CASPs

assembly

template-based modeling

there have been enormous improvements in this area. However, the overall accuracy improvements that v 10 years of CASP remained unmatched until CASP12 (2016), when a new burst of progress happened [Kn

In two years from 2014 to 2016, the backbone accuracy of the submitted models improved more than in t The next CASP continued the trend [Croll et al, 2019], and the 2014-2018 model accuracy improvement of 2014 (see left plot). Several factors contributed to this, including more accurate alignment of the target se available templates, combining multiple templates, improved accuracy of regions not covered by template of models, and better selection of models from decoy sets due to improved methods for estimation of models

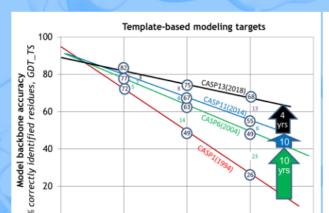
Models based on templates identified by sequence similarity remain the most accurate. Over the course of

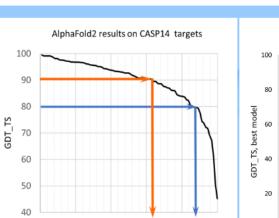
modeling template based

experimental accuracy (GDT TS>90) for ~2/3 of the targets and of high accuracy (GDT TS>80) for almost (middle plot). The accuracy of CASP14 models for TBM targets significally superseeded accuracy of models simple transcription of information from templates, and reached the level of GDT TS=92 on average, which than the corresponding averages in previous two CASPs (right plot).

CASP14 marked an extraordinary increase in the accuracy of the computed three-dimensional protein stru

emergence of the advanced deep learning method AlphaFold2. Models built with this method proved to be





CASP2 (1996) CASP1 (1994)

C A S P 2 0 2

0

#	GR code	GR name	Domains Count	SUM Zscore (>-2.0)	Rank SUM Zscore (>-2.0)	AVG Zscore (>-2.0)	Rank AVG Zs (>-2.0)
1	427	AlphaFold2	92	244.0217	1	2.6524	1
2	473	BAKER	92	90.8241	2	0.9872	2
3	403	BAKER-experimental	92	88.9672	3	0.9670	3
4	480	FEIG-R2	92	72.5351	4	0.7884	4
5	129	Zhang	92	67.9065	5	0.7381	5
6	009	tFold_human	92	61.2858	7	0.6661	8
7	420	MULTICOM	92	63.2689	6	0.6877	7
8	042	QUARK	92	60.0226	10	0.6524	11
9	324	Zhang-Server	92	60.8875	8	0.6618	9
10	488	tFold-IDT_human	92	57.6435	11	0.6266	12
11	368	tFold-CaT_human	92	60.5423	9	0.6581	10
12	334	FEIG-R3	92	48.4424	20	0.5265	23
13	039	ropius0QA	92	55.7086	12	0.6055	13
14	293	MUFOLD_H	92	47.7806	21	0.5194	24
15	031	Zhang-CEthreader	92	49.5742	18	0.5389	21
16	032	MESHI	92	53.0953	14	0.5771	15
17	216	EMAP_CHAE	92	53.1597	13	0.5778	14
18	209	BAKER- ROSETTASERVER	92	46.1861	25	0.5020	28
19	379	Wallner	92	51.7365	15	0.5624	16
20	498	VoroMQA-select	92	51.4288	17	0.5590	18
21	220	McGuffin	92	49.5443	19	0.5385	22
22	252	Phattachania	02	E4 E220	16	0.5600	17

CASP14--2020

CASP2 (1996) CASP1 (1994)

▶ Initiatives

Data Archive

Proceedings

CASP Measures

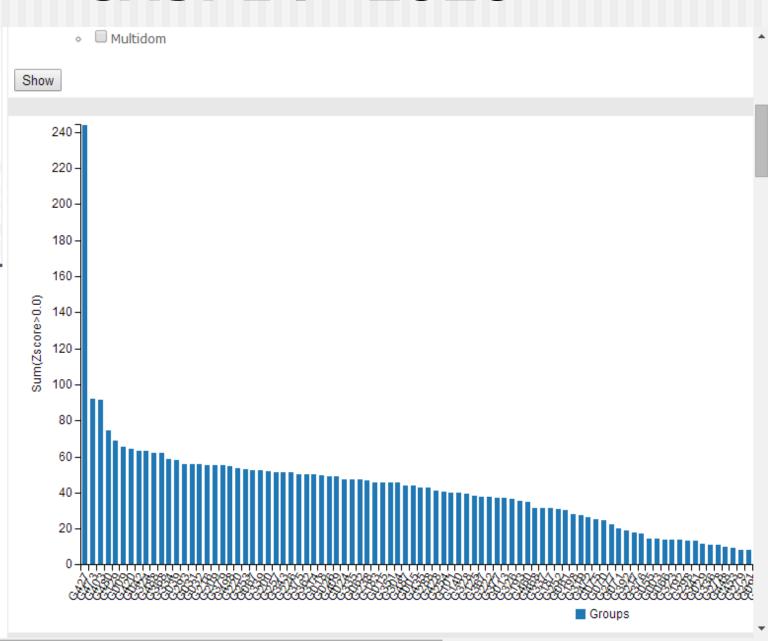
Feedback

Assessors

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Community Resources

<u>Job Fair</u>



AlphaFold

蛋白质三维结构预测是生物学最严峻的 挑战之一。继围棋、国际象棋等竞技项目之 后. 近日谷歌旗下DeepMind开发的人工智能 程序AlphaFold在两年一次的蛋白质结构预测 挑战赛CASP中再次大幅胜出。该程序在根据 蛋白质氨基酸序列确定蛋白质三维结构方面 取得巨大飞跃,准确性可与冷冻电子显微术 (又称冷冻电镜) (Cryo-EM) 和X-射线晶体 学等实验技术相媲美。

第一代AlphaFold依托蛋白质数据库PDB作为 训练数据集,构建神经网络,采用深度学习预测氨 基酸残基间的方向和距离,混合传统算法Rosetta 对蛋白质结构进行同源建模、结构优化;与此不同 的是,第二代AlphaFold则将折叠蛋白质视为"空 间图",基于神经网络系统进行"端到端"的训练。 使用了进化相关的氨基酸序列, 多序列比对以及对 氨基酸对的评估来优化结构预测。研究人员使用蛋 白质数据库中接近17万个不同的蛋白质结构。通过 不断地迭代。AlphaFold系统学习到了基于氨基酸 序列精确预测蛋白结构的能力。这一基于原子坐标 近乎"暴力"的算法是全新的途径,是全新算法与 强大算力的强强联合。

AlphaFold

正如马里兰大学帕克分校计算生物学家, CASP共同创始人John Moult所言, 从某种程 度上而言, 结构预测问题得到了解决。根据 氨基酸序列准确预测蛋白质结构的能力将对 生命科学和医学带来巨大的好处。这将极大 地加快对细胞组成模块的理解. 对于更快更 先进的药物发现显然有很大帮助。Nature使 用"它将会改变一切"来报道这一关键成 果……

History

- Many of the first bioinformatics programs were written in order to "solve the protein folding problem".
- Even though the field is more than 40 years old, protein structure prediction continues to be one of the most active areas in all of bioinformatics research.

Article

Highly accurate protein structure prediction with AlphaFold 预测人类蛋白质组的结构

https://doi.org/10.1038/s41586-021-03819-2

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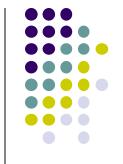
Open access



2021, 7, 15 Nature John Jumper^{1,4, Richard}, Richard Evans^{1,4}, Alexander Pritzel^{1,4}, Tim Green^{1,4}, Michael Figurnov^{1,4}, Olaf Ronneberger^{1,4}, Kathryn Tunyasuvunakool^{1,4}, Russ Bates^{1,4}, Augustin Žídek^{1,4}, Anna Potapenko^{1,4}, Alex Bridgland^{1,4}, Clemens Meyer^{1,4}, Simon A. A. Kohl^{1,4}, Andrew J. Ballard^{1,4}, Andrew Cowie^{1,4}, Bernardino Romera-Paredes^{1,4}, Stanislav Nikolov^{1,4}, Rishub Jain^{1,4}, Jonas Adler¹, Trevor Back¹, Stig Petersen¹, David Reiman¹, Ellen Clancy¹, Michal Zielinski¹, Martin Steinegger^{2,3}, Michalina Pacholska¹, Tamas Berghammer¹, Sebastian Bodenstein¹, David Silver¹, Oriol Vinyals¹, Andrew W. Senior¹, Koray Kavukcuoglu¹, Pushmeet Kohli¹ & Demis Hassabis^{1,4} ≅

Proteins are essential to life, and understanding their structure can facilitate a mechanistic understanding of their function. Through an enormous experimental effort¹⁻⁴, the structures of around 100,000 unique proteins have been determined⁵, but this represents a small fraction of the billions of known protein sequences^{6,7}. Structural coverage is bottlenecked by the months to years of painstaking effort required to determine a single protein structure. Accurate computational approaches are needed to address this gap and to enable large-scale structural bioinformatics. Predicting the three-dimensional structure that a protein will adopt based solely on its amino acid sequence—the structure prediction component of the 'protein folding problem'⁸—has been an important open research problem for more than 50 years⁹. Despite recent progress^{10–14}, existing methods fall far short of atomic accuracy, especially when no homologous structure is available. Here we provide the first computational method

蛋白质组 (Proteome)



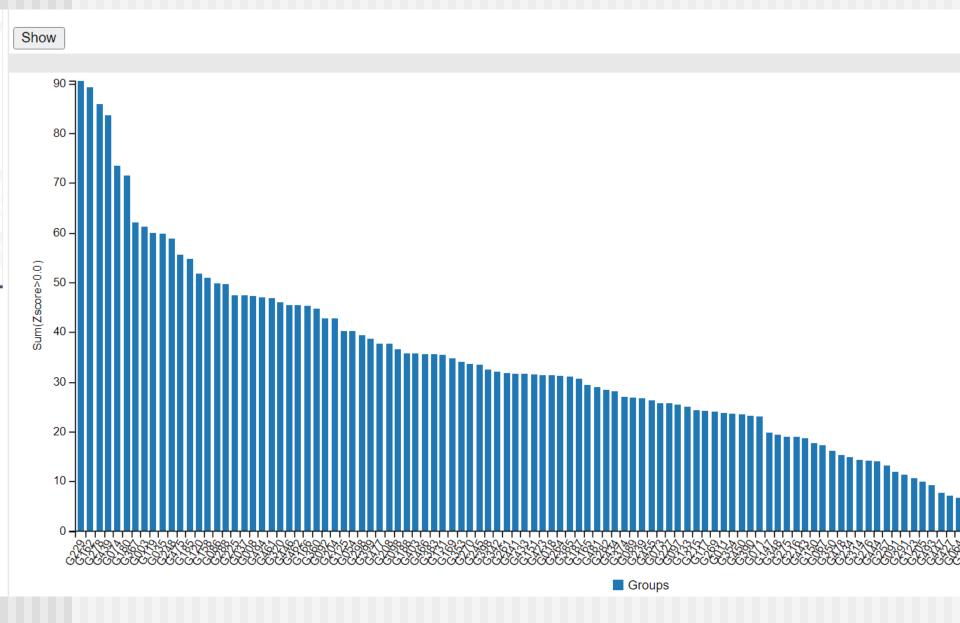
一个基因组表达产生的所有蛋白质的 总体。(抽象)一意义不大。

"the complete set of proteins expressed by an organism"

在某种内在和外在条件下,一个基因组表达产生的所有蛋白质的总体。(具体) — 通常所指。

"the complete set of proteins expressed by a cell or a tissue in a definitive situation"

CASP15--2022



- Two years later, AlphaFold still dominates the competition.
- Deepmind itself did not participate in this round, but AlphaFold has been open source since 2021 and the most successful participants have integrated Deepmind's Al system into their approaches.

In predicting the shape of individual proteins, participating teams achieved moderate improvements in accuracy. "The accuracy is already so high that it's hard to improve on it," ...

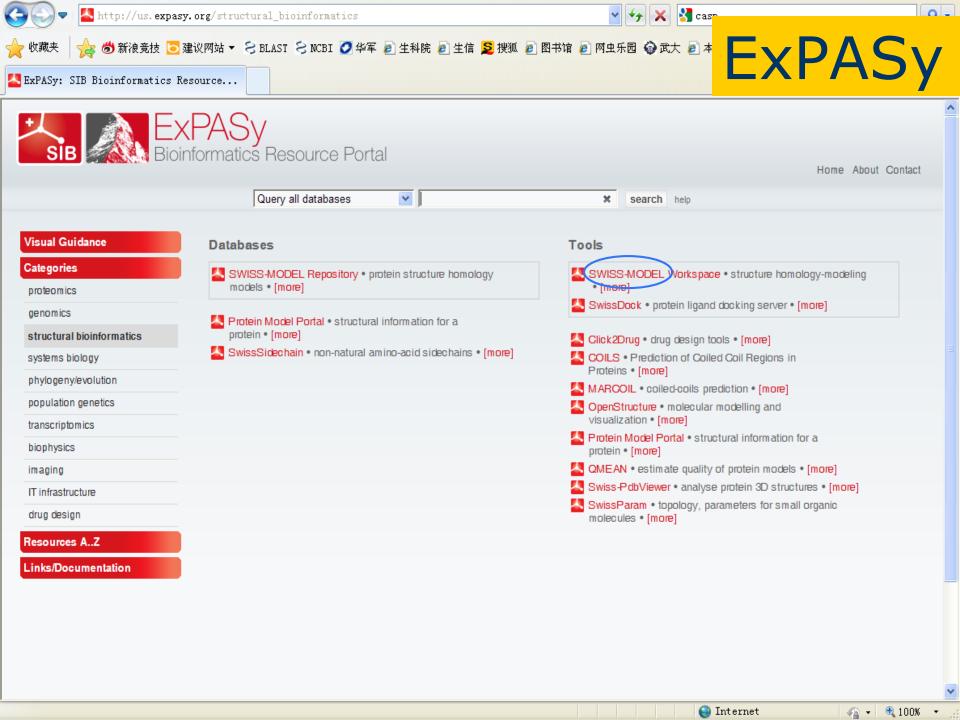
Aside from AlphaFold's proven capabilities, several teams this year also demonstrated how the Al system can be used with modifications to predict protein interactions. Compared to CASP14, systems using such AlphaFold variants have made significant improvements and are slowly approaching the accuracy of experimental methods.

Strategy

Preliminary sequence analysis

Prediction methods

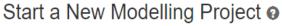
- Ab initio:
- 3 --- Ab initio prediction
 - Knowledge-based:
- 1 Comparative modeling (Homology modeling)
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 - Knowledge-trained:
- 2 Secondary structure prediction

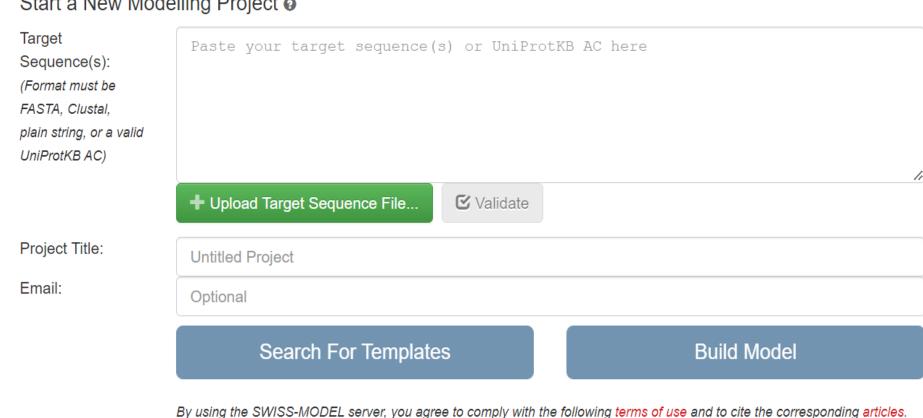




Modelling

Repository T





You are currently not logged in - to take advantage of the workspace, please log in or create an account.

(There is no requirement to create an account to use any part of SWISS-MODEL, however you will gain the benefit of seeing a list of your previous modelling project

Modelling Projects in Session

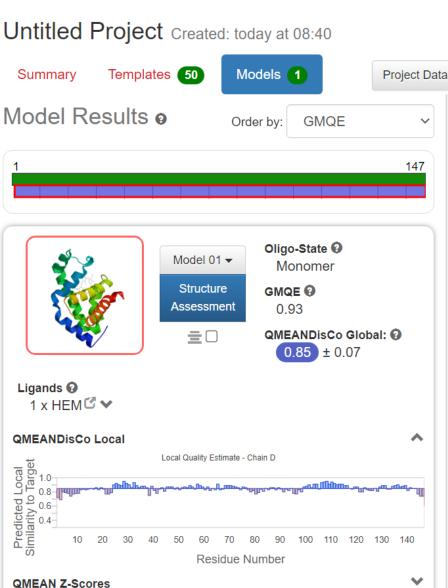


(2022)

Modelling

Repository Tools

Documer



Template

