



GPS Manual

Group-based Prediction System

Version 5.0

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The software is only free for academic research.

The latest version of GPS software is available from <http://gps.biocuckoo.cn/download.php>.

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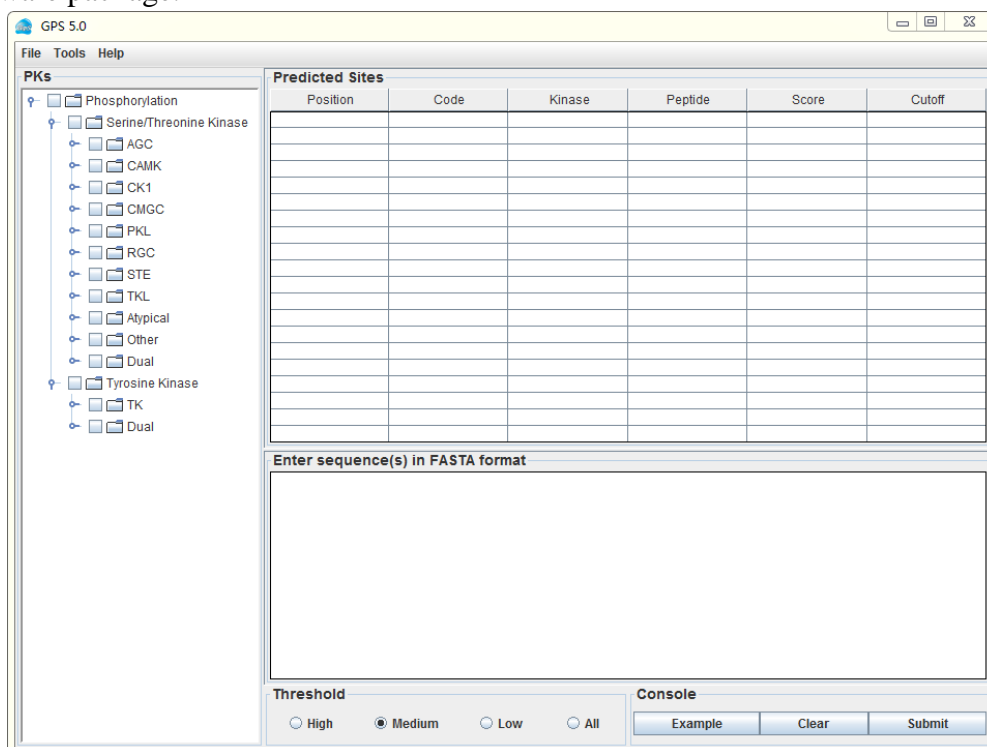
Statement

1. **Implementation.** The softwares of the CUCKOO Workgroup are implemented in JAVA (J2SE). Usually, both of online service and local stand-alone packages will be provided.
2. **Availability.** Our softwares are freely available for academic researches. For non-profit users, you can copy, distribute and use the softwares for your scientific studies. Our softwares are not free for commercial usage.
3. **GPS.** Previously, we used the GPS to denote our Group-based Phosphorylation Scoring algorithm. Currently, we are developing an integrated computational platform for post-translational modifications (PTMs) of proteins. We re-denote the GPS as Group-based Prediction Systems. This software is an indispensable part of GPS.
4. **Usage.** Our softwares are designed in an easy-to-use manner. Also, we invite you to read the manual before using the softwares.
5. **Updation.** Our softwares will be updated routinely based on users' suggestions and advices. Thus, your feedback is greatly important for our future updation. Please do not hesitate to contact with us if you have any concerns.
6. **Citation.** Usually, the latest published articles will be shown on the software websites. We wish you could cite the article if the software has been helpful for your work.
7. **Acknowledgements.** Funding for open access charge: Special Project on Precision Medicine under the National Key R&D Program [2017YFC0906600, 2018YFC0910500]; Natural Science Foundation of China [31671360, 81701567, 31801095]; National Program for Support of Top-Notch Young Professionals; Changjiang Scholars Program of China; The program for HUST Academic Frontier Youth Team; China Postdoctoral Science Foundation [2018M642816, 2018M632870].

Introduction

Identification of phosphorylation sites with their cognate protein kinases (PKs) is the foundation for understanding the functional dynamics and plasticity of various cellular processes. Although nearly 10 kinase-specific predictors were developed, numerous PKs were casually classified into sub-groups without a standard rule. And for large-scale predictions, the false positive rate (FPR) was also never addressed. Here we adopted a well-established rule to classify PKs with their verified sites into a hierarchical structure with four levels, including group, family, subfamily and single PK¹. Then we constructed the **GPS (Group-based Prediction System, ver 2.0)** software, with a modified version of GPS (Group-based Phosphorylation Scoring) algorithm^{2,3}. As the first stand-alone software for computational phosphorylation, GPS 2.0 was implemented in JAVA and could predict kinase-specific phosphorylation sites for **408** human PKs in hierarchy.

Currently, as more and more phosphorylation sites have been experimentally validated, a more accurate and reliable kinases-specific phosphorylation sites prediction tool is in urgent need. At the same time, several advanced GPS algorithm have been developed and achieved higher performance, including GPS 2.1 and GPS 2.2. To improve the performance of phosphorylation sites prediction, we applied the GPS 2.2 algorithm and enlarged the training set. In addition, we included the phosphorylation sites prediction in 161 species. For dual-specific kinases, we also provided the prediction for these kinases. Finally, we constructed the **GPS (Group-based Prediction System, ver 5.0)** software package.

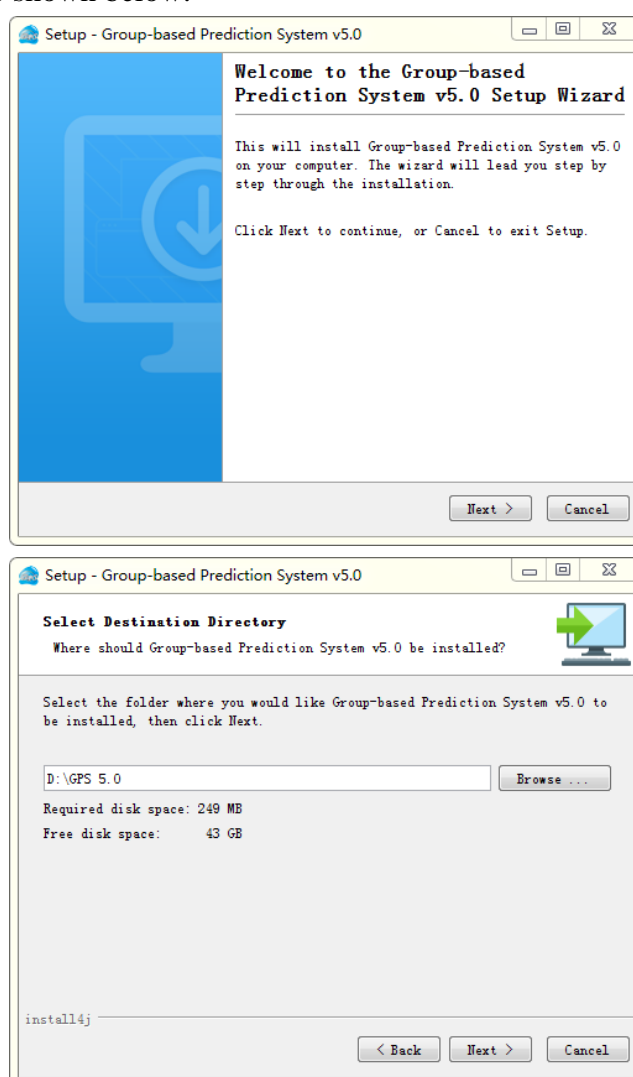


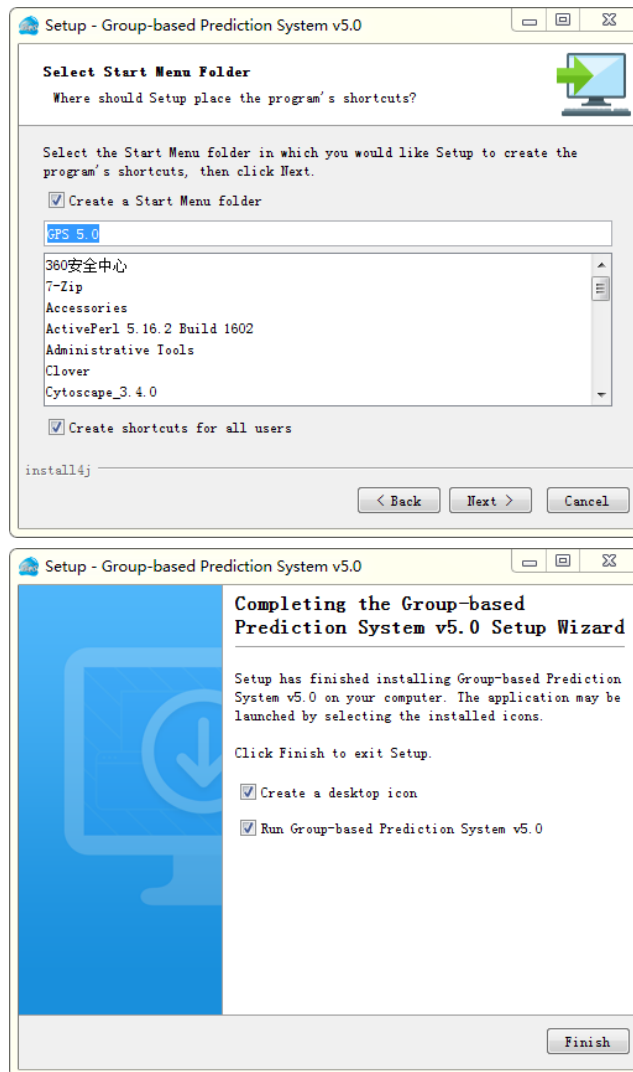
Group-based Prediction System v5.0 User Interface

Download & Installation

The software of GPS 5.0 was implemented in JAVA, and could be installed on Windows systems. GPS 5.0 distributions for Windows can be found at <http://gps.biocuckoo.cn/download.php>. We recommend that users could download the latest release.

After downloading, please double-click on the file *GPS_5.0_windows.exe* to begin installation. Follow the user prompts through the installation. And snapshots of the setup program are shown below:



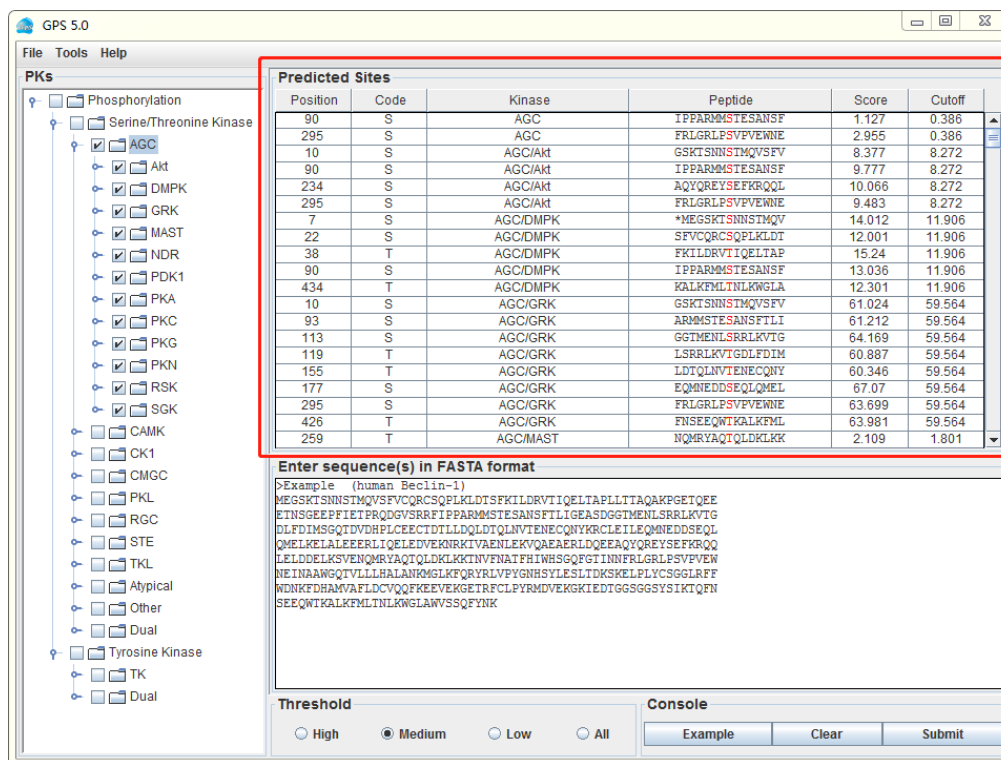


Click on the **Finish** button to complete the setup program.

For convenience, the GPS 5.0 allows users to input their protein sequences into the “TEXT form” for prediction. One or more protein sequences should be prepared in **FATSA** format as below:

Please note: All irregular words, including non-amino acid word (eg, number) and blank, will be removed automatically. As an instance, we put **human Beclin-1** protein sequence as an example for GPS 5.0. Users could click on the “Example” button to access the example.





Batch Prediction

We also provide an alternative approach for processing multiple protein sequences. If the file is large, the **Batch Predictor** will be convenient for users.

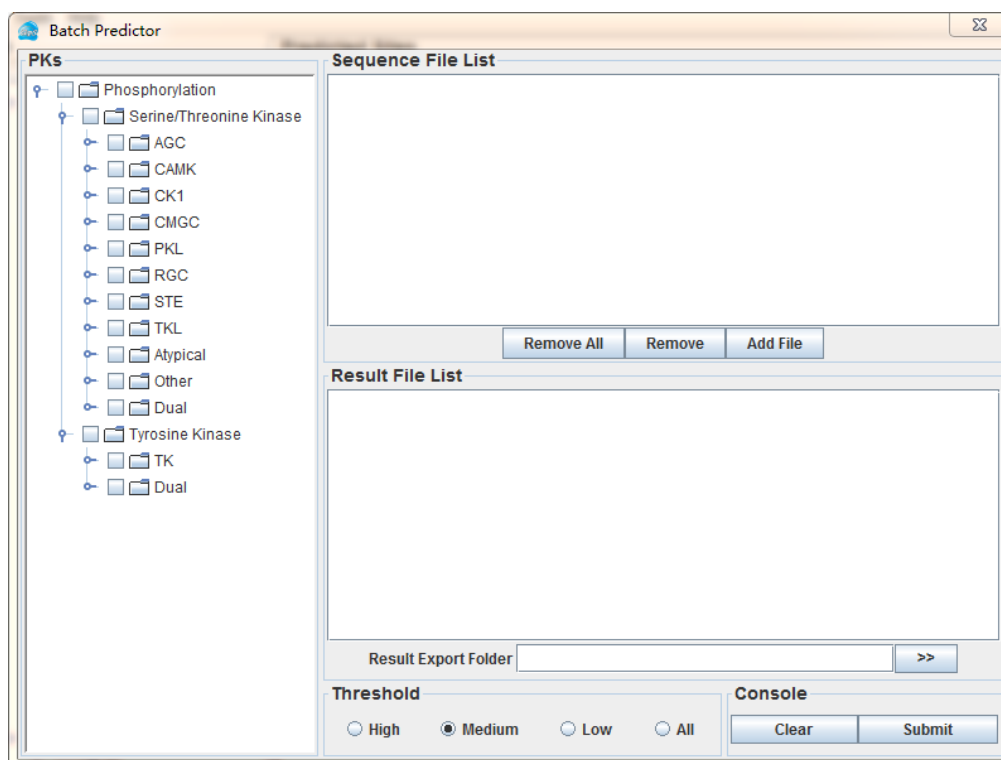
The following steps show you how to use it:

Put protein sequences into a file with **FATSA** format as below:

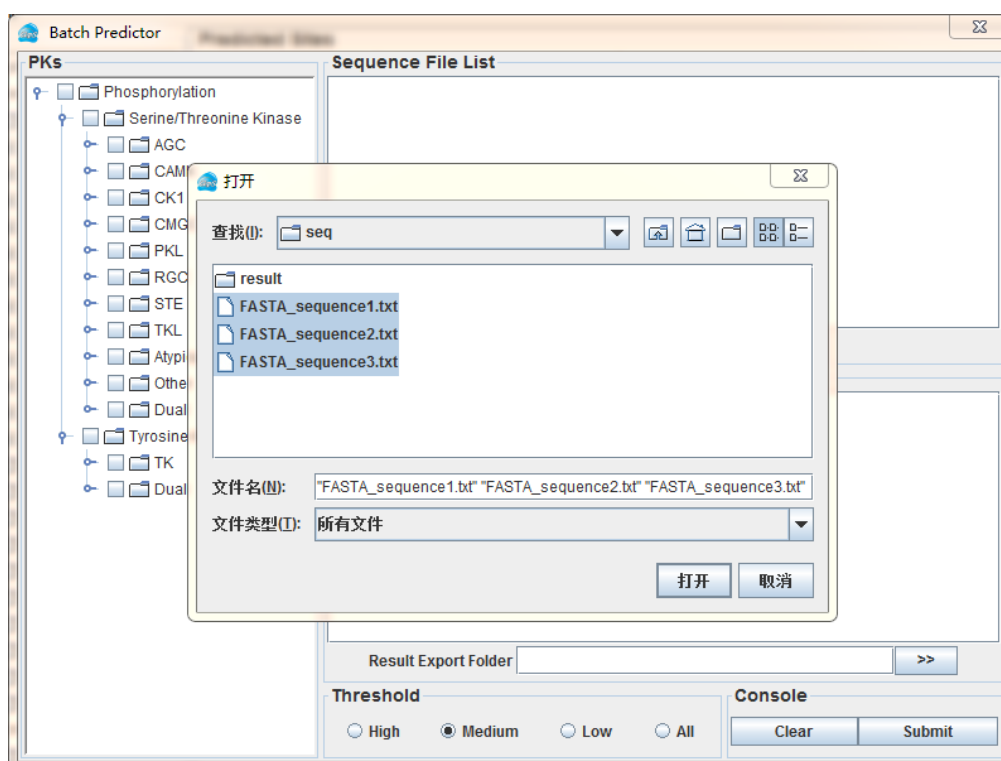
```
>protein1
XXXXXXXXXXXXXXXXXX
XXXXXXXXXX
>protein2
XXXXXXXXXXXXXXXXXXXX...
>protein3
XXXXXXXXXXXXXXXXXX
```

The names of proteins are necessary (the line with ">" and a protein name/accession number).

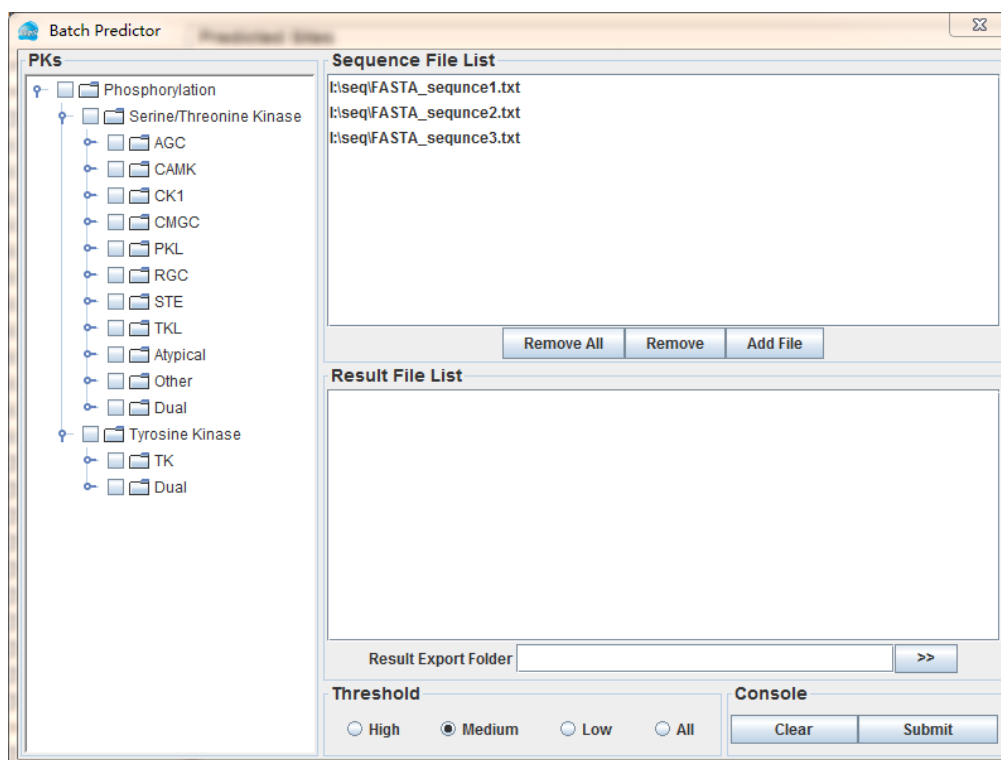
To run the Batch Predictor just select the **Batch Predictor** option in the **Tools** menu.



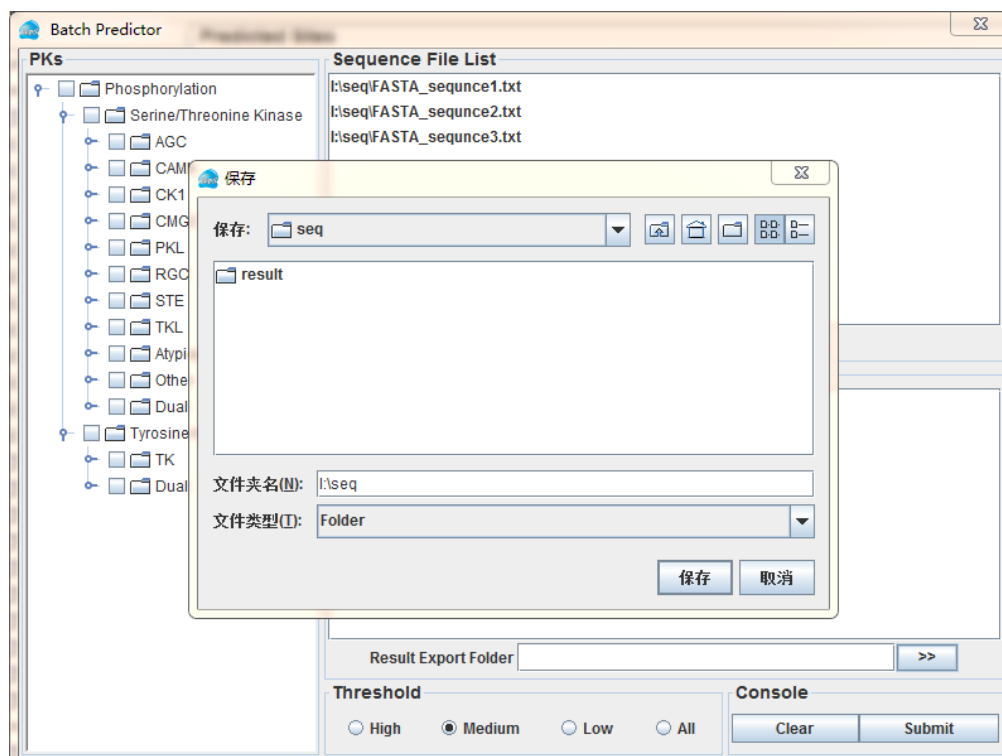
Click on the **Add File** button and add one or more protein sequence files in your hard disk.



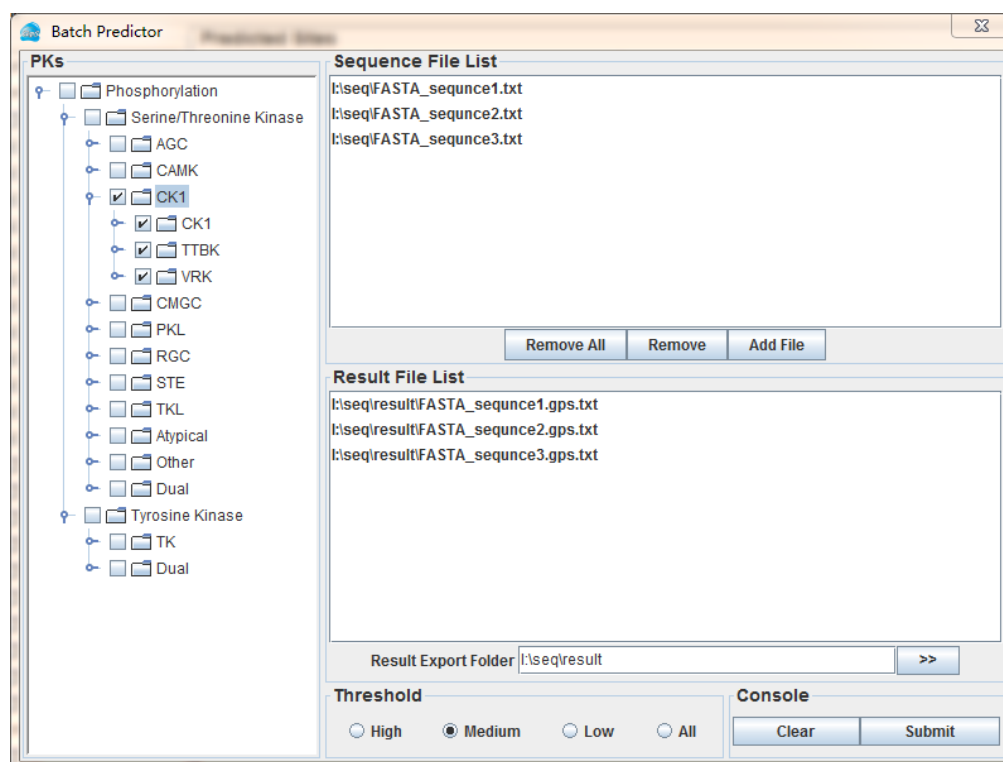
The name of added files will be shown in the **Sequence File List**



The output directory of prediction results should also be defined. Please click on the “>>” button to specify the export file fold.



5. Choose one or more kinases from **Kinase Hierarchy Tree**, and then pick a **Threshold**, click on the **Submit** button, then the **Batch Predictor** will begin to process all of the sequence files that have been added to the list. The results of predictions will be exported to the **Prediction Export Fold**, and the name of result files will be shown in the **Prediction**

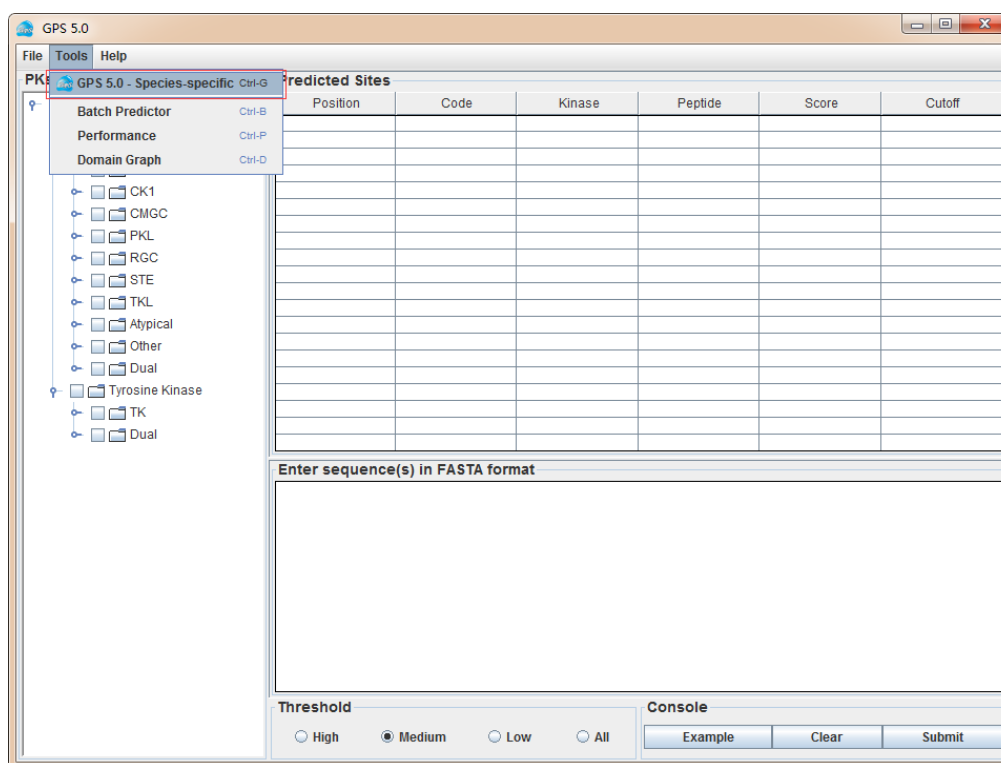
File List.

Prediction of Species-specific Kinase-specific Phosphorylation Sites

Direct Prediction

An advance feature of GPS 5.0 is Prediction of Species-specific Kinase-specific Phosphorylation Sites. Users can predict phosphorylation sites in different species. Since there isn't a standard principle classification for all animals and plants, we applied the kinases classification for eukaryotes from a new published database - iEKPd, which is a eukaryotic protein kinases and protein phosphatases database. Kinases and phosphatases in 164 species are classified in a hierarchy structure, including group, family and single protein.

To predict phosphorylation sites in 161 species, users could open the species-specific prediction interface by clicking the item “GPS 5.0 – Species-specific” item in “Tools” menu.

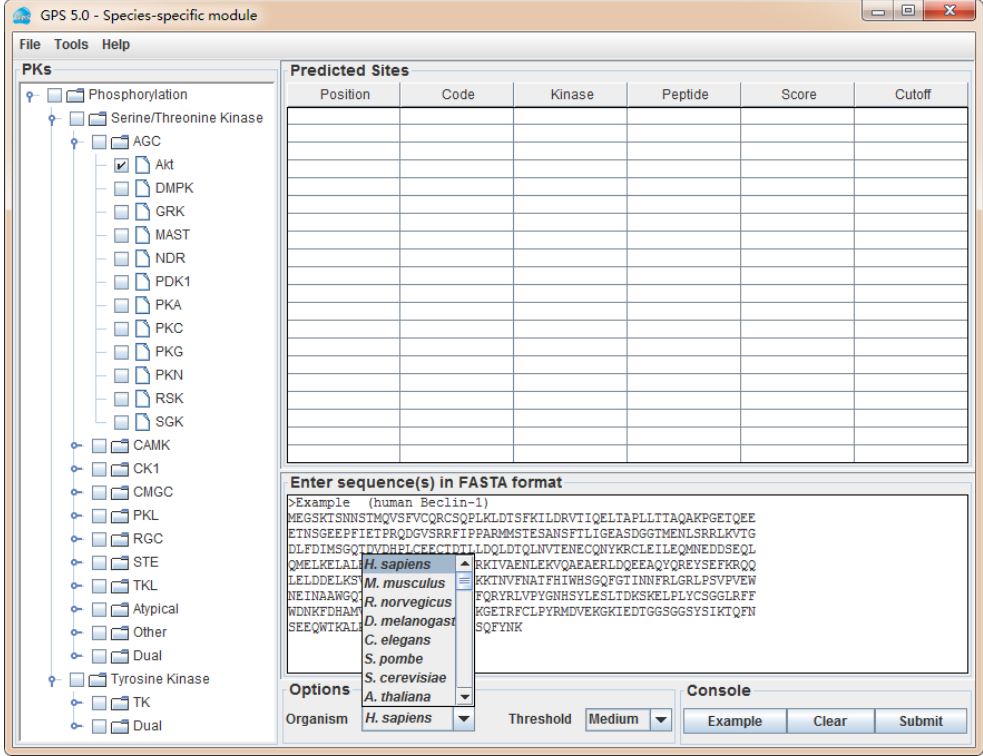


Similar with the classical version, this prediction interface also allows users to input their protein sequences into the “TEXT form” for prediction. One or more protein sequences should be prepared in **FATSA** format as below:

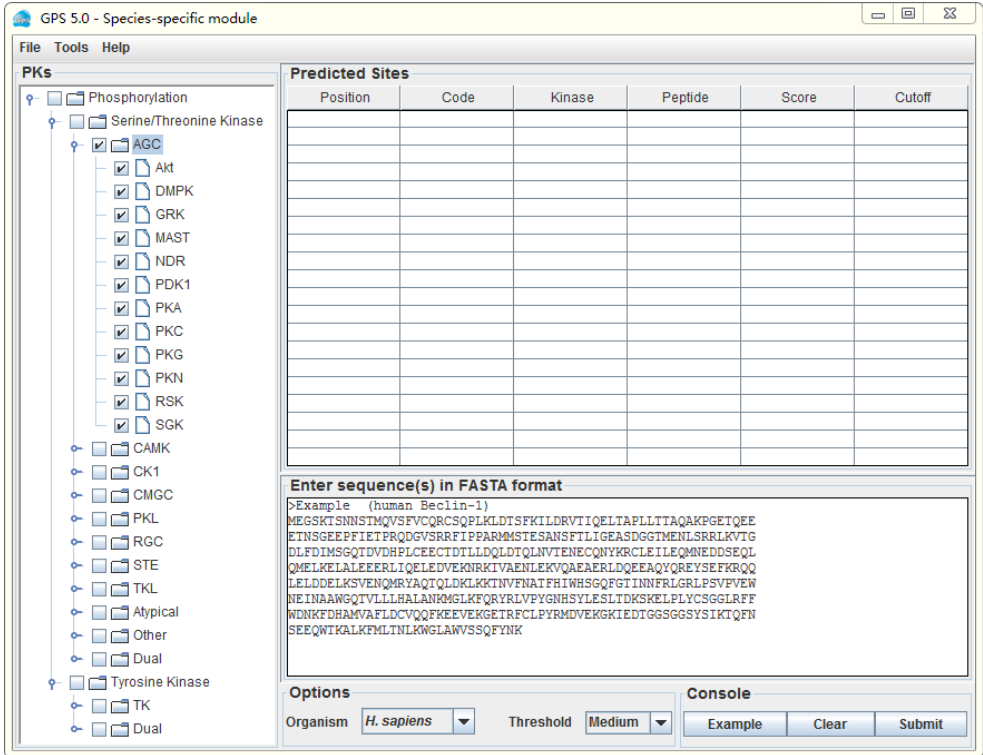
>protein1

[illegible]

13



Choose one or more kinases from the **Kinase Hierarchy Tree**



Choose a **Threshold** what you need, default is **Medium**

[illegible]

Click on the **Submit** button, then the predicted phosphorylation sites will be shown.

GPS 5.0 - Species-specific module

File

Tools

Help

PKs

Phosphorylation

Serine/Threonine Kinase

AGC

Akt

DMPK

GRK

MAST

NDR

PKD1

PKA

PKC

PKG

PKN

RSK

SGK

CAMK

CK1

CMGC

PKL

RGC

STE

TKL

Atypical

Other

Dual

Tyrosine Kinase

TK

Dual

Predicted Sites

Position	Code	Kinase	Peptide	Score	Cutoff
90	S	AGC	IPPARMSTESANSF	1.127	0.386
295	S	AGC	FRLGRLSPVPEVNE	2.955	0.386
10	S	AGC/AKT	GSKTSNNSTMVQSFV	8.377	8.272
90	S	AGC/AKT	IPPARMSTESANSF	9.777	8.272
234	S	AGC/AKT	AQYQREYSEFKRQQL	10.066	8.272
295	S	AGC/AKT	FRLGRLSPVPEVNE	9.483	8.272
7	S	AGC/DMPK	*MEGSKTSNNSTMVQ	14.012	11.906
22	S	AGC/DMPK	SFVCQRCSPQLKLDLT	12.001	11.906
38	T	AGC/DMPK	FKTLDRITQELTAP	15.24	11.906
90	S	AGC/DMPK	IPPARMSTESANSF	13.036	11.906
434	T	AGC/DMPK	KALKFMLTNLKWGLA	12.301	11.906
10	S	AGC/GRK	GSKTSNNSTMVQSFV	61.024	59.564
93	S	AGC/GRK	ARMSTESANSFTLI	61.212	59.564
113	S	AGC/GRK	GGTMENTLSRLKVTG	64.169	59.564
119	T	AGC/GRK	LSRRLKVTGDLFDIM	60.887	59.564
155	T	AGC/GRK	LDITQLNVTENECQNY	60.346	59.564
177	S	AGC/GRK	EQMNEDDSEQLQML	67.07	59.564
295	S	AGC/GRK	FRLGRLSPVPEVNE	63.699	59.564
426	T	AGC/GRK	FNSEEQTKALKFML	63.981	59.564
259	T	AGC/MAST	NQMRYAQTLQDKLKK	2.109	1.801

Enter sequence(s) in FASTA format

>Example (human Beclin-1)

MEGSKTSNNSTMVQSFVCQRCSPQLKLDLTSFKILDRVTIQLTAPLLTTAQAKPGTQEE

ETNSGEEPFIEIPRQDGVSRRIIPAPARMSTESANSFTLIGESADGGTGMENTLSRLKVTG

DLFDIMSGQITVDVHPLCEECTDILLDLQITQLNVTENECQNYKRCLEILEQMNEDDSEQL

QMLKELALEEERLIELEDVEKKNKIVAENLEKVAEAEERLDQEEAQYQREYSEFKRQQL

LELDDKLSVENQMRYAQTLQDKLKTINVTNATFHIWHSQGFGINNFRGLRSPVPEV

NEINAAWGQITVLLHALANKMGLKFRQYRVLVPGNHSYLESLTDKSKLPLCYSGGLRFF

WDNKFDHAMVAFDLVCQFKEEVEKGETRFLPYRMDVEKGIETDGSGGSSYSIKTQFN

SEEQWTKALKFMLTNLKWGLAWVSQFYNK

Options

Organism

H. sapiens

Threshold

Medium

Console

Example

Clear

Submit

Batch Prediction

In GPS 5.0 – Species-specific prediction interface, we also provide an alternative approach for processing multiple protein sequences. If the file is large, the **Batch Predictor** will be convenient for users. The usage is similar to the **Batch Prediction** of **Direct Prediction**.

Algorithms and Prediction Performance

Algorithm Design

To predict kinase-specific phosphorylation sites, a series of GPS (Group-based Phosphorylation Scoring) algorithm were maintained and improved in the past few years^{2,3}. The basic hypothesis of GPS algorithm is that highly similar short peptides bear similar biochemical properties for the modification. Therefore, we defined a *phosphorylation site peptide* PSP(m, n) as a pS, pT or pY amino acid flanked by m residues upstream and n residues downstream. Then we used the amino acid substitution matrix BLOSUM62 to calculate the similarity between two PSP (m, n) peptides.

In GPS 5.0, a new hypothesis is that long flanking regions around p-sites might be important for the recognition of PKs, which are bulky molecules to interact with phosphorylatable residues. Thus the GPS 5.0 is upgraded in basis of GPS 2.1 and contained two parts: the scoring strategy and performance improvement.

The scoring strategy defined the average similarity score (S) between a PSP(30,30) peptide P and peptides around all known p-sites in the training data set:

$$S = \frac{1}{N} \sum_{j=-30}^{L-31} \left(\sum_{i=1}^N M_{train}[P_j, T_{ij}] \right) \times W_j$$

Where L is the length of the PSP(30, 30) peptide and equal to 61 to represent a relatively long flanking region. N is the number of known p-sites in the positive data set. T_{ij} is the amino acid at position j around a known p-site T_i ($i = 1, 2, 3, \dots, N$). W_j is the weight value of position j , and M_{train} represents the optimized amino acid substitution matrix.

The performance improvement comprises two parts, the MLS and MaM in GPS 2.1 were upgraded into PWD and SMO, respectively.

(i) PWD: The amino acid substitution matrix BLOSUM62 ($M_{BLOSUM62}$) was used to calculate an average similarity score at the position j of a PSP(30, 30) peptide P as S'_j :

$$S'_j = W_j \frac{1}{N} \sum_{i=1}^N M_{BLOSUM62}[P_j, T_{ij}]$$

Initially, the weight value of each position W_j in the PSP(30, 30) peptide was set to 1. Then the one-vs-rest (OVR) classifier with the ridge (L2) penalty of the LR algorithm was used to optimize W_j values, by applying the "newton-cg" solver in the class

LogisticRegressionCV of scikit-learn v0.21.0 (<https://scikit-learn.org/>). To avoid over-fitting, such a procedure was repeated for 100 times and 10-fold cross-validation was conducted to determine the inverse of regularization strength of each time. Receiver operating characteristic (ROC) curves were illustrated, and area under curve (AUC) values were calculated. The optimal W_j vectors were determined based on the highest AUC value:

$$W_j = W_{-30}, \dots, W_{-1}, W_0, W_1, \dots, W_{30}$$

In order to evaluate position-specific contributions of flanking regions around p-sites for different PK clusters, the W_j vectors were normalized into -1 to 1 based on the maximum absolute value.

(ii) SMO: The average similarity score of an amino acid a in the given PSP(30, 30) peptide P and a residue b in peptides around all known p-sites was defined as S_{ab} :

$$S_{ab} = \frac{1}{N} \sum_{j=-30}^{L-31} C_j \times M_{BLOSUM62}[a, b] \times W_j$$

Where C_j is the number of ab amino acid pairs at position j . In BLOSUM62, there were 24 types of characters including 20 types of amino acids and 4 non-canonical characters (B, Z, X and *). Thus, a number of $[24 \times (24+1)]/2 = 300$ unique S_{ab} scores ($S_{ab} = S_{ba}$) were generated. Then, the same LR algorithm was used to optimize all of S_{ab} scores to produce a new matrix M_{train} :

$$M_{train} = (S_{AA}, S_{AC}, S_{AD}, \dots, S_{**})_{300}$$

Evaluation of Prediction Performances

Performance measurements

To evaluate the prediction performances, four standard measurements were used, including accuracy (Ac), sensitivity (Sn), specificity (Sp) and Mathew correlation coefficient (MCC). Accuracy (Ac) represents the correct ratio between both positive (+) and negative (-) data sets, while sensitivity (Sn) and specificity (Sp) illustrate the correct prediction ratios of positive (+) and negative data (-) sets respectively. Since the number of positive data and negative data differed too much from each other, the Mathew correlation coefficient (MCC) was also included. The value of MCC ranges from -1 to 1, and a larger MCC value stands for better prediction performance.

Among the data with positive hits by GPS 5.0, the real positives were defined as *true positives* (TP), while the others were defined as *false positives* (FP). Among the data with negative predictions, the real positives were defined as *false negatives* (FN), while the others were defined as *true negatives* (TN). The four measurements of sensitivity (Sn), specificity (Sp), accuracy (Ac), and Mathew correlation coefficient (MCC) were defined as below:

$$Sn = \frac{TP}{TP + FN}, \quad Sp = \frac{TN}{TN + FP},$$

$$Ac = \frac{TP + TN}{TP + FP + TN + FN}, \text{ and}$$

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

The self-consistency, leave-one-out validation and n-fold cross-validation

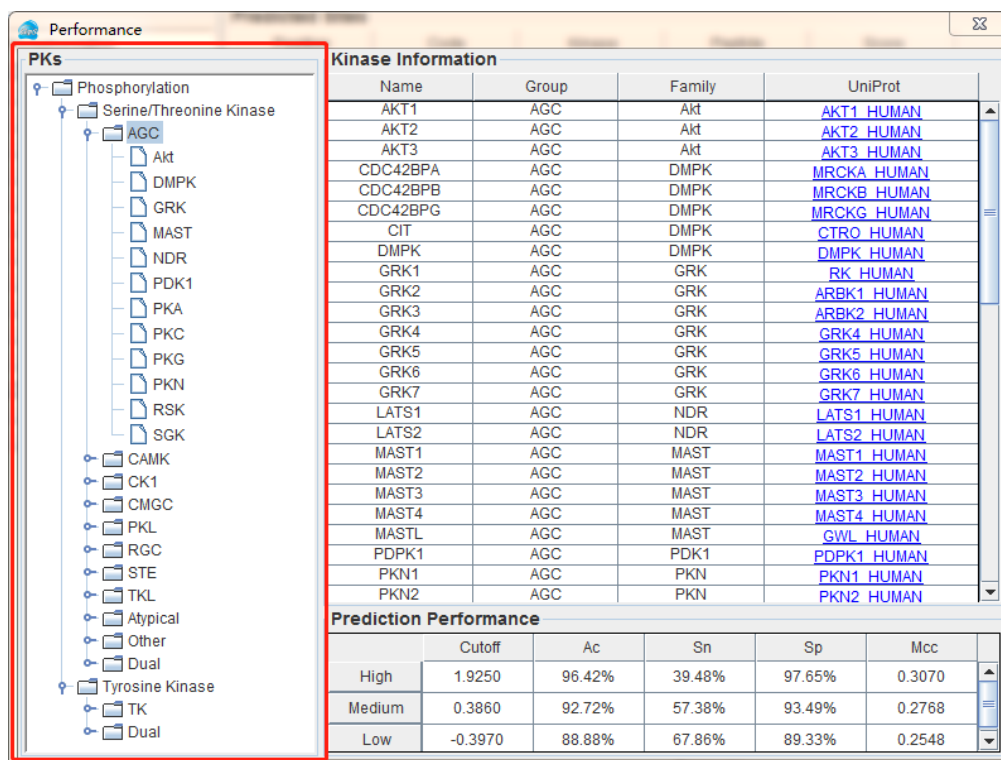
The self-consistency used the training positive data and negative data directly to evaluate the prediction performance, and represented the computational power of the prediction system. However, the robustness and stability of the software should also be evaluated by leave-one-out validation and n -fold cross-validation. In the leave-one-out validation, which is also called as Jack-Knife validation, each sites in the dataset was picked out in turn as an independent test sample, and all the remaining sites were regarded as training data. This process was repeated until each site was used as test data one time. In n -fold cross-validation all the (+) sites and (-) sites were combined and then divided equally into n parts, keeping the same distribution of (+) and (-) sites in each part. Then $n-1$ parts were merged into a training data set while the remanent part was taken as a testing data set. This process was repeated twenty times and the average performance of n -fold cross-validation was used to estimate the performance. In this work, the performances of validation were calculated for all PK groups. And the 10-fold cross-validations were performed for PK groups with the number of positive data ≥ 30 and leave-one-out validation were calculated

for PK groups with the number of positive data < 30 . All validation performances of GPS 5.0 have been included in **Performance** option in the **Tools** menu.

Accessing of prediction performances of GPS 5.0 with related information

To check the prediction performance of the **479** human PKs just select the **Performance** option in the **Tools** menu.

Choose a kinase what you want check from the **Kinase Hierarchy Tree**.



PKs

- Phosphorylation
 - Serine/Threonine Kinase
 - AGC**
 - Akt
 - DMPK
 - GRK
 - MAST
 - NDR
 - PDK1
 - PKA
 - PKC
 - PKG
 - PKN
 - RSK
 - SGK
 - CAMK
 - CK1
 - CMGC
 - PKL
 - RGC
 - STE
 - TKL
 - Atypical
 - Other
 - Dual
 - Tyrosine Kinase
 - TK
 - Dual

Kinase Information

Name	Group	Family	UniProt
AKT1	AGC	Akt	AKT1_HUMAN
AKT2	AGC	Akt	AKT2_HUMAN
AKT3	AGC	Akt	AKT3_HUMAN
CDC42BPA	AGC	DMPK	MRCKA_HUMAN
CDC42BPB	AGC	DMPK	MRCKB_HUMAN
CDC42BPG	AGC	DMPK	MRCKG_HUMAN
CIT	AGC	DMPK	CTRO_HUMAN
DMPK	AGC	DMPK	DMPK_HUMAN
GRK1	AGC	GRK	RK_HUMAN
GRK2	AGC	GRK	ARBK1_HUMAN
GRK3	AGC	GRK	ARBK2_HUMAN
GRK4	AGC	GRK	GRK4_HUMAN
GRK5	AGC	GRK	GRK5_HUMAN
GRK6	AGC	GRK	GRK6_HUMAN
GRK7	AGC	GRK	GRK7_HUMAN
LATS1	AGC	NDR	LATS1_HUMAN
LATS2	AGC	NDR	LATS2_HUMAN
MAST1	AGC	MAST	MAST1_HUMAN
MAST2	AGC	MAST	MAST2_HUMAN
MAST3	AGC	MAST	MAST3_HUMAN
MAST4	AGC	MAST	MAST4_HUMAN
MASTL	AGC	MAST	GWL_HUMAN
PDPK1	AGC	PDK1	PDPK1_HUMAN
PKN1	AGC	PKN	PKN1_HUMAN
PKN2	AGC	PKN	PKN2_HUMAN

Prediction Performance

	Cutoff	Ac	Sn	Sp	Mcc
High	1.9250	96.42%	39.48%	97.65%	0.3070
Medium	0.3860	92.72%	57.38%	93.49%	0.2768
Low	-0.3970	88.88%	67.86%	89.33%	0.2548

Then **Kinase Information** and **Prediction Performance** of the kinase are shown in the tables.

Performance

PKs

- Phosphorylation
 - Serine/Threonine Kinase
 - AGC
 - Akt
 - DMPK
 - GRK
 - MAST
 - NDR
 - PDK1
 - PKA
 - PKC
 - PKG
 - PKN
 - RSK
 - SGK
 - CAMK
 - CK1
 - CMGC
 - PKL
 - RGC
 - STE
 - TKL
 - Atypical
 - Other
 - Dual
 - Tyrosine Kinase
 - TK
 - Dual

Kinase Information

Name	Group	Family	UniProt
AKT1	AGC	Akt	AKT1 HUMAN
AKT2	AGC	Akt	AKT2 HUMAN
AKT3	AGC	Akt	AKT3 HUMAN
CDC42BPA	AGC	DMPK	MRCKA HUMAN
CDC42BPB	AGC	DMPK	MRCKB HUMAN
CDC42BPG	AGC	DMPK	MRCKG HUMAN
CIT	AGC	DMPK	CTRO HUMAN
DMPK	AGC	DMPK	DMPK HUMAN
GRK1	AGC	GRK	RK HUMAN
GRK2	AGC	GRK	ARBK1 HUMAN
GRK3	AGC	GRK	ARBK2 HUMAN
GRK4	AGC	GRK	GRK4 HUMAN
GRK5	AGC	GRK	GRK5 HUMAN
GRK6	AGC	GRK	GRK6 HUMAN
GRK7	AGC	GRK	GRK7 HUMAN
LATS1	AGC	NDR	LATS1 HUMAN
LATS2	AGC	NDR	LATS2 HUMAN
MAST1	AGC	MAST	MAST1 HUMAN
MAST2	AGC	MAST	MAST2 HUMAN
MAST3	AGC	MAST	MAST3 HUMAN
MAST4	AGC	MAST	MAST4 HUMAN
MASTL	AGC	MAST	GWL HUMAN
PDPK1	AGC	PDK1	PDPK1 HUMAN
PKN1	AGC	PKN	PKN1 HUMAN
PKN2	AGC	PKN	PKN2 HUMAN

Prediction Performance

	Cutoff	Ac	Sn	Sp	Mcc
High	1.9250	96.42%	39.48%	97.65%	0.3070
Medium	0.3860	92.72%	57.38%	93.49%	0.2768
Low	-0.3970	88.88%	67.86%	89.33%	0.2548

If you want get more kinase information, you can click on the hyperlinks in the table.

Performance

PKs

- Phosphorylation
 - Serine/Threonine Kinase
 - AGC
 - Akt
 - DMPK
 - GRK
 - MAST
 - NDR
 - PDK1
 - PKA
 - PKC
 - PKG
 - PKN
 - RSK
 - SGK
 - CAMK
 - CK1
 - CMGC
 - PKL
 - RGC
 - STE
 - TKL
 - Atypical
 - Other
 - Dual
 - Tyrosine Kinase
 - TK
 - Dual


Kinase Information

Name	Group	Family	UniProt
AKT1	AGC	Akt	AKT1 HUMAN
AKT2	AGC	Akt	AKT2 HUMAN
AKT3	AGC	Akt	AKT3 HUMAN
CDC42BPA	AGC	DMPK	MRCKA HUMAN
CDC42BPB	AGC	DMPK	MRCKB HUMAN
CDC42BPG	AGC	DMPK	MRCKG HUMAN
CIT	AGC	DMPK	CTRO HUMAN
DMPK	AGC	DMPK	DMPK HUMAN
GRK1	AGC	GRK	RK HUMAN
GRK2	AGC	GRK	ARBK1 HUMAN
GRK3	AGC	GRK	ARBK2 HUMAN
GRK4	AGC	GRK	GRK4 HUMAN
GRK5	AGC	GRK	GRK5 HUMAN
GRK6	AGC	GRK	GRK6 HUMAN
GRK7	AGC	GRK	GRK7 HUMAN
LATS1	AGC	NDR	LATS1 HUMAN
LATS2	AGC	NDR	LATS2 HUMAN
MAST1	AGC	MAST	MAST1 HUMAN
MAST2	AGC	MAST	MAST2 HUMAN
MAST3	AGC	MAST	MAST3 HUMAN
MAST4	AGC	MAST	MAST4 HUMAN
MASTL	AGC	MAST	GWL HUMAN
PDPK1	AGC	PDK1	PDPK1 HUMAN
PKN1	AGC	PKN	PKN1 HUMAN
PKN2	AGC	PKN	PKN2 HUMAN

Prediction Performance

	Cutoff	Ac	Sn	Sp	Mcc
High	1.9250	96.42%	39.48%	97.65%	0.3070
Medium	0.3860	92.72%	57.38%	93.49%	0.2768
Low	-0.3970	88.88%	67.86%	89.33%	0.2548

The hyperlinks will access the **UniProt** database and show you the detailed information of the kinases.



UniProtKB

Advanced Search

BLAST Align Retrieve/ID mapping Peptide search Help Contact

UniProtKB - P31749 (AKT1_HUMAN)

Basket

Display

BLAST Align Format Add to basket History

Other tutorials and videos Help video Feedback

Entry

Publications

Feature viewer

Feature table

Protein

Gene

Organism

Status

None

Function

Names & Taxonomy

Subcellular location

RAC-alpha serine/threonine-protein kinase

AKT1

Homo sapiens (Human)

Reviewed - Annotation score: ●●●●● - Experimental evidence at protein level²

Functionⁱ

AKT1 is one of 3 closely related serine/threonine-protein kinases (AKT1, AKT2 and AKT3) called the AKT kinase, and which regulate many processes including metabolism, proliferation, cell survival, growth and angiogenesis. This is mediated through serine and/or threonine phosphorylation of a range of downstream substrates. Over 100 substrate candidates have been reported

References

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Release Note

1. Jan. 1st, 2008, the online service and the local stand-alone packages of GPS 2.0 were released. The stand-alone software of GPS 2.0 could support Windows Operating Systems.
2. Jan. 29th, 2008, a bug was found that the version 2.0 couldn't be used under non-English Operating Systems. We fixed the bug and released the version 2.0.1 beta version. We thank Dr. Miguel Angel Sanchez (Malaga, Spain) and Dr. Gilles Vachon (Universite J. Fourier, France) to send us feedbacks.
3. Apr. 13th, 2008, The GPS 2.0.1 was released, with online service and local packages, to support three major Operating Systems, including Windows, Linux/Unix and Mac. Also, the GPS 2.0.1 manual was updated and included in the packages.
4. Mar. 1st, 2009, The GPS 2.1 was released, with online service and local packages and support three major Operating Systems including Windows, Linux/Unix and Mac. In this version, the newly developed motif length selection (MLS) method was introduced and the robustness of the prediction system was greatly improved.
5. Jul. 21st, 2009, The GPS 2.1.1 was released, with online service and local packages and support three major Operating Systems including Windows, Linux/Unix and Mac.
6. Sep. 13th, 2012, The GPS 2.1.2 was released, with online service and local packages and support three major Operating Systems including Windows, Linux/Unix and Mac.
7. Dec. 14th, 2014, The GPS 3.0 was released, with online service and local packages and support three major Operating Systems including Windows, Linux/Unix and Mac. In this version, *k-means* clustering, peptide selection (PS), and weight training (WT) procedures were added to enhance the prediction performance.
8. Jul. 20th, 2019, The GPS 5.0 was released, with online service and local packages and support three major Operating Systems including Windows, Linux/Unix and Mac. In this version, two novel methods of position weight determination (PWD) and scoring matrix optimization (SMO) were developed to improve the performance for kinase-specific phosphorylation sites prediction. Meanwhile, the GPS 5.0 manual was updated and included in the packages.