

## Special Article - Animal Husbandry

# The pLoc\_bal-mGneg Predictor is a Powerful Web-Server for Identifying the Subcellular Localization of Gram-Negative Bacterial Proteins based on their Sequences Information Alone

Kuo-Chen Chou<sup>1,2\*</sup><sup>1</sup>Gordon Life Science Institute, USA<sup>2</sup>Center for Informational Biology, University of Electronic Science and Technology of China, China

\*Corresponding author: Kuo-Chen Chou, Gordon Life Science Institute, Center for Informational Biology, University of Electronic Science and Technology of China, Chengdu, 610054, China

Received: November 29, 2019; Accepted: December 10

2019; Published: December 17, 2019

## Short Communication

Recently a very powerful web-server has been developed for predicting the subcellular localization of Gram-negative bacterial proteins purely according to their sequences information for the multi-label systems [1], in which a same protein may appear or move between two or more location sites and hence its ID (Identification) needs two or more labels for distinction, namely the “multi-label mark” [2].

The web-server is called as “pLoc\_bal-mGneg”, where “bal” means that the predictor has been treated by balancing or quasi-balancing out the training dataset [3-9], and “m” means that the predictor is with the capacity to study the multi-label systems. How the web-server is working can be clearly seen via the showcase below.

Clicking the link at [http://www.jci-bioinfo.cn/pLoc\\_bal-mGneg/](http://www.jci-bioinfo.cn/pLoc_bal-mGneg/), you will see the top page of the pLoc\_bal-mGneg web-server prompted on your computer’s screen Figure 1. Then, just simply following the commands given in the Step 2 and Step 3 of [4], you will see Figure 2

**pLoc\_bal-mGneg: predict subcellular localization of Gram-negative proteins by balancing training dataset and general PseAAC**

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Enter query sequences

Enter the sequences of query proteins in FASTA format ([Example](#)): the number of proteins is limited at 10 or less for each submission.

**Submit** **Cancel**

Or, upload a file for batch prediction

Enter your e-mail address and upload the batch input file ([Batch-example](#)). The predicted result will be sent to you by e-mail once completed; it usually takes 1 minute or so for each protein sequence

Upload file:  **Browse**  
Your Email:   
**Batch submit** **Cancel**

Figure 1: A semi screenshot for the top page of pLoc\_bal-mGneg (Adapted from [4] with permission).

Covered by pLoc\_bal-mGneg are the following 8 subcellular locations

(1) Cell inner membrane	(2) Cell outer membrane
(3) Cytoplasm	(4) Extracellular
(5) Fimbrium	(6) Flagellum
(7) Nucleoid	(8) Periplasm

Predicted results

Protein ID	Subcellular location or locations
>P22340	2
>P04032	8
>P04825	1, 3
>P22251	4, 6

[Continue Test](#)

Figure 2: A semi screenshot for the webpage obtained by following Step 3 of Section 3.5 (Adapted from [4] with permission).

on the screen of your computer. The corresponding predicted reports were given [4]. As you can see from there: nearly all the success rates achieved by the web-server for the Gram-negative bacterial proteins in each of the 8 subcellular locations are within the range of 98-100%, which is far beyond the reach of any of its counterparts.

In addition to the advantages of high accuracy and easy to use, the web-server has been constructed by strictly observing the “Chou’s 5-steps rule” and hence bears the following fantastic merits as agreed and supported by many investigators (see, e.g., [10-91] as well as three comprehensive review papers [2,92,93]): (1) crystal clear in logic development, (2) completely transparent in operation, (3) easily to repeat the reported results by other investigators, (4) with high potential in stimulating other sequence-analyzing methods, and (5) very convenient to be used by the majority of experimental scientists.

Besides, the approach [94-96] of PseAAC (Pseudo Amino Acid Composition) has also been applied during developing the web-server predictor. It is a very powerful approach for representing the protein samples by catching their special or key features, as done by many scientists as well [97-222].

Furthermore, the IHTS (Inserting Hypothetical Training Samples) treatment has also been applied to balance or quasi-balance out the training dataset [57,60,84].

For the amazing and awesome roles of the “5-steps rule” and general “PseAAC” in driving proteome, genome analyses and drug development, see a series of recent papers [2,93,223-232] where both

the “5-steps rule” and the general “PseAAC” as well as their huge impacts have been very impressively recollected at different angles or from varieties of aspects.

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