Comprehensive Statistical Analysis on Estimated Errors of Averagine Model for Intact Proteins

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Abstract: Averagine Model (AM) is a very popular and practical computing tool in top-down proteomics, which is usually employed to predict the monoisotopic mass for an unknown protein or a peptide to be of interest. However, with the significant advancement on high-resolution and high-accuracy mass spectrometry (MS) instrumentation, AM's limitation on its accuracy became more and more significant. Here we studied statistically AM's mass errors using all proteins in the Human databases. Both the mass errors of estimated monoisotopic mass and average mass for all proteins from the Human protein database are analysed comprehensively in this paper. According to the results obtained, we then found the error range difference between these two different types of mass errors and then we further analysed the error contributions on the individual elemental level of C, H, N, O and S which constitute the proteins. Our analysis will provide an experimental basis to further improve the average model in the top-down proteomics.

1 INTRODUCTION

Modern mass spectrometry (MS) is widely utilized to identify proteins in the field of proteomics no matter the strategy being used is bottom-up or top-down. The isotopic clusters of the protein in MS data are typically the most abundant information also being the most potential element when protein needs to be identified.

Under the circumstance of the known protein identity, i.e., the protein formula needed to be studied, of which the isotopic distribution can be obtained by theoretical predictions using the simulation algorithms, such as emass. However, if the identity of one protein is unknown, in order to search it out, we will need to estimate its monoisotopic or average mass from the database by using its molecular weight (MW). To solve these types of problem, the averagine model, utilizing a kind of an average amino acid, has been developed and widely used in estimating molecular weight of proteins by using the MS data.

Although this model can be utilized to identify proteins in a relatively high accuracy for the most of times, it still has some limitations when applied to modern MS data which has level of high resolution and high accuracy. Therefore the model of averagine is generally required when an unknown protein is needed to be identified. Averagine is an average amino acid based on the occurrences of all amino acids from protein database (PIR) and it was proposed in order to estimate the average mass of the targeted protein and find their corresponding estimated formula.

Through past study, it could be discovered that although Averagine Model (AM) may estimate the true average molecular mass with an error up to 0.5 Dalton, it can still be improved further to reduce this mass errors.

The issue that will be addressed in this paper is about how to reduce the estimated errors in the realworld MS data sets when applying AM to proteins with a relatively large MW.

Firstly, the mass error distribution which covered the full mass range for all proteins from the Human database was computed. After this more emphasis was put on the average mass errors as well as the monoisotopic mass errors estimated by AM.

The reason why we chose these two types of mass errors is that these two types of protein masses are the most typical and crucial information that are required to identify a certain protein when searching the database.

136

Che. Y.

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Later in order to improve the AM model, the analysis of the individual elemental contributions is furthered which is to estimate the errors may be caused from C, H, N, O and S.

2 EXPERIMENT AND METHOD

Comprehensive simulation analysis can help us find the essential pattern hidden behind the complicated data sets in most cases. To find out the regular patterns of mass errors estimated when applying the averagine model on all human proteins, an in-house program was developed using the MATLAB toolbox which has multiple functions and bioinformatics tools that can deal with massive amount of protein data, as well as its capacity of transferring massive digital results to visualized diagrams such as scatters and bars conveniently.

The averagine model is used in the experimental fundamentals which in this case offers the basic idea of how to estimate unknown large proteins as well.

Here the molecular information for each protein in human protein database were utilized and then the estimated masses were compared with the actual theoretical masses calculated using the formula provided from the database. Both the average mass errors and the monoisotopic mass errors are obtained along with the different mass ranges.

All the statistical calculations presented here are based on Human protein database, which is a collection of 20,341 sequences of proteins (June, 2019). The primary task of our study is to get the mass error distribution covered the full mass range, which will provide the experimental foundation to improve AM by reducing its estimated errors when applied to large proteins with MW larger than 30 kDa.

2.1 Main Analysis Process

To get the estimated mass errors, four computational steps are conducted as below (figure 1):

- Step 1: Computing every formula of protein in the Human Protein Database;
- Step 2: Using the obtained formula result from the first step and the emass algorithm to compute the theoretical isotopic distributions;
- Step 3: Using the AM and the average mass provided in the second step, estimate the formula for each protein;
- Step 4: Generating two types of mass errors, i.e., average mass errors and monoisotopic mass errors.



Figure 1: Diagram of the four computing steps.

estmation mass	unit of element	isotopic masses / average mass
	averagine model	C4.9384 H7.7583 N1.3577 O1.4773 S0.0417

Figure 2: Key process of AM application.

Although average mass is widely used for large molecule mass estimation, the monoisotopic mass still represents the most accurate mass for a compound.

Here in this experiment, two sets of errors are computed through the four computing processes introduced which are monoisotopic and average element mass. (figure 2)

The reason why for taking both errors in consideration is that the former error could offer hints on how to improve AM while the latter error offers the information related to the unknown large molecules validated by the information from the database.

2.2 Simulation on the Estimated Mass Errors for All Proteins from Human Database

We statistically computed two types of mass errors between Averagine-fit and theoretical isotopic clusters. According to the distribution, we then compared the differences between the mass error ranges for both average masses and monoisotopic masses. The results showed that the mass accuracy can be improved remarkably for large proteins in terms of the monoisotopic mass errors.

However, this is not enough for high-resolution mass spectrometers, therefore, futher analysis of the elemental contribution are provided to estimate the mass errors from all individual elements which are C, H, N, O, and S.

More detailed results will be shown in next section.

3 RESULT AND CONCLUSION

As stated previously, the estimated average mass

errors and monoisotopic mass errors for all proteins in the human database are firstly analysed respectively, which their distributions covered the whole mass range were acquired.

3.1 Estimated Mass Error Distributions

Figure 3 shows the relationship between the estimated average mass error in Dalton (vertical axis) and the corresponding average mass (also in Dalton, horizontal axis) for the 20,431 proteins in Human Protein Database.

Each blue-cross designates one protein. For this distribution, we can find the range of errors is between [-0.5, 0.5] Dalton, which is due to the estimated number of Hydrogen atom when computing errors in the rounding process of AM.



Figure 3: The estimated average mass error using AM with the average mass (unit: Dalton).



Figure 4: The estimated monoisotopic mass error with the monoisotopic mass (unit: Dalton).

To our surprise, these errors' ranges are much larger than those of the corresponding average mass errors, about [-2.5, 2.5] Dalton. We speculate that this

difference is caused by the different contribution of the constituent elements: C, H, N, O and S (will be detailed in later section).

We then extracted the absolute value of these two types of errors, i.e. the average mass errors and the monoisotopic mass errors, as shown in Figures 5 and 6. Unlike the relatively small errors, less than 0.5 Dalton for average mass errors across the full mass range, the monoisotopic mass error is smaller under the low mass, such as below 3000 Dalton.

However, the errors become larger as the mass increases, which limited the application of AM when applied to larger protein molecules.



Figure 5: Absolute monoisotopic mass error distribution.



Figure 6: Absolute average mass error distribution.

3.2 Comparisons of Mass Errors between Close Masses

The masses with their corresponding estimated errors on two straight lines around 13,000 Dalton and 35,000 Dalton are shown in Figure 7 after zooming in for average mass error (left) or monoisotopic mass error (right). Under this circumstance, we found that even for the similar mass of molecules, they have totally different estimated errors, whether for average mass or monoisotopic mass. This has indicated that the different element may play a differently important role in the total contribution of mass errors caused.

In order to give a real-world validation of this phenomenon, the function *"isotopicdist"* of MATLAB was used to this comprehensive analysis.



Figure 7: Examples of estimated error distributions for both average and monoisotopic mass.

Figure 8 shows that similar nominal masses could generate a remarkable difference for their most abundant masses.



Figure 8: Computed isotopic distributions using *isotopicdist* with the protein nominal mass around 35,000 Dalton.

3.3 Error Contribution on the Element Level

In order to get the information on which element contributes most to the total estimated errors, we also analysed the individual contribution to mass errors on the elemental level for all those 20,431 proteins from the database.

Figure 9 shows the element contribution based on isotopic masses, which indicates that Carbon plays the most important mass role while Sulfur plays the most important role based on element coefficient.

However, in no matter what circumstances, Hydrogen always contributes the least in a single element mass.



Figure 9: Element contribution using different relative error calculation.

4 CONCLUSION

In order to further improve the accuracy of averagine model (AM), we systematically analysed two types of mass errors, i.e. monoisopis mass errors and average mass errors estimated by AM.

A method of calculation of massive error simultaneously has been developed through the process of attempting to figure out what element gives more contribution when forming a compound.

Our results on 20,431 human proteins (all of human proteins) shows that the mass error ranges are remarkablely different form these two types of errors.

Analysis of ours on the elemental level indicates that the element Carbon has the most important mass contribution while Sulfur contributes most in terms of the element coefficients.

All of these studies will provide a clue on how to further improve the performance of the averagine model.

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