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The Architectural Network for Protein Secondary Structure Prediction

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ABSTRACT

Over the past 25 years, the accuracy of proteins secondary structure prediction has improved substantially. Recently evolutionary information taken from the deviation of proteins in some structural family have again enhance prediction accuracy for all these residues predicted correctly is in one of the three sates helix, strands and others . The new methods developed over the past few years may be interesting in context of improvements which is achieved through combination of the existing methods. Evolutionary divergences profile posses' adequate information to improve protein secondary structure prediction accuracy. These profiles can also able to correctly predict long stretches of identical residues in other secondary structure. This sequence structure relationship may help to help to developed tool which can efficiently predict the protein secondary structure from its amino acid sequence.

KEYWORDS: Secondary structure, Evolution, Algorithm, Tool, programme

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Secondary structure is significantly determined by the hydrogen bonds made by biopolymers. That biopolymer was resolved by an atomic resolution structure. The building blocks of the secondary structure predicted from the information present into the amino acids sequences and analyzed through molecular modeling simulation. Proteins performed the major key role in almost all biometabolic process and their functional properties depend upon their structural folds (Yen-Ru Chen et al. 2008). Protein secondary structure prediction helps to determine a meaningful analysis of biological function. The linear sequence of amino acid means the primary structure is the basic information of a protein from which four state secondary structures can be predicted. Genome sequencing technologies now a day's widely used because it is relatively chief accurate and fast comparison with protein structure. Prediction from genome sequencing data linear amino acid chain can be determine by using various computational tools(Adams PD et al. 2013).Most accurate protein secondary structure prediction is a necessary step for improve modeling of a protein fold (Pirovano W and Heringa J. 2010) and determination of its biological function also (Sleator RD. 2012) for the prediction of three dimensional design (Das R. and Baker D.2008) and enzymatic function(Kiss G et al. 2013) as well as in drug design with development, we should model the secondary structure of a protein (Winter C et al. 2012). Linus Pauling accurately assumed the structural configuration of helixes and strands (Pauling L., and Corey R. B. 1951; Pauling L. et al. 1951).The theoretical concept of Pauling was verified with the first Xray structure published (Kendrew J.C. et al. 1960; Perutz M.F. et al. 1960). The Ramachandran angles present in the polypeptide chain and the rotation of the polypeptide backbone phi and psi bonds are described, which is present around the polypeptide. To determine the distribution of the R Ramachandran or torsion angles of a protein Ramachandran Plot is very much useful (Ramachandran G.N. et. al. 1963). Szent and his groupalready designed a method for the predition secondary structure from a primary sequences (Szent-Gyorgyi A.G. and Cohen C. 1957) based on each twenty amino acids propancity values, The first generation productive method became very popular later the segment of amino acids residues are taken to calculate from the previously used propancity values (Rost B. and Sander C. 2000). Although the accuracy level reached just above 60% due to the imaginable algorithm applied to calculate the percentage of residue present in a protein which is helix strand and others. The result of

three statehow partially some errordue to some restriction of local information. So we should introduce some global informatics parameter into the local ones (Dickerson R.E et al. 1976). Multiple sequence alignment information can improve the secondary structure prediction level (Zvelebil M. J, 1987).The third generation method for this prediction is designed from this multiple sequence alignment information and this concept is applied into an automatic prediction method increases the accuracy level upto 70% but this alignment method require a large number of dataset with more advance time management algorithm (Rost B. and Sander C. 2000 and 1993).The evolutionary significant data are the key component of the data set which is used in this new method.Sequence contain more than 35% pairwise identical residues with more than hundred align residues have similar structure isolated from natural source (Rost B. 1999).The natural mutational process shows sequence divergence increases the stability against the environmental hazards.Most of the mutations result in proteins that will not protect against environment only by the formation of globular structure. Substitution with lower number of residues shows adaptation against the extreme condition of environment. Exchange of amino acids shows specificity means position specific profiling gives important and crucial information about structure. This evolutionary divergence data set was the major informative key password for the prediction of secondary structure of protein in third generation. The most successful logic applies for secondary structure prediction apply machine learning algorithm which maximize the relationship between the primary sequence between the protein and their corresponding secondary structure(Kabsch W and Sander C. 1983). The DSSP programme successfully predict and improve the accuracy level above 80% but it depends upon the sample of protein sequences and their coordinate data set (Rost B. 2001).The coordinate data set of few proteins was stored into protein data bank and their corresponding secondary structure prediction began. In 1980 the first depositd protein structure in protein data bank data base was in membrane protein, which contain membrane helix as well as β strand (Westbrook J.D. et al. 2003; Engelman DM et al. 1986).Lather another way became very popular name homology modeling which can precisely predict both secondary and tertiary structure (Jones D.T. et al. 1992). Homology modeling accurately predicts the fold of corresponding structure by comparing closely related sequential data set deposited into the protein data bank (Sutcliffe M.J. 1987).Later in the 1990 the concept of

neural network and hidden Markov models were designed to improve the accuracy level of secondary structure predicted by homology modeling. Later it can be concluded that homology modeling based on the logical and theoretical concept neural network and hidden Markov models (Rost B. 1997; Rost B. 2001; Eyrich V A. 2001). Partialy solving the protein fold gradually increases the accuracy level of secondary structure prediction are most important in protein chemistry. The prediction accuracy directly or indirectly affect on how protein are to be analyzed and annotated to specify proteome analysis (Cozzetto D et al. 2005;Rost B et al. 2004). Critical assessment of structure prediction (CASP)method which shows more accuracy than the other structure prediction model which directly can predict secondary structure from primary sequences (Westbrook JD et al. 2003).The structural information of secondary structure helps to predict three dimensional model as well as many protein represent numerous important information, these informationof amino acids chain to determine their corresponding three dimensional structure. The explosive growth of sequence structural relationship information results the numerous growth of denovo prediction from sequence (Anfinsen C.B. 1973 and 1962).To determine secondary structure of a protein the de novo folding based approaches has been taken called state-of-the-art. Which is based on the sequence structural similarity present in to the structural data base and the similar fragments are assembled by using empirical intermolecular force fields. Such logical approaches have worked favorably in cases for smaller peptides (Bradley P et al. 2005; Raman S. et al. 2010; Lange O F. et al 2012). Our present effort for the better structural prediction enabled us for a clear assumption about the proteins' structure-function relationship. We have generated some software tools which more efficiently predict the stress withstanding abilities in a protein form its amino acid sequence. We also made an extensive effort to develop software for the secondary structure prediction from the sequence of a protein. Conclusively, studies on proteins structure can generate a strong base of understanding the organismal behavior/existence and speciation/species proliferation on course of long evolutionary period and even for the course of the coming period.

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