

# Protein Structure Prediction Using Robust Principal Component Analysis and Support Vector Machine

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## ABSTRACT

Existence of bioinformatics is to increase the further understanding of biological process. Proteins structure is one of the major challenges in structural bioinformatics. With former knowledge of the structure, the quality of secondary structure, prediction of tertiary structure, and prediction function of amino acid from its sequence increase significantly. Recently, the gap between sequence known and structure known proteins had increase dramatically. So it is compulsory to understand on proteins structure to overcome this problem so further functional analysis could be easier. The research applying RPCA algorithm to extract the essential features from the original high-dimensional input vectors. Then the process followed by experimenting SVM with RBF kernel. The proposed method obtains accuracy by 84.41% for training dataset and 89.09% for testing dataset. The result then compared with the same method but PCA was applied as the feature extraction. The prediction assessment is conducted by analyzing the accuracy and number of principal component selected. It shows that combination of RPCA and SVM produce a high quality classification of protein structure

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## 1. Introduction

The functional and structural annotation of protein domain is one of the important roles in bioinformatics. In this context, protein structure information plays an important information key of their structural part also the features related to the biological function (S.S. Sahu et al., 2009) such as prediction of DNA binding site, implementation of a heuristic approach to find tertiary structure, reduction of conformation search space and also characterizing the folding type of a protein or its domain. S. Zhang et al. (2012) state that the exponentially growth of newly discovered protein sequences by different scientific community caused a large gap between the number of sequence-known and the number of structure-known proteins. Hence, there exist critical challenges to develop automated method for fast and accurate determination of the structures of proteins in order to reduce gap. Therefore, there is a compulsory to implement reliable and effective computational methods for identifying the structural class of newly discovered protein based on their primary sequences.

## 2. Objectives

The purposes of this research are: 1. To implement Robust Principal Component Analysis (RPCA) to determine the number of principal component. 2. To implement Support Vector Machine (SVM) for protein structure classification. 3. To evaluate the performance of RPCA and SVM based on accuracy

## 3. Methodology

Firstly, the current issues of protein structure prediction are investigated followed by collecting research materials such as journals, articles, conference paper and others. The data preprocessing conducted to gain higher and better prediction success rate and system performance. It also help to minimizing error in preparation be validated by machine learning algorithm. Datasets by Ding and Dubchak (2012) filtered to remove unnecessary values and information. Research continues by applying Principal component analysis (PCA) and RPCA (Croux and Ruiz-Gazen, 2005)) algorithm to extract the essential features from the original high-dimensional input vectors. The process continued by experimenting SVM with RBF kernel using the reduced and normalized features by PCA and RPCA. The final phase is the prediction assessment of the application of RPCA and SVM by the comparison of recognition ratio compared between different methods and methods used by previous researcher. Performance testing of this research by comparing classification result of protein by overall accuracy that expressed in equation 1.

$$\text{correctly recognize protein} = \frac{\text{correctly recognize number of query protein}}{\text{total number of protein}} \quad (1)$$

## 4. Result and Discussion

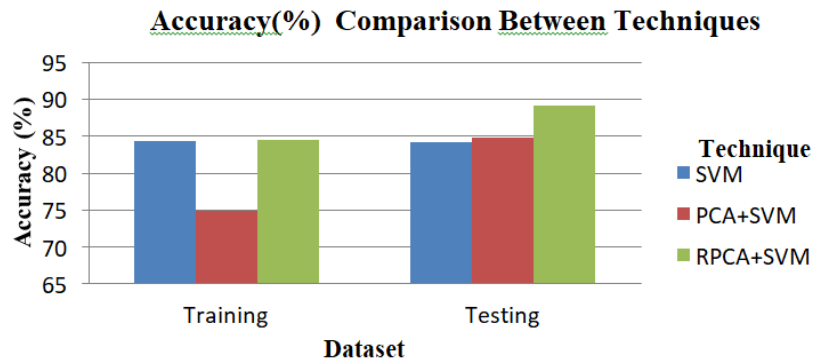
The experiment was conducted by using three approaches in order to analyse the performance of RPCA and SVM. In order to gives a clear view on performance of RPCA, the method was compared with the PCA (the basic of RPCA) and SVM. In order to select the components that contain >60% of variance, the number of PC selected are different accordingly. Table 1 shows that number of selected PC in training dataset is lower compared to testing dataset. Table 2 shows the accuracy percentage of tested approach divided by training and testing datasets.

**Table 1.** Number of PC selected for classification

Feature extraction	Number of PC selected for classification	
	Training	Testing
PCA	2	3
RPCA	2	4

**Table 2.** Comparison of SVM, PCA + SVM and RPCA + SVM

Technique	Training Dataset Accuracy (%)	Testing Dataset Accuracy (%)
SVM	84.25	84.16
PCA+SVM	74.79	84.68
RPCA+SVM	84.41	89.09



**Fig 4.** Accuracy comparison between techniques.

Based on this analysis, it can be assumed that difference between data characteristics will influence the number of sufficient PCs required in both PCA and RPCA approaches. The number of PCs required for the training dataset is less than for the testing dataset since the size of the training dataset is larger, so it may contain higher information and better interpretation of features compared to the testing dataset.

From the results in Figure 1, it can be seen that the non-extracted features technique (only SVM) gains a high percentage of accuracy (84.25% and 84.16%). However, the result can be doubted since models built on extracted features may be of higher quality, because the data is described by fewer, more meaningful attributes. Results obtained by the combination of PCA and SVM are 74.79% on the training dataset and 84.68% on the testing dataset. The accuracy on both datasets is quite high but still lower than the combination of RPCA and SVM technique (84.41% on the training dataset and 89.09% on the testing dataset). The gap seems to be higher in the training dataset, possibly because of the larger number of outliers. RPCA seems to perform the best since this method is not influenced much by outliers and its ability to detect exact fit situations.

Table 3 shows the comparison of accuracy percentages of PCA and RPCA combinations with SVM. Even according to the number of components, the RPCA method always seems to lead in terms of accuracy. This proves the effectiveness of the RPCA approach. Table 3 also shows the increasing pattern of accuracy for both datasets. It can be assumed that a higher number of PCs contain much more data information, leading to higher accuracy.

**Table 3.** Comparison of PCA+SVM and RPCA+SVM based on number of components

Number of Principal Component (PC)	Accuracy (%) for Training Dataset		Accuracy (%) for Testing Dataset	
	PCA +SVM	RPCA+SVM	PCA+SVM	RPCA+SVM
1	51.91	80.60	55.84	77.92
2	74.79	84.41	84.68	84.94
3	82.75	86.90	87.53	88.05

L. Singh, G.Chetty and D.Sharma (2012) apply the same dataset (feature vector described by Ding and Dubchak, 2001) to predict protein structure using PCA and LDA based in Extreme Learning Machine (ELM). According to Table 4, it can be seen that the proposed method used in this research shows promising results in terms of accuracy obtained compared to the proposed method proposed by L. Singh, G.Chetty and D.Sharma (2012). This shows that feature extraction using

RPCA and classification using SVM is an efficient method for protein structure prediction. It also shows that method proposed by L. Singh, G.Chetty and D.Sharma (2012) has drawbacks in due to the outliers and low ability in detection of exact fit situation.

**Table 4.** Accuracy comparison between method

Method	Accuracy (%)
LDA-ELM	77.67
PCA-ELM	82.45
RPCA-SVM	89.09

## 5. Conclusion

This research focus is on protein structural classification. Protein Structure classification is important for identification of protein function. As the protein structure classification is a first and key step in protein structure prediction, it becomes an increasingly challenging task. Recently, the exponentially increase of sequence data protein cause the increasing of the requirements for reliable and effective computational method for protein structure classification. Protein structure classification is very important in bioinformatics field. Proposed feature extraction method, Robust Principal Component Analysis (RPCA) combines with Support Vector Machine (SVM) shows that data with extracted features can obtain higher accuracy (84.41% for training dataset and 89.09% for testing dataset). It also shows that RPCA works well with highly corrupted data especially dataset with outliers.

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