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**Abstract Topic:-** Cancer

**Abstract Title:-** Classification of Pathogenicity from Germline Missense Variants using Machine Learning Algorithms

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**Aims:-**The present study had proposed a novel machine learning-based tool to classify between benign and pathogenic missense variants.

**Methods:-** There were 222,540 germline missense variants obtained from 64 gastric cancer patients via next generation sequencing (NGS). The missense variants were subjected to pathogenicity prediction tools: MutationTaster, FATHMM, LR, LRT, RadialSVM, SIFT, and Polyphen2 to acquire appropriate class labels (pathogenic or benign). A missense variant was labeled pathogenic if and only if all the above-seven pathogenicity prediction tools predicted it as pathogenic; otherwise it had been labeled as benign. The dataset contained 200171 benign and 22369 pathogenic missense variants with 62 features. The dataset was randomly divided into 70% for training and 30% for testing. Extra trees classifier algorithm was implemented to extract the features of importance from 62 independent features using 5-fold cross validation via gridsearchCV on training dataset and obtained the best estimators. A set of 5-ensemble algorithms was chosen to classify between pathogenic and benign missense variants: Random Forest (RF), Bagging classifier (BC), Extra Trees (ET), AdaBoost (AB) and Gradient Boosting (GB). The dataset was divided into five sets based on feature significance: top 30 features, top 20 features, top 10 features, top 8 features, and all features.

**Results:-** The five ensemble models' hyper-parameters were tuned using 5-fold gridsearchCV on training dataset:  
ExtraTreesClassifier(max\_depth=100, n\_estimators=50),  
BaggingClassifier(estimator=DecisionTreeClassifier(), n\_estimator=100),  
RandomForestClassifier(max\_depth=50, n\_estimators=50), AdaBoostClassifier(n\_estimators=1000) and GradientBoostingClassifier(n\_estimators=1000). RF, BC and ET models had shown outstanding performances on all five test datasets: accuracy, precision, recall and f1\_score of 99% each individual evaluation metric with Matthew's correlation coefficient (MCC) of 1.0 and precision\_recall (PR) curve of 1.0 for both benign and pathogenic classes. The presented models showed that top 8 features: phyloP46way\_placental, VEST3\_score, ExAC\_AFR, AFR.sites.2015\_08, AF\_afr, SiPhy\_29way\_logOdds, and ALL.sites.2015\_08 scores can be playing a significant role in determining the pathogenicity of the missense variant.

**Conclusions:-** The performance of generated in-house machine learning tool had clearly paved way to improve the patients' diagnosis rate and to identify novel disease-specific variants with high probability; furthermore, this knowledge transfer will aid in personalized-precision genomic medicine to reduce the cancer incidence and early diagnosis.

**Keywords:-** germline missense variants, gastric cancer, machine learning, pathogenicity, precision\_recall curve