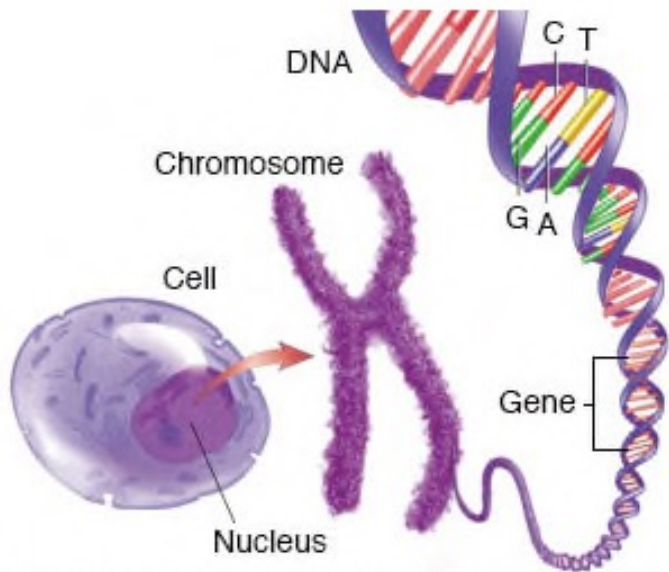


Genetic Variant Classifications

Ni-Ting Chiou

Genetic variants come from the changes of DNA sequences



Classes of human genetic variants.

Single nucleotide variant

SNP

Insertion-deletion variant

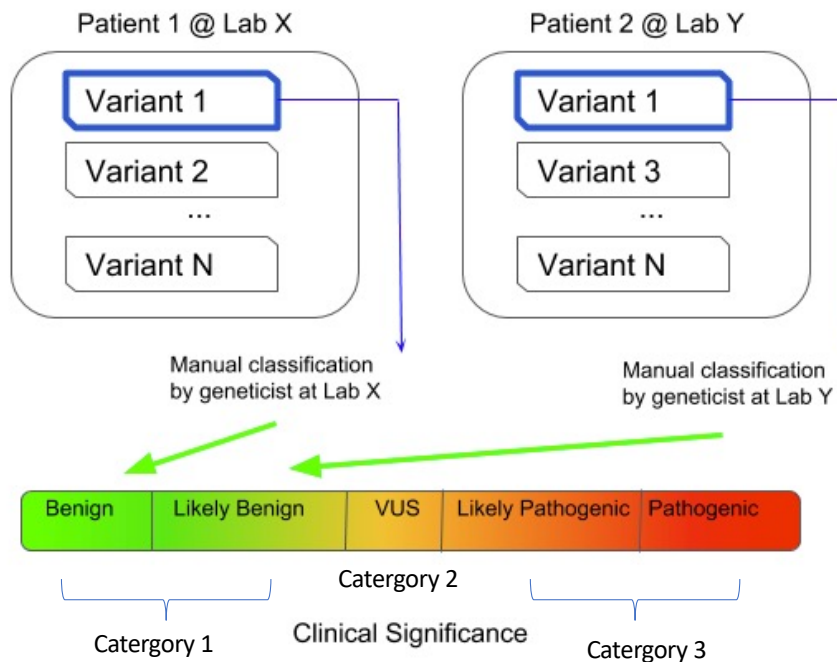
Indel

ATTGGCCTTAACCC	CCGATTATCAGGAT	REF
ATTGGCCTTAACCT	CCGATTATCAGGAT	Sequence of interest
ATTGGCCTTAACCC	GATCCGATTATCAGGAT	REF
ATTGGCCTTAACCC	---CCGATTATCAGGAT	Sequence of interest

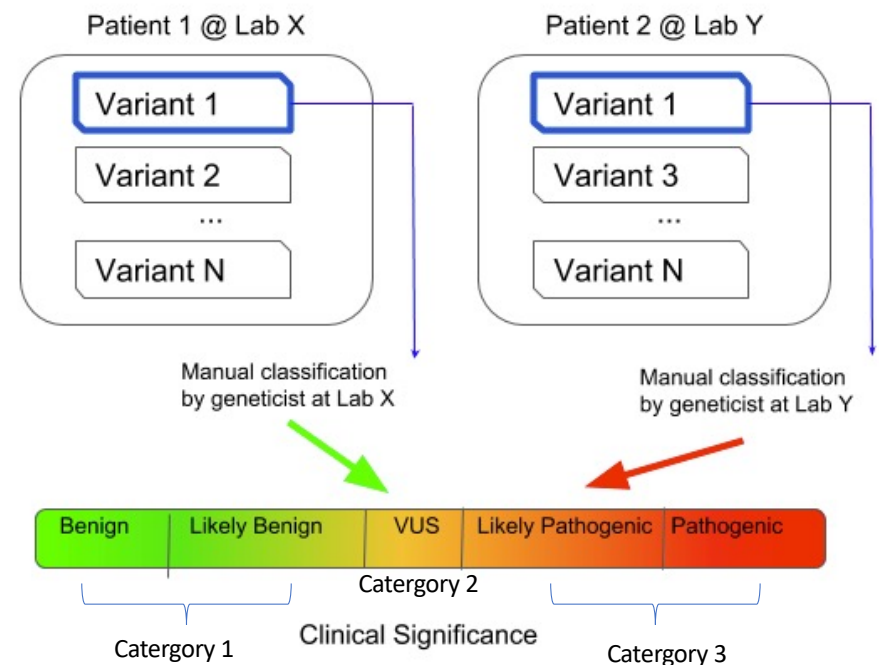
- Variants might have negative, little or no effect to diseases

Genetic variants are classified manually which resulting in conflicting classification

Concordant Variant Classification - Class: 0



Conflicting Variant Classification - Class: 1



Data exploration analysis

clinvar_conflicting.csv (Kaggle)
(46 features)

Remove features

1. Redundant
2. Not correlated
3. Have > 90% nan

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 65188 entries, 0 to 65187
Data columns (total 9 columns):
#   Column      Non-Null Count  Dtype
---  -
0   CHROM        65188 non-null  object
1   CLNVC        65188 non-null  object
2   MC           64342 non-null  object
3   IMPACT       65188 non-null  object
4   SYMBOL       65172 non-null  object
5   AF_ESP       65188 non-null  float64
6   LoFtool      60975 non-null  float64
7   CADD_PHRED   64096 non-null  float64
8   CLASS        65188 non-null  int64
dtypes: float64(3), int64(1), object(5)
memory usage: 4.5+ MB
```

Categorical features :

CHROM- chromosome

CLNVC - Variant Type

MC - Molecular consequence

IMPACT - the impact of the variants

SYMBOL - Gene Name

Numerical features:

AF_ESP - Allele frequencies of variants

LoFtool - Loss of Function tolerance score

CADD_PHRED - Scoring the deleteriousness of the variants

Target:

class 0 (concordant variant classification)

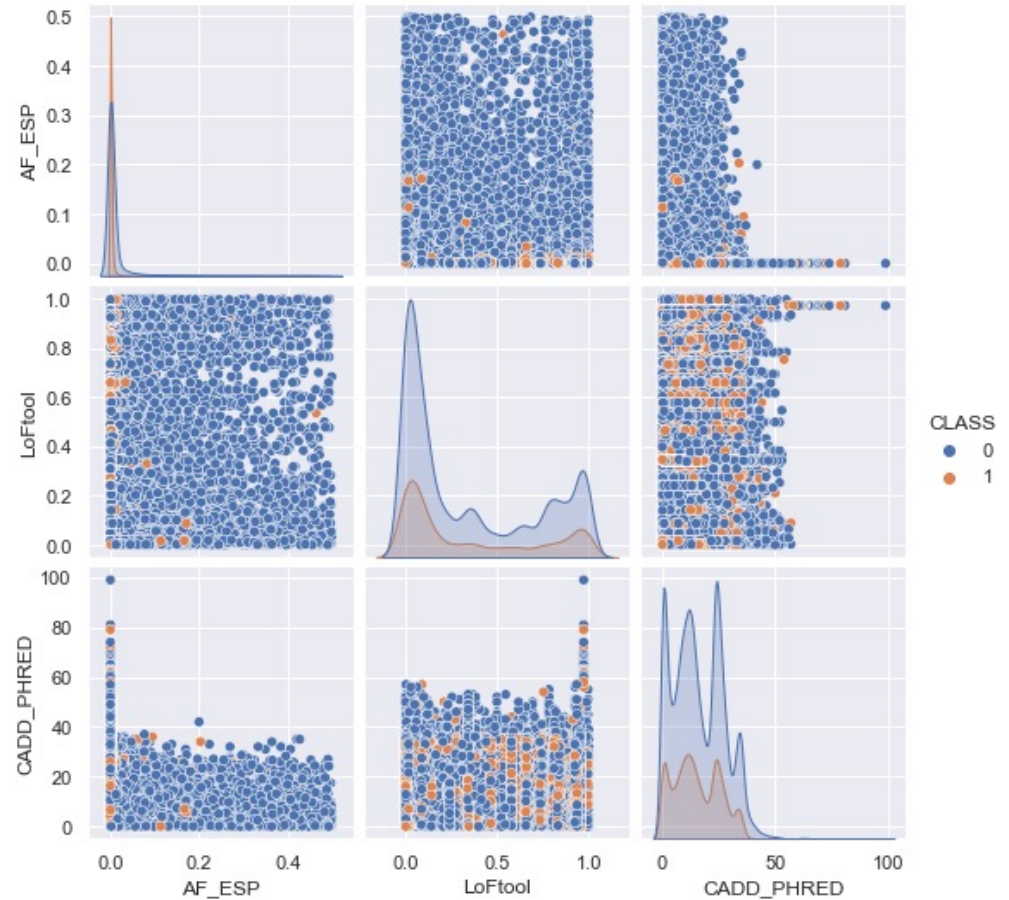
class 1 (conflicting variant classification)

AF_ESP and SYMBOL
are more distinguishable
among 2 classes

Chi2 test for **categorical features**
(p-value)

	CLASS
CHROM	1.407244e-05
CLASS	NaN
IMPACT	1.856664e-191
SYMBOL	6.362397e-309

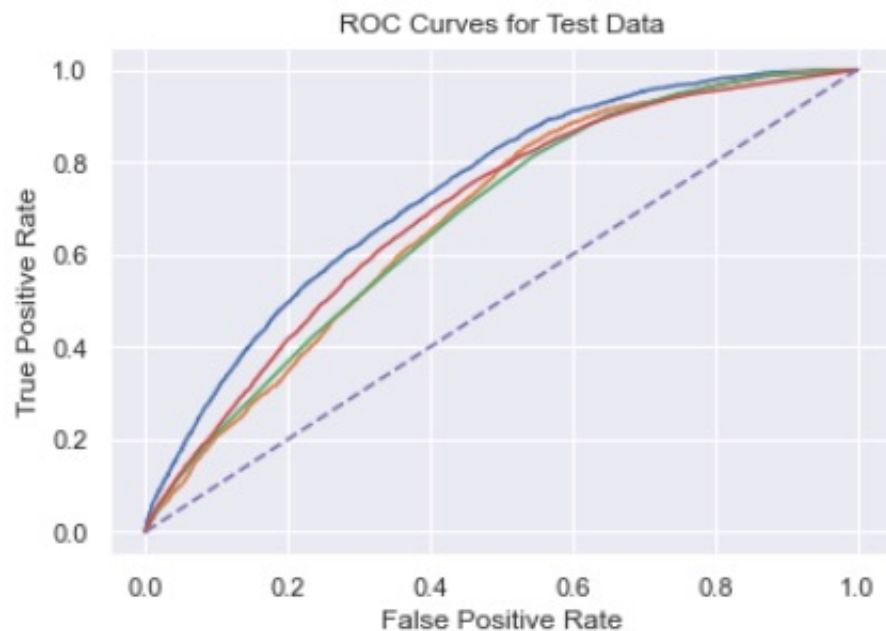
Pairplot for **numerical features**



Data preprocessing

- Convert categorical features into dummies variables
- Scale the numerical data
- Class rebalance
- Model fitting (GridSearch for hyperparameter optimalization)

XGBoost tree has better performance for the prediction



```
XGBClassifier(base_score=0.5, booster='gbtree', colsample_bylevel=1,
               colsample_bynode=1, colsample_bytree=0.8,
               enable_categorical=False, gamma=0, gpu_id=-1,
               importance_type=None, interaction_constraints="",
               learning_rate=0.05, max_delta_step=0, max_depth=4,
               min_child_weight=3, missing=nan, monotone_constraints='()',
               n_estimators=300, n_jobs=8, num_parallel_tree=1, predictor='auto',
               random_state=0, reg_alpha=0, reg_lambda=1, scale_pos_weight=1,
               subsample=0.8, tree_method='exact', validate_parameters=1,
               verbosity=None)
```

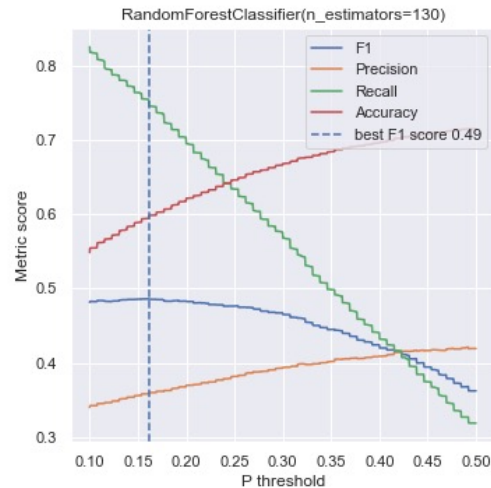
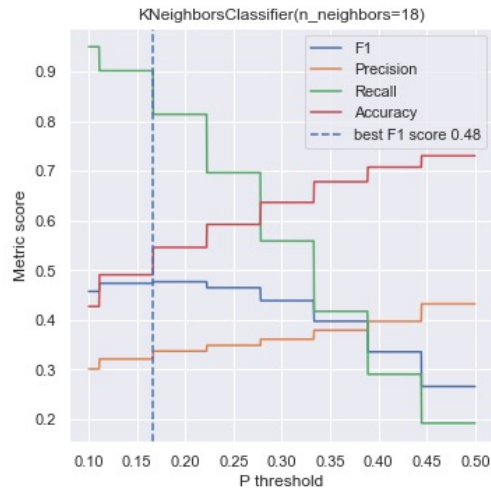
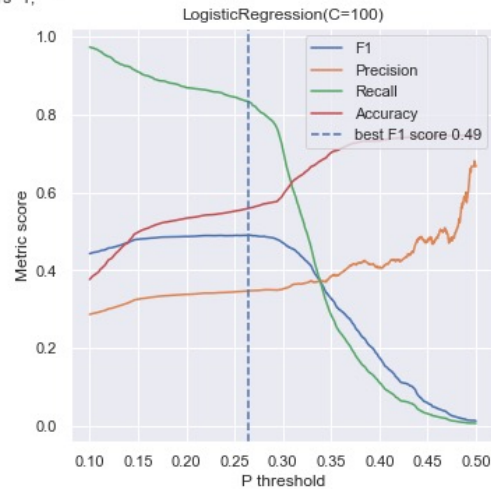
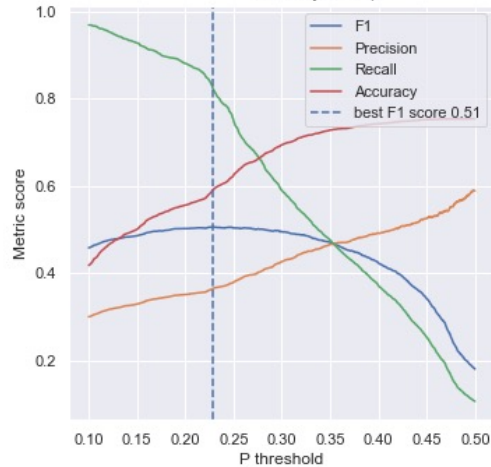
— LogisticRegression(C=100)

— KNeighborsClassifier(n_neighbors=18)

— RandomForestClassifier(n_estimators=130)

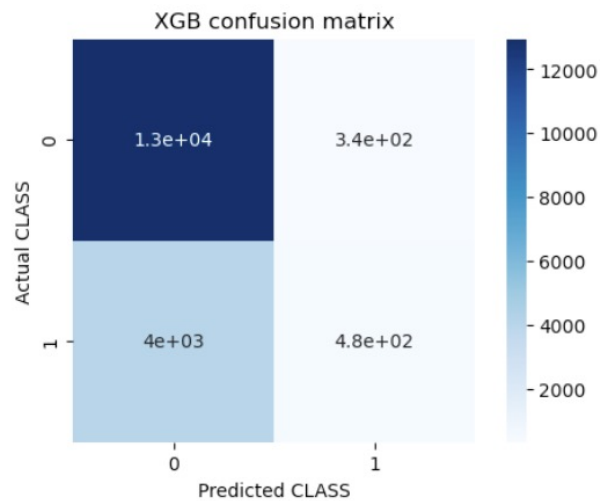
- - - No Skill

```
XGBClassifier(base_score=0.5, booster='gbtree', colsample_bylevel=1,
             colsample_bynode=1, colsample_bytree=0.8,
             enable_categorical=False, gamma=0, gpu_id=-1,
             importance_type=None, interaction_constraints='',
             learning_rate=0.05, max_delta_step=0, max_depth=4,
             min_child_weight=3, missing=nan, monotone_constraints=(),
             n_estimators=300, n_jobs=8, num_parallel_tree=1, predictor='auto',
             random_state=0, reg_alpha=0, reg_lambda=1, scale_pos_weight=1,
             subsample=0.8, tree_method='exact', validate_parameters=1,
             verbosity=None)
```



XGB model has the highest F1 scores at the P threshold of 0.22

AF (variant frequency), symbol (gene names) and MC(molecular consequences) are the important features

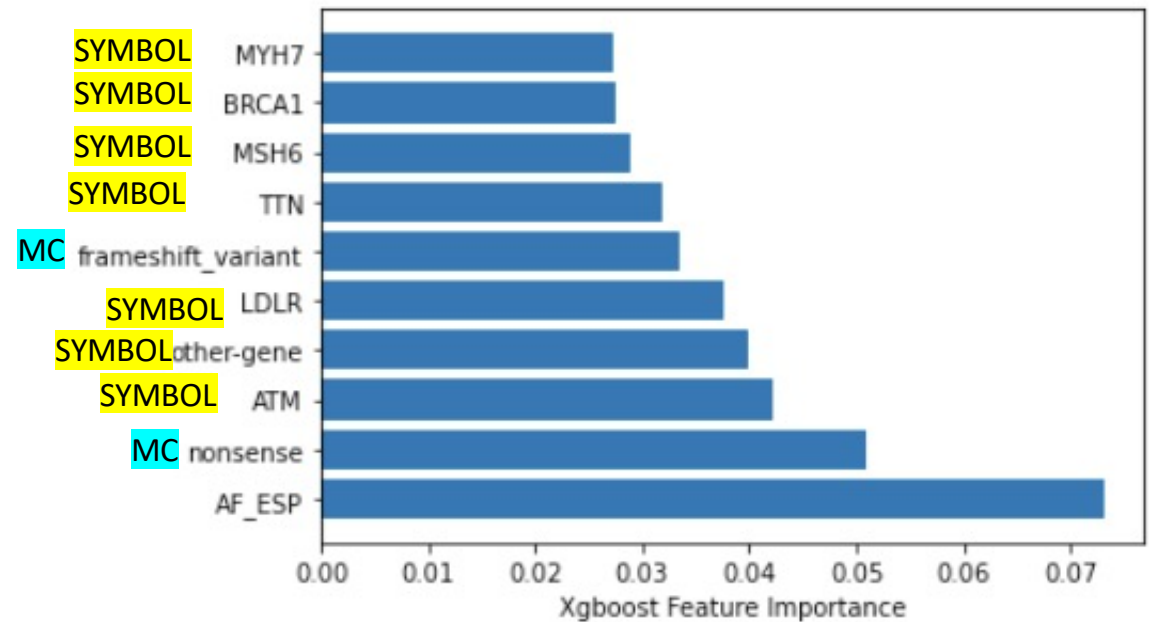


Training set

f1 score: 0.19
Precision score: 0.65
Recall score: 0.11
Accuracy score: 0.76

Test set

f1 score: 0.18
Precision score: 0.59
Recall score: 0.11
Accuracy score: 0.75



Discussion and future work

- The variants with the low allele frequency (AF), do not have the known deleterious molecular consequence (MC) and located at the cancer genes (SYMBOL) tend to have the conflicting classification.
- The variants within the conflicting classification can be compared with the cancer variant databases to be classified better.