# CSE 472 - Machine Learning Sessional

Offline-2 Report

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#### How to run the code:

- 1. First, install the ucmlrepo package by running pip install ucimlrepo.
- 2. main\_test is the main function of getting the metrics of normal logistic regression(average of 9), stacking, and voting ensemble. Input your choice ( 'adult', 'churn', or 'credit\_card') to get the metrics of the corresponding dataset. You can also pass in parameters like learning rate, lambda 1, lambda 2, regularization choice, maximum epoch, and the value of k(in the selection of top k features), in the function instead of using the default hyperparameters. I have kept the respective calls for each dataset in separate cells to make it easy to run any of your choices.

```
def main_test(choice, lr=1, max_epoch=1000, n_estimators=9, regularization='l2', l1_lambda=0.01, l2_lambda=0.01, k=20, feature_selection='correlation'):

churn_metrics_df,metrics_churn = main_test('churn')
    # churn_metrics_df,metrics_churn = test('churn', lr=0.1, max_epoch=1000, n_estimators=9, regularization='l2', l1_lambda=0.01, l2_lambda=0.01, k=20, feature_churn_metrics_df

24.7s

Dataset: churn

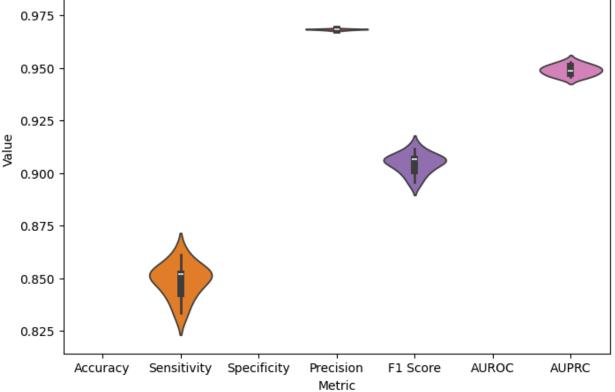
adult_metrics_df,metrics_adult = main_test('adult')
    # adult_metrics_df,metrics_adult = test('adult', lr=0.1, max_epoch=1000, n_estimators=9, regularization='l2', l1_lambda=0.01, l2_lambda=0.01, k=20, feature_adult_metrics_df, metrics_adult = test('adult', lr=0.1, max_epoch=1000, n_estimators=9, regularization='l2', l1_lambda=0.01, l2_lambda=0.01, k=20, feature_adult_metrics_df, metrics_credit_card = main_test('credit_card')
    # credit_metrics_df, metrics_credit_card = test('credit_card', lr=0.1, max_epoch=1000, n_estimators=9, regularization='l2', l1_lambda=0.01, l2_lambda=0.01
    # credit_metrics_df, metrics_credit_card = test('credit_card', lr=0.1, max_epoch=1000, n_estimators=9, regularization='l2', l1_lambda=0.01, l2_lambda=0.01
    # credit_metrics_df, metrics_credit_card = test('credit_card', lr=0.1, max_epoch=1000, n_estimators=9, regularization='l2', l1_lambda=0.01, l2_lambda=0.01
    # credit_metrics_df, metrics_credit_card = test('credit_card', lr=0.1, max_epoch=1000, n_estimators=9, regularization='l2', l1_lambda=0.01, l2_lambda=0.01
    # credit_metrics_df, metrics_credit_card = test('credit_card', lr=0.1, max_epoch=1000, n_estimators=9, regularization='l2', l1_lambda=0.01, l2_lambda=0.01
    # credit_metrics_df, metrics_credit_card = test('credit_card', lr=0.1, max_epoch=1000, n_estimators=9, regularization='l2', l1_lambda=0.01, l2_lambda=0.01
    # credit_metrics_df, metrics_credit_card = test('credit_card', lr=0.1, max_epoch=1000, n_estimators=9, regularization='l2', l1_lambda=0.01, l2_lambda=0.01
    # credit_metrics_df, metrics_credit_card = test('credit_card', lr=0.1, max_epoch=1000, n_
```

# **Telco-customer churn:**

Method	Accuracy	Sensitivity	Specificity	Precision	F1-score	AUROC	AUPRC
	0.801977	0.508207	0.900179	0.630024	0.562361	0.844862	0.622140
LR	± 0.002194	± 0.018722	± 0.005171	± 0.005927	± 0.010936	± 0.001314	± 0.002311
Voting ensembl e	0.802135	0.502841	0.902184	0.632143	0.560127	0.845351	0.624513
Stacking ensembl e	0.800712	0.536932	0.888889	0.617647	0.574468	0.839853	0.616814

Violin Plot of Model Metrics for credit\_card

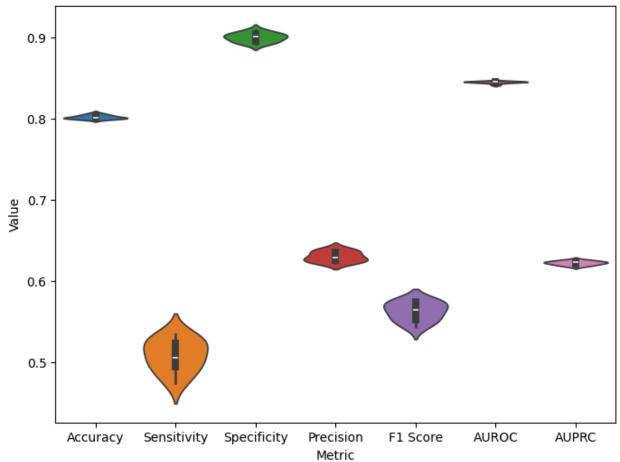




## Adult:

Method	Accuracy	Sensitivity	Specificity	Precision	F1-score	AUROC	AUPRC
LR	0.841647	0.524846	0.944226	0.752922	0.618502	0.895662	0.742335
	±	±	±	±	±	±	±
	0.000845	0.006208	0.001103	0.001924	0.003987	0.000202	0.000829
Voting ensembl e	0.841647	0.525544	0.944	0.752398	0.618836	0.89572	0.742479
Stacking ensembl e	0.841954	0.526801	0.944	0.752843	0.619857	0.895754	0.743677

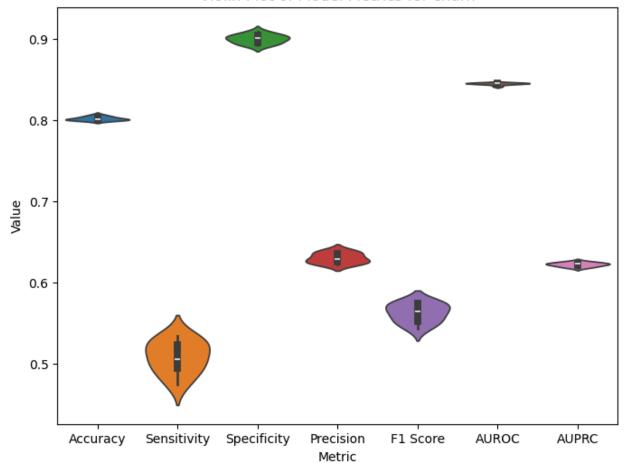




### **Creditcard fraud:**

Method	Accuracy	Sensitivity	Specificity	Precision	F1-score	AUROC	AUPRC
LR	0.995253 ± 0.000200	0.848765 ± 0.007560	0.999243 ± 0.000000	0.968307 ± 0.000275	0.904587 ± 0.004423	0.993636 ± 0.002241	0.948934 ± 0.002314
Voting ensembl e	0.995335	0.851852	0.999243	0.968421	0.906404	0.994629	0.949548
Stacking ensembl e	0.995335	0.851852	0.999243	0.968421	0.906404	0.988471	0.947661





#### Observation:

1. For 9 9 LR models and Voting ensemble, F1 score decreases drastically if learning rate is decreases