ML Project Report: Lend or Lose (Checkpoint 2)

Team Members

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Github Link: https://github.com/truptikhodwe/Lend-or-Lose

Task

Training ML models to predict which individuals are at the highest risk of defaulting on their loans is essentially a **binary classification problem**, where the target variable is the likelihood of default. For checkpoint 2, we were supposed to train the dataset on the following models:

- Support Vector Machine (SVM)
- Logistic Regression
- Neural Networks

And also apply skew reduction to reduce bias in the data.

For EDA, pre-processing and results for models of checkpoint 1, please refer to the previous report: Report 1: Lend or Lose

Results for various models (Checkpoint 2)

Sr. no	Model Name	Validation Accuracy	Test Accuracy	Best hyperparameters
1.	SVM without skew reduction	88.59	88.447	<pre>Default: {'C': 1.0, 'break_ties': False, 'cache_size': 200, 'class_weight': None, 'coef0': 0.0, 'decision_function_shape': 'ovr', 'degree': 3, 'gamma': 'scale', 'kernel': 'rbf', 'max_iter': -1, 'probability': False, 'random_state': None, 'shrinking': True, 'tol':</pre>

				0.001, 'verbose': False}
2.	SVM with skew reduction	67.99	68.241	<pre>{'C': 1.0, 'break_ties': False, 'cache_size': 200, 'class_weight': None, 'coef0': 0.0, 'decision_function_shape': 'ovr', 'degree': 3, 'gamma': 'scale', 'kernel': 'rbf', 'max_iter': -1, 'probability': False, 'random_state': None, 'shrinking': True, 'tol': 0.001, 'verbose': False}</pre>
3.	Neural Networks without skew reduction	88.49	88.44	The optimal number of units in the first densely-connected layer is 128, in the second layer is 288 and the optimal learning rate for the optimizer is 0.0001.
4.	Neural Networks with skew reduction	56.74	87.52	The optimal number of units in the first densely-connected layer is 448, in the second layer is 224 and the optimal learning rate for the optimizer is 0.0001.
5.	Logistic Regression without skew reduction	88.48	88.586 (Highest Accuracy)	N/A
6.	Logistic Regression with skew reduction	42.06	48.682	N/A

Results for various models (checkpoint 1)

Sr. no	Model Name	Validation Accuracy	Test Accuracy	Best hyperparameters
1.	Decision Tree without feature selection	88.40	87.879	{'criterion': 'entropy', 'max_depth': 10, 'min_samples_leaf': 4, 'min_samples_split': 10}
2.	Decision Tree with feature selection	88.54	88.388	{'criterion': 'entropy', 'max_depth': 10, 'min_samples_leaf': 4, 'min_samples_split': 10}
3.	K Nearest Neighbour s	88.44	88.214	{'metric': 'manhattan', 'n_neighbors': 11, 'weights': 'uniform'}
4.	Random Forest Classifier	88.81	88.686	{'max_depth': 20, min_samples_leaf': 2, 'min_samples_split': 10, 'n_estimators': 100}
5.	XGBoost without feature selection	88.709	88.774	{'colsample_bytree': 0.8, 'learning_rate': 0.1, 'max_depth': 3, 'n_estimators': 200, 'subsample': 0.8}
6.	XGBoost with feature selection	88.74	88.791 (Highest Accuracy)	{'colsample_bytree': 0.8, 'learning_rate': 0.2, 'max_depth': 3, 'n_estimators': 100, 'subsample': 0.8}
7.	Gradient Boosting without feature selection	88.696	88.748	{'learning_rate': 0.1, 'max_depth': 3, 'min_samples_split': 5, 'n_estimators': 100, 'subsample': 0.8}

8.	Gradient Boosting with feature selection	88.714	88.768	{'learning_rate': 0.1, 'max_depth': 3, 'min_samples_split': 5, 'n_estimators': 100, 'subsample': 0.8}
9.	Adaboost	88.672	88.742	{'learning_rate': 1.0, 'n_estimators': 100}
10.	Naive Bayes Classifier	88.505	88.484	N/A
11.	Linear & Polynomial Regressio n without feature selection	9.96	88.490	N/A
12.	Linear & Polynomial Regressio n with feature selection	10.14	88.447	N/A

Details about the models used

- 1. Support vector machine
 - a. The preprocessing which I did was the same as before. I applied the SVM model in two cases, one where I did skew reduction and one where I did not do skew reduction.
 - b. Results when I did not do skew reduction:
 - i. I trained the model using default hyper parameters, which

```
Were: {'C': 1.0, 'break_ties': False, 'cache_size':
200, 'class_weight': None, 'coef0': 0.0,
'decision_function_shape': 'ovr', 'degree': 3,
'gamma': 'scale', 'kernel': 'rbf', 'max_iter': -1,
'probability': False, 'random_state': None,
'shrinking': True, 'tol': 0.001, 'verbose': False}
```

ii. Model accuracy with default hyper parameters was 88.59

- iii. Applied the same model on the test.csv testing data and got a score of 88.447 on kaggle.
- iv. Saved the results in the following file: submission_svm_default_without_smote.csv
- v. Note: I tried applying GridSearchCV for finding the best hyper parameters but it was taking a lot of time, so I stopped running it after almost 40 hrs.

```
Fitting 5 folds for each of 32 candidates, totalling 160 fits
[CV] END ......C=0.1, gamma=1, kernel=rbf; total time=35.8min
[CV] END ......C=0.1, gamma=1, kernel=rbf; total time=37.7min
[CV] END ......C=0.1, gamma=1, kernel=rbf; total time=39.3min
[CV] END ..... total time=210.9min
[CV] END ......footal time=36.5min
[CV] END ......total time=236.2min
```

- c. Results when I did skew reduction:
 - i. I trained the model using default hyper parameters, which

```
Were: {'C': 1.0, 'break_ties': False, 'cache_size':
200, 'class_weight': None, 'coef0': 0.0,
'decision_function_shape': 'ovr', 'degree': 3,
'gamma': 'scale', 'kernel': 'rbf', 'max_iter': -1,
'probability': False, 'random_state': None,
'shrinking': True, 'tol': 0.001, 'verbose': False}.
```

- ii. Model accuracy with default hyper parameters was 67.99
- iii. Applied the same model on the test.csv testing data and got a score of 68.241 on kaggle.
- iv. Class distribution after SMOTE:

Default:

- 0 144328
- 1 144328
- v. Saved the results in the following file: submission_svm_default_with_smote.csv

2. Neural Networks:

- a. The pre-processing remains the same as before. The model was trained twice, once with skew reduction and once without.
- b. Without skew reduction:
 - i. Before applying the model, I tried hyperparameter tuning using Bayesian optimization. I set the tuner values to:

```
max trials = 10, executions per trial = 2
```

And the tuner search ran for 10 epochs.

ii. This gave the following best parameters:

```
The optimal number of units in the first densely-connected layer is 128, in the second layer is 288 and the optimal learning rate for the optimizer is 0.0001.
```

iii. The model was then fitted on the training data split with epochs = 50.

```
Epoch 1/50
/usr/local/lib/python3.10/dist-packages/keras/src/layers/core/dense.py:87: UserWarning: Do not pass an `input_shape`/`input_dim
  super().__init__(activity_regularizer=activity_regularizer, **kwargs)
                              14s 3ms/step - accuracy: 0.7980 - loss: 50.8748 - val_accuracy: 0.8855 - val_loss: 10.6825
Epoch 2/50
4086/4086 -
                             - 9s 2ms/step - accuracy: 0.8045 - loss: 23.9302 - val_accuracy: 0.8340 - val_loss: 7.3250
Epoch 3/50
4086/4086
                              10s 2ms/step - accuracy: 0.7998 - loss: 17.4185 - val_accuracy: 0.8854 - val_loss: 31.3737
Epoch 4/50
4086/4086 -
                             - 8s 2ms/step - accuracy: 0.8035 - loss: 11.6698 - val accuracy: 0.6049 - val loss: 4.3434
Epoch 5/50
4086/4086
                              9s 2ms/step - accuracy: 0.8007 - loss: 8.2261 - val_accuracy: 0.8854 - val_loss: 6.1486
Epoch 6/50
                             - 9s 2ms/step - accuracy: 0.8050 - loss: 5.7862 - val_accuracy: 0.3283 - val_loss: 3.7733
4086/4086 -
Epoch 7/50
                             - 9s 2ms/step - accuracy: 0.8043 - loss: 3.6193 - val_accuracy: 0.8854 - val_loss: 1.6016
4086/4086 -
Epoch 8/50
4086/4086 -
                              • 10s 2ms/step - accuracy: 0.8060 - loss: 2.3307 - val_accuracy: 0.8858 - val_loss: 0.5740
Epoch 9/50
4086/4086 -
                              11s 3ms/step - accuracy: 0.8159 - loss: 1.5069 - val_accuracy: 0.8857 - val_loss: 0.5377
Epoch 10/50
4086/4086 -
                              11s 3ms/step - accuracy: 0.8270 - loss: 0.8329 - val_accuracy: 0.8854 - val_loss: 1.0164
Epoch 11/50
4086/4086
                               20s 2ms/step - accuracy: 0.8517 - loss: 0.5307 - val_accuracy: 0.8854 - val_loss: 0.3676
Epoch 12/50
4086/4086
                              9s 2ms/step - accuracy: 0.8634 - loss: 0.4775 - val_accuracy: 0.8854 - val_loss: 0.3895
Epoch 13/50
                              11s 2ms/step - accuracy: 0.8681 - loss: 0.4446 - val accuracy: 0.8854 - val loss: 0.3634
4086/4086
Epoch 50/50
4086/4086 -
                             - 9s 2ms/step - accuracy: 0.8828 - loss: 0.3613 - val_accuracy: 0.8854 - val_loss: 0.3560
                             - 2s 2ms/step - accuracy: 0.8857 - loss: 0.3555
1277/1277 -
[test loss, test accuracy]: [0.3570156395435333, 0.8849373459815979]
```

This gave an accuracy of 88.493.

- iv. I applied the same model on the data from test.csv and got an accuracy of 88.447 on kaggle.
- c. With skew reduction:
 - i. First i checked the class distribution of the training data:

```
Class distribution before SMOTE:

Default

0 180524

1 23753
Name: count, dtype: int64

print(skew(train_df['Default']))

2.394081993720339
```

ii. I then applied SMOTE to reduce the skewness of the data:

```
Class distribution after SMOTE: print(skew(y_resampled))

Counter({0: 180524, 1: 180524})

0.0
```

iii. I then did hyperparameter tuning using Bayesian Optimization, with the following tuner values:

```
max_trials = 5, executions_per_trial = 1
```

And the tuner search ran for 8 epochs.

iv. This gave the following best parameters:

```
The optimal number of units in the first densely-connected layer is 448, in the second layer is 224 and the optimal learning rate for the optimizer is 0.0001.
```

v. The model was then fitted on the training data split with epochs = 50.

```
Epoch 1/50
/usr/local/lib/python3.10/dist-packages/keras/src/layers/core/dense.py:87: UserWarning: Do not pass an `input_shape`/`input_dim`
 super().__init__(activity_regularizer=activity_regularizer, **kwargs)
                              18s 2ms/step - accuracy: 0.5330 - loss: 108.6645 - val_accuracy: 0.5133 - val_loss: 63.5835
7221/7221
Epoch 2/50
7221/7221 -
                              15s 2ms/step - accuracy: 0.5507 - loss: 53.2121 - val_accuracy: 0.5420 - val_loss: 41.2264
Epoch 3/50
7221/7221 -
                              15s 2ms/step - accuracy: 0.5583 - loss: 45.6762 - val_accuracy: 0.6370 - val_loss: 16.9277
Epoch 4/50
                               21s 2ms/step - accuracy: 0.5736 - loss: 38.4998 - val_accuracy: 0.6103 - val_loss: 27.8290
Epoch 5/50
                               15s 2ms/step - accuracy: 0.5890 - loss: 28.6887 - val_accuracy: 0.6044 - val_loss: 17.3542
7221/7221 -
Epoch 6/50
7221/7221 -
                               15s 2ms/step - accuracy: 0.5949 - loss: 26.2433 - val_accuracy: 0.6209 - val_loss: 8.5408
Epoch 7/50
                               15s 2ms/step - accuracy: 0.5985 - loss: 24.4179 - val accuracy: 0.6277 - val loss: 11.5755
7221/7221
Epoch 8/50
                               15s 2ms/step - accuracy: 0.6101 - loss: 20.8767 - val accuracy: 0.6205 - val loss: 8.2261
7221/7221 -
Epoch 9/50
                               20s 2ms/step - accuracy: 0.6151 - loss: 20.1680 - val accuracy: 0.6740 - val loss: 8.6744
7221/7221 -
Epoch 10/50
                               20s 2ms/step - accuracy: 0.6213 - loss: 16.7428 - val accuracy: 0.5577 - val loss: 19.1021
7221/7221 -
Epoch 11/50
7221/7221 -
                               21s 2ms/step - accuracy: 0.6284 - loss: 14.1247 - val_accuracy: 0.6680 - val_loss: 9.7753
Epoch 12/50
                               21s 2ms/step - accuracy: 0.6286 - loss: 13.2772 - val_accuracy: 0.6863 - val_loss: 4.2657
7221/7221 -
Epoch 13/50
7221/7221 -
                               21s 2ms/step - accuracy: 0.6355 - loss: 11.8272 - val_accuracy: 0.6802 - val_loss: 7.0885
Epoch 50/50
7221/7221 -
                               15s 2ms/step - accuracy: 0.6958 - loss: 0.9898 - val_accuracy: 0.5698 - val_loss: 1.5891
                              3s 1ms/step - accuracy: 0.5672 - loss: 1.6018
2257/2257 -
[test loss, test accuracy]: [1.5914316177368164, 0.5674560070037842]
```

This gave an accuracy of 56.745.

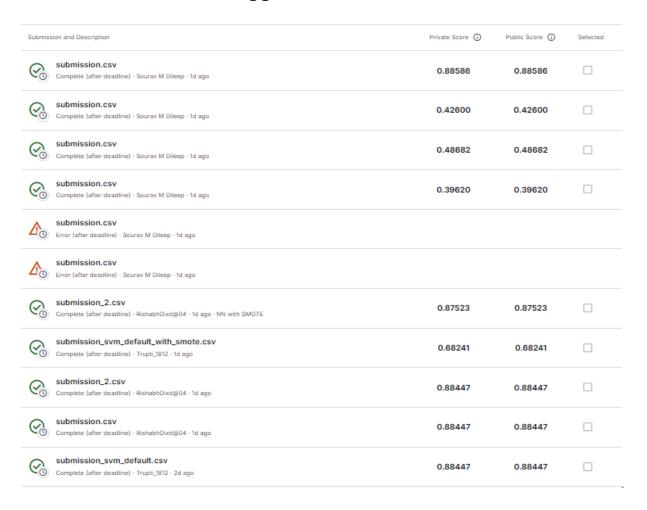
vi. I applied the same model on the data from test.csv and got an accuracy of 87.523 on kaggle.

3. Logistic Regression

- a. Preprocessing was the same as mentioned in the preprocessing section. The model was trained two times, one with skew reduction(SMOTE) and one without skew reduction.
- b. Without using SMOTE:
 - We got a validation accuracy of 88.48 after applying the LR model.
 - ii. After replicating the preprocessing on data of test.csv and applying the model, we got a test accuracy of 88.586.
 - iii. This was the highest accuracy we could achieve using these 3 models.
- c. When using SMOTE:
 - i. We oversampled the data by adding extra rows using the SMOTE function.
 - ii. After the remaining preprocessing steps and application of the LR model, validation accuracy dropped to 42.06.

iii. When this data was submitted on kaggle, the test accuracy dropped to 48.682.

Kaggle Submissions



Required Installations

- 1. pip install imbalance-learn : to remove skewness from data
- 2. !pip install keras-tuner : for hyperparameter tuning of NN model (through Bayesian Optimization)