# Bankruptcy Prediction for American Companies

## Final Data Science Project Report

Dataset Source:

**Kaggle: Bankruptcy Prediction** 

### Objective

Build predictive models to classify whether a public company will go bankrupt in the following year based on financial/accounting ratios.

```
# • Importing Libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import LabelEncoder, RobustScaler
from sklearn.model_selection import train_test_split, GridSearchCV
from sklearn.pipeline import Pipeline
from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score, roc_auc_score, classification_report
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.neighbors import KNeighborsClassifier
from xgboost import XGBClassifier
from imblearn.over_sampling import SMOTE
```

### Data Preprocessing & Balancing

```
# Load dataset
data = pd.read csv('american bankruptcy.csv')
# Label Encoding
le = LabelEncoder()
data['status label'] = le.fit transform(data['status label'])
# Drop unnecessary columns
X = data.drop(columns=['status_label', 'company_name', 'year'])
y = data['status_label']
# Train-Test Split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, stratify=y, random_state=42)
# SMOTE Balancing
smote = SMOTE(random state=42)
X_train_bal, y_train_bal = smote.fit_resample(X_train, y_train)
```

### Model Training & Comparison

```
models = {
    'DecisionTree': DecisionTreeClassifier(),
    'RandomForest': RandomForestClassifier(),
    'GradientBoosting': GradientBoostingClassifier(),
    'KNN': KNeighborsClassifier(),
    'XGBoost': XGBClassifier(use_label_encoder=False, eval_metric='logloss')
}
param_grid = {
    'DecisionTree': {'model__max_depth': [5, 10, 20]},
    'RandomForest': {'model__n_estimators': [100]},
    'GradientBoosting': {'model__n_estimators': [100]},
```

```
'KNN': {'model__n_neighbors': [5, 7]},
    'XGBoost': {'model n estimators': [100]}
}
results = []
best_models = {}
for name in models:
    pipeline = Pipeline([('scaler', RobustScaler()), ('model', models[name])])
    grid = GridSearchCV(pipeline, param_grid[name], cv=3, scoring='f1', n_jobs=-1)
    grid.fit(X_train_bal, y_train_bal)
    best_model = grid.best_estimator_
    best_models[name] = best_model
    y_pred = best_model.predict(X_test)
    y_prob = best_model.predict_proba(X_test)[:, 1]
    results.append({
        'Model': name,
        'Accuracy': accuracy_score(y_test, y_pred),
        'Precision': precision_score(y_test, y_pred),
        'Recall': recall_score(y_test, y_pred),
        'F1-Score': f1_score(y_test, y_pred),
        'ROC AUC': roc_auc_score(y_test, y_prob)
   })
results_df = pd.DataFrame(results).set_index('Model')
results_df
```



/usr/local/lib/python3.11/dist-packages/xgboost/core.py:158: UserWarning: [15:15:14] WARNING: /workspace/src/learner.cc Parameters: { "use\_label\_encoder" } are not used.

warnings.warn(smsg, UserWarning)

	Accuracy	Precision	Recall	F1-Score	ROC AUC
Model					
DecisionTree	0.814069	0.179932	0.506705	0.265562	0.678165
RandomForest	0.917964	0.402832	0.490421	0.442333	0.857835
GradientBoosting	0.689649	0.122790	0.598659	0.203782	0.713521
KNN	0.804283	0.210302	0.707854	0.324265	0.818558
XGBoost	0.814641	0.204107	0.618774	0.306961	0.811582



best\_model = best\_models['KNN']
y\_pred = best\_model.predict(X\_test)
print(classification\_report(y\_test, y\_pred))

<b>→</b>		precision	recall	f1-score	support
	0	0.98	0.81	0.89	14693
	1	0.21	0.71	0.32	1044
	accuracy			0.80	15737
	macro avg	0.59	0.76	0.60	15737
	weighted avg	0.92	0.80	0.85	15737

from IPython.display import display, FileLink
display(FileLink("/content/US Bankruptcy Prediction Model Analysis.pdf"))



/content/US Bankruptcy Prediction Model Analysis.pdf

Start coding or generate with AI.