

Problem:

Client: a Portuguese banking institution

- "Term deposits" a major source of income
- Direct marketing campaigns (phone calls)
- Goal predict if the customer will convert.

Data: Kaggle - Banking Dataset - Marketing Targets

(Train: 45,211 rows; Test: 4,521 rows)

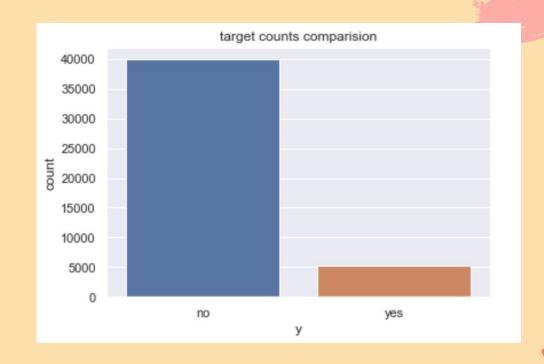


Data Cleaning & EDA

Drop "duration" feature

Link for the reason

Class imbalance: Oversampling













Metric Choosing

F-beta(beta = 2) ROC_AUC Baseline model

Logistic regression

Model Selection

6 model's comparison

Retrain Model

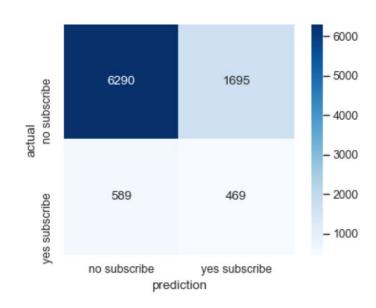
Retrain on (train+val)

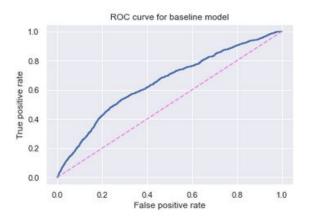
Model evaluation & Interpretation



Baseline Model

Features: numerical features only - age, balance, day, campaign, pdays, previous





F-beta: 0.7582

ROC_AUC: 0.6502



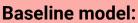
Improving Baseline Model

Baseline Model

- + Scaling
- + Feature Engineering

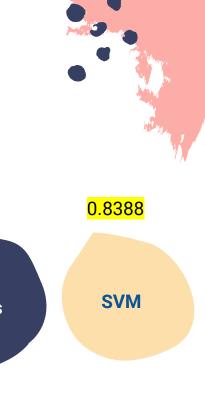
F-beta: 0.7933 ROC_AUC: 0.7687

Candidate Models



F-beta: 0.7582

ROC_AUC: 0.6502



F-beta(beta = 2)

0.7933

0.7632

<mark>0.8333</mark>

<mark>0.8329</mark>

0.8016

Logistic Regression

20-NN

Random Forest

XGBoost

Naive Bayes

0.7687

0.7392

<mark>0.7928</mark>

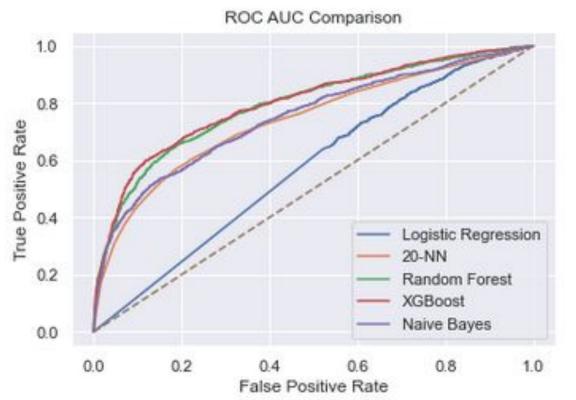
0.7992

0.7487

N/A

ROC_AUC





Winner: XGBoost



Current performance: F-beta: 0.8329 ROC AUC: 0.7992

Step 1: Tuning hyperparameter for XGBoost

Best score on val: F-beta: 0.8536 ROC_AUC: 0.7983

Step 2: Retrain model with (train+val) data

Final Score on Test: F-beta: 0.9946 ROC_AUC: 0.9998

Model Evaluation & Interpretation



Future Work

- 1. Double check code settings
- 2. Check feature importance
- 3. Try different random_state (see the score change for all candidate models)
- 4. Try Random Forest to see if get similar result
- 5. Tuning hyperparameters for SVM to see how good it can be.

Appendix 1: Final Model

```
gbm2 = xgb.XGBClassifier(
                        n estimators=10000.
                        max depth=11,
                        objective='binary:logistic', #new objective
                        learning rate=0.05.
                        subsample=.8,
                        min_child_weight=2,
                        colsample bytree=.8
eval_set=[(X_tr_rs, y_tr_rs),(X_test_dum,y_test)] #using test set
fit model = qbm2.fit(
                    X tr rs, y tr rs,
                    eval set=eval set.
                    eval metric='auc', #new evaluation metric: classification error (could also use AUC, e.g.)
                                    #can "eval metric" be F-beta score
                    early stopping rounds=20.
                    verbose=False
y_pred_test = gbm2.predict(X_test_dum, ntree_limit=gbm2.best_ntree_limit)
y prob pred test = gbm2.predict proba(X test dum)[:,1]
print(f"train f-beta score: {fbeta_score(y_tr_rs,qbm2.predict(X_tr_rs),average='weighted',beta = 2)}")
print(f"train ROC AUC score: {roc_auc_score(y_tr_rs,gbm.predict(X_tr_rs))}")
print(f"test f-beta score: {fbeta_score(y_test,y_pred_test,average='weighted',beta = 2)}")
print(f"test ROC AUC score: {roc auc score(y test, y prob pred test)}")
train f-beta score: 0.9967927081981969
train ROC AUC score: 0.7477706527729072
test f-beta score: 0.9946969122222279
```

test ROC AUC score: 0.9998795585412668