Deep 02. Learning

CONTENTS

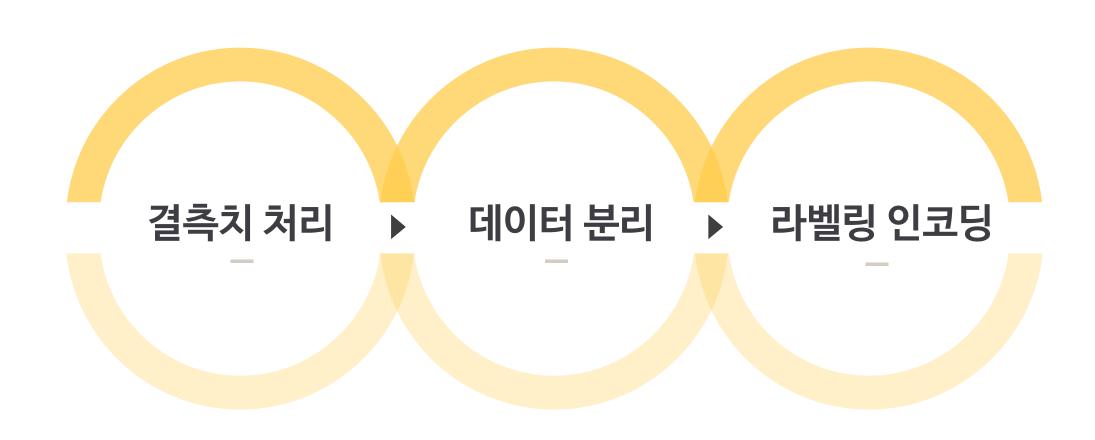
1 데이터 전처리

2 모델링

3 결론

PART1

데이터전처리



● 결측치 처리 및 데이터 분리

결측치 처리

```
data = data.fillna(-1)
data.drop("Unnamed: 0",axis=1,inplace=True)
데이터 분리 및 Index 초기화
from sklearn.model_selection import train_test_split
trn, val = train_test_split(data, test_size=0.2,shuffle=True, stratify=data["class"], random_state=42)
valid = data.iloc[val.index,:]
train = data.iloc[trn.index, :]
train = train.reset_index(drop=True)
valid = valid.reset index(drop=True)
```

● 결측치 처리 및 데이터 분리

결측치 처리

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```
train = train.reset_index(drop=True)
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```

1. 다른 값들과 다른 CLASS로 두기 위해 (Tree 계열 모델)

2. 변수들의 값들의 분포가 넓기 때문에

● 결측치 처리 및 데이터 분리

결측치 처리

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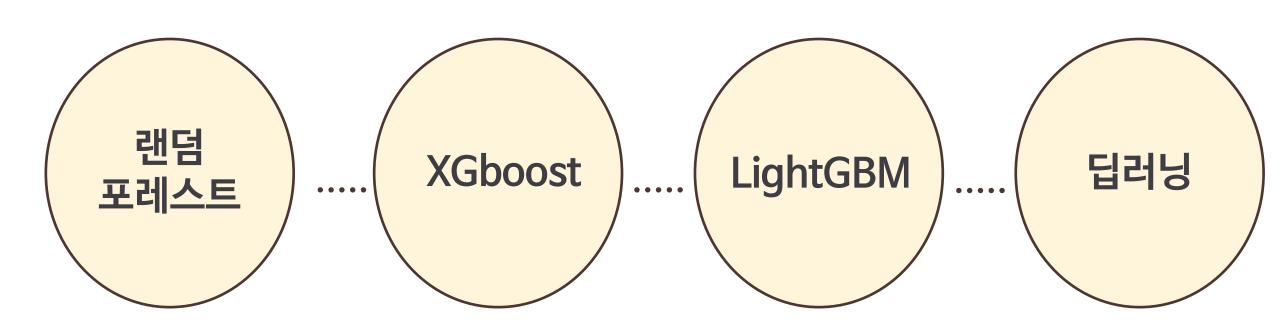
● 레이블 인코딩

Target 변수 encoding

```
3 0
4 0
...
45595 0
45596 0
45597 0
45598 0
45599 0
Name: class, Length: 45600, dtype: int64
```

PART2

모델링



랜덤 포레스트

```
from sklearn.ensemble import RandomForestClassifier
rf = RandomForestClassifier(n_jobs= 2, random_state=42)
rf.fit(x_train,y_train)
rf_pred = rf.predict(x_valid)
rf_pred
y_valid.values
```

```
cost(rf_pred,y_valid.values)
```

27720

XGboost

19190

```
from xgboost import XGBClassifier
 xgb = XGBClassifier(max_depth = 10, learning_rate=0.05,n_estimators=1600,
 xgb.fit(x_train,y_train)
 xgb pred = xgb.predict(x valid)
|XGBClassifier(base score=0.5, booster='gbtree', colsample bylevel=1,
              colsample_bynode=1, colsample_bytree=1, gamma=0, gpu_id=0,
              importance_type='gain', interaction_constraints=''.
              learning_rate=0.05, max_delta_step=0, max_depth=10,
              min_child_weight=1, missing=nan,
n_estimators=1600, n_jobs=0, num_parallel_tree=1, random_state=0,
reg_alpha=0, reg_lambda=1, scale_pos_weight=1, subsample=1,
tree_method='gpu_hist', validate_parameters=1, verbosity=None)
 cost(xgb_pred,y_valid.values)
```

LightGBM

딥러닝

● 정규화

```
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
x_train.iloc[:,:] = sc.fit_transform(x_train.iloc[:,:])
x_valid.iloc[:,:] = sc.transform(x_valid.iloc[:,:])
```

● 딥러닝 모델

딥러닝

k-Fold

● 모델 실행

앙상블

```
final_pred = preds_test*0.25 * Igb_pred_proba*0.15 * xgb_pred_proba*0.6

cost(np.argmax(final_pred,axis=1),y_valid)
19190
```

PART3

결론

3. 결론

결과값

Accuracy Score

Cost

Confusion Matrix

0.9748

155220

 18519
 172

 307
 2

3. 결론

고찰

```
final_pred = xgb_pred_proba2*0.1 + lgb_pred_proba2*0.3 + preds_test2*0.6
```

Thank you