# BookReview-Classification-Supervised

January 20, 2021

```
[]: ! pip install --upgrade scikit-learn
```

# 1 Book Crossing - Classification

```
[1]: %matplotlib inline
   import scipy
   import seaborn as sns
   import matplotlib.pyplot as plt
   import pandas as pd
   import numpy as np
   sns.set()
   palette = sns.color_palette("icefire")
   plt.style.use('ggplot')
   sns.set_context("talk")
[2]: dataset = pd.read_csv('book_crossing.classification.cleaned.csv')
[3]: dataset['age'] = dataset['age'].astype(np.float64)
   dataset['book_rating'] = dataset['book_rating'].astype('category')
   dataset['book_title'] = dataset['book_title'].astype('category')
   dataset['book_author'] = dataset['book_author'].astype('category')
   dataset['year_of_publication'] = dataset['year_of_publication'].astype(np.
     →float64)
   dataset['publisher'] = dataset['publisher'].astype('category')
   dataset['country'] = dataset['country'].astype('category')
[4]: dataset.head()
[4]:
       age book_rating
                                          publisher country
   0 34.0
                   mid
                        ... HarperFlamingo Canada
                                                     canada
   1 30.0
                        ... HarperFlamingo Canada
                   high
                                                     canada
   2 34.0
                        ... HarperFlamingo Canada
                  high
                                                     canada
   3 34.0
                  high
                        ... HarperFlamingo Canada
```

```
4 34.0
                  high ... HarperFlamingo Canada canada
   [5 rows x 7 columns]
[5]: dataset.info()
   <class 'pandas.core.frame.DataFrame'>
   RangeIndex: 364570 entries, 0 to 364569
   Data columns (total 7 columns):
        Column
                             Non-Null Count
                                              Dtype
       _____
                             _____
                                            ____
    0
                            364570 non-null float64
        age
    1
        book_rating
                             364570 non-null category
    2
                            364570 non-null category
       book_title
    3
       book author
                             364570 non-null category
        year_of_publication 364570 non-null float64
    5
       publisher
                             364570 non-null category
        country
                             364570 non-null category
   dtypes: category(5), float64(2)
   memory usage: 19.1 MB
[6]: from sklearn import set_config
   from sklearn.compose import ColumnTransformer
   from sklearn.pipeline import Pipeline
   from sklearn.impute import SimpleImputer
   from sklearn.preprocessing import StandardScaler, OneHotEncoder, LabelEncoder, u
     \rightarrowMinMaxScaler
   from sklearn.svm import SVC, LinearSVC
   from sklearn.model_selection import train_test_split
   from sklearn.metrics import classification_report, plot_confusion_matrix,_
    →confusion_matrix, accuracy_score, balanced_accuracy_score
   from sklearn.neural_network import MLPClassifier
   from sklearn.tree import DecisionTreeClassifier
   from sklearn.ensemble import RandomForestClassifier, AdaBoostClassifier
   from sklearn.discriminant_analysis import QuadraticDiscriminantAnalysis
   from sklearn.linear_model import RidgeClassifier, PassiveAggressiveClassifier
   from sklearn.preprocessing import MinMaxScaler
   from sklearn.preprocessing import minmax_scale
   from sklearn.preprocessing import MaxAbsScaler
   from sklearn.preprocessing import StandardScaler
   from sklearn.preprocessing import RobustScaler
   from sklearn.preprocessing import Normalizer
   from sklearn.preprocessing import QuantileTransformer
   from sklearn.preprocessing import PowerTransformer
```

set\_config(display='diagram')

```
[7]: | X, y = dataset.drop('book_rating', axis=1), dataset['book_rating']
 [8]: target_names = ['low', 'mid', 'high']
 [9]: X.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 364570 entries, 0 to 364569
    Data columns (total 6 columns):
         Column
                              Non-Null Count
                                                Dtype
         _____
                              _____
     0
                              364570 non-null float64
         age
     1
         book_title
                              364570 non-null category
         book_author
                              364570 non-null category
         year_of_publication 364570 non-null float64
     4
         publisher
                              364570 non-null category
                              364570 non-null category
         country
    dtypes: category(4), float64(2)
    memory usage: 18.8 MB
[10]: X.head()
[10]:
         age
                book_title
                                             publisher
                                                        country
       34.0 Clara Callan
                           ... HarperFlamingo Canada
                                                         canada
     1 30.0 Clara Callan ... HarperFlamingo Canada
                                                         canada
     2 34.0 Clara Callan
                           ... HarperFlamingo Canada
                                                         canada
     3 34.0 Clara Callan
                                 HarperFlamingo Canada
                                                         canada
     4 34.0 Clara Callan
                                 HarperFlamingo Canada
                                                         canada
     [5 rows x 6 columns]
[11]: y.head()
[11]: 0
          mid
     1
         high
     2
         high
     3
         high
         high
     Name: book_rating, dtype: category
     Categories (3, object): ['high', 'low', 'mid']
[12]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,_
      →random state=0)
[13]: numeric_features = ['age', 'year_of_publication']
     numeric transformer = Pipeline(steps=[
         ('imputer', SimpleImputer(strategy='median')),
         ('scaler', StandardScaler())])
[14]: categorical_features = ['book_title', 'book_author', 'publisher', 'country']
     categorical_transformer = OneHotEncoder(handle_unknown='ignore')
```

[24]: (291656, 177473)

If we were to store all the all data, and each cell occupying 4 bytes, this is how much space we would require

```
[30]: f'{(dummy.shape[0] * dummy.shape[1] * 4) // 1024 // 1024 // 1024} giga bytes'
[30]: '192 giga bytes'
```

# 1.1 Supervised Models

- DecisionTreeClassifier
- SVC
- LinearSVC
- KNeighborsClassifier
- PassiveAggressiveClassifier
- AdaBoostClassifier
- GaussianProcessClassifier

```
# get the classification report
  print(f"\nClassification Report for {algorithm.__class__.__name__})")
  print(classification_report(y_true, y_pred, target_names=target_names,_
→labels=target_names))
  acc_score = accuracy_score(y_true, y_pred)
  bal_score = balanced_accuracy_score(y_true, y_pred)
  print(f"\nAccuracy Score: {acc_score}")
  print(f"Balanced Accuracy Score: {bal_score}")
  print()
  # show the confusion matrix
  cmmat_table = pd.DataFrame({'y_true': y_true, 'y_pred': y_pred})
  conmat = pd.crosstab(cmmat_table.y_true, cmmat_table.y_pred,_
→rownames=['Actual'], colnames=['Predicted'], margins=True, normalize='all')
  ax = plt.axes()
  sns.set(rc={'figure.figsize':(9, 7)})
  sns.heatmap(conmat, annot=True, ax=ax)
  ax.set_title(f'{algorithm.__class__.__name__}')
  plt.show()
  print()
  t2 = time()
  print(f'Trained {algorithm.__class__.__name__} in {(t2 - t1)}s')
  return clf
```

## 1.2 DecisionTree Classifier

```
[33]: dtc = DecisionTreeClassifier(max_depth=100)

[34]: clf = fit_model(algorithm=dtc, data=(X_train, X_test, y_train, y_test),

→preprocessor=preprocessor)

Started Training DecisionTreeClassifier on X_train: (291656, 6) y_train: (291656,)

Evaluating model on X_test: (72914, 6) y_test: (72914,)

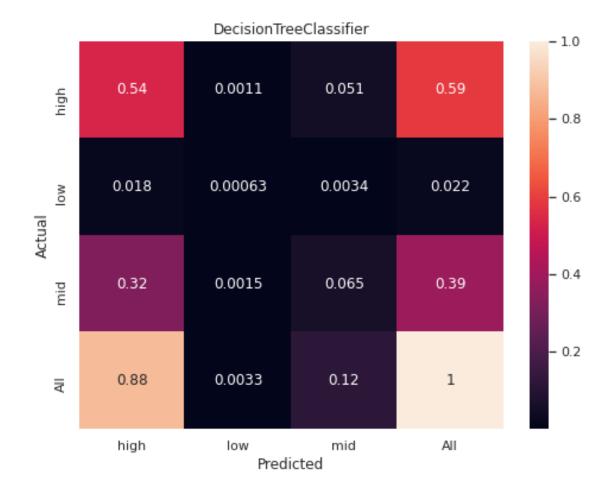
Classification Report for DecisionTreeClassifier

precision recall f1-score support
```

low	0.19	0.03	0.05	1603
mid	0.55	0.17	0.25	28483
high	0.61	0.91	0.73	42828
accuracy			0.60	72914
macro avg	0.45	0.37	0.35	72914
weighted avg	0.58	0.60	0.53	72914

Accuracy Score: 0.6010368379186439

Balanced Accuracy Score: 0.36884514138535285



Trained DecisionTreeClassifier in 146.20573687553406s

[35]: clf

```
[35]: Pipeline(steps=[('preprocessor',
                      ColumnTransformer(transformers=[('num',
                                                        Pipeline(steps=[('imputer',
     SimpleImputer(strategy='median')),
                                                                         ('scaler',
     StandardScaler())]),
                                                        ['age',
                                                          'year_of_publication']),
                                                       ('cat',
     OneHotEncoder(handle_unknown='ignore'),
                                                        ['book_title', 'book_author',
                                                          'publisher', 'country'])])),
                     ('classifier', DecisionTreeClassifier(max_depth=100))])
    1.3 SVC (Support Vector Classification)
[31]: svc = SVC(gamma='auto', C=1.0, kernel='rbf')
 | ]: clf = fit_model(algorithm=svc, data=(X_train, X_test, y_train, y_test),__
      →preprocessor=preprocessor)
    Started Training SVC on X_train: (291656, 6) y_train: (291656,)
 []: clf
```

# 1.4 Passive Aggressive Classifier

```
pac = PassiveAggressiveClassifier(early_stopping=True, verbose=1, n_jobs=-1)

[20]: clf = fit_model(algorithm=pac, data=(X_train, X_test, y_train, y_test), ______
__preprocessor=preprocessor)

Started Training PassiveAggressiveClassifier on X_train: (291656, 6) y_train: (291656,)

[Parallel(n_jobs=-1)]: Using backend ThreadingBackend with 2 concurrent workers.

-- Epoch 1-- Epoch 1

Norm: 46.21, NNZs: 53157, Bias: -0.504273, T: 262490, Avg. loss: 0.105049

Total training time: 0.11 seconds.

-- Epoch 2

Norm: 119.01, NNZs: 149844, Bias: 0.084396, T: 262490, Avg. loss: 0.905595

Total training time: 0.18 seconds.

-- Epoch 2
```

Norm: 66.08, NNZs: 70363, Bias: -0.521754, T: 524980, Avg. loss: 0.086252 Total training time: 0.29 seconds.

### -- Epoch 3

Norm: 164.90, NNZs: 161910, Bias: 0.072078, T: 524980, Avg. loss: 0.789480 Total training time: 0.36 seconds.

### -- Epoch 3

Norm: 79.79, NNZs: 79611, Bias: -0.583871, T: 787470, Avg. loss: 0.077301 Total training time: 0.46 seconds.

### -- Epoch 4

Norm: 198.89, NNZs: 164254, Bias: 0.097328, T: 787470, Avg. loss: 0.731145 Total training time: 0.54 seconds.

### -- Epoch 4

Norm: 90.21, NNZs: 85297, Bias: -0.646251, T: 1049960, Avg. loss: 0.072339 Total training time: 0.63 seconds.

### -- Epoch 5

Norm: 225.85, NNZs: 164894, Bias: 0.119736, T: 1049960, Avg. loss: 0.696331 Total training time: 0.72 seconds.

#### -- Epoch 5

Norm: 98.61, NNZs: 89244, Bias: -0.634484, T: 1312450, Avg. loss: 0.070211 Total training time: 0.80 seconds.

### -- Epoch 6

Norm: 248.75, NNZs: 165157, Bias: 0.069723, T: 1312450, Avg. loss: 0.674325 Total training time: 0.89 seconds.

### -- Epoch 6

Norm: 105.81, NNZs: 92113, Bias: -0.705083, T: 1574940, Avg. loss: 0.067957 Total training time: 0.99 seconds.

Convergence after 6 epochs took 1.04 seconds

### -- Epoch 1

Norm: 268.11, NNZs: 165276, Bias: 0.086452, T: 1574940, Avg. loss: 0.659386 Total training time: 1.10 seconds.

Convergence after 6 epochs took 1.16 seconds

Norm: 117.95, NNZs: 148735, Bias: -0.110863, T: 262490, Avg. loss: 0.897951 Total training time: 0.15 seconds.

## -- Epoch 2

Norm: 163.59, NNZs: 161578, Bias: -0.108644, T: 524980, Avg. loss: 0.784290 Total training time: 0.30 seconds.

## -- Epoch 3

Norm: 197.51, NNZs: 164105, Bias: -0.107356, T: 787470, Avg. loss: 0.725686 Total training time: 0.45 seconds.

### -- Epoch 4

Norm: 224.64, NNZs: 164845, Bias: -0.101451, T: 1049960, Avg. loss: 0.692328 Total training time: 0.59 seconds.

#### -- Epoch 5

Norm: 247.11, NNZs: 165150, Bias: -0.105241, T: 1312450, Avg. loss: 0.670131 Total training time: 0.73 seconds.

### -- Epoch 6

Norm: 266.48, NNZs: 165274, Bias: -0.121605, T: 1574940, Avg. loss: 0.654409 Total training time: 0.89 seconds.

## -- Epoch 7

 ${\tt Norm:\ 283.55,\ NNZs:\ 165349,\ Bias:\ -0.141001,\ T:\ 1837430,\ Avg.\ loss:\ 0.642702}$ 

Total training time: 1.03 seconds.

-- Epoch 8

Norm: 298.48, NNZs: 165405, Bias: -0.106583, T: 2099920, Avg. loss: 0.634828

Total training time: 1.18 seconds.

-- Epoch 9

Norm: 312.23, NNZs: 165423, Bias: -0.145596, T: 2362410, Avg. loss: 0.626431

Total training time: 1.33 seconds.

Convergence after 9 epochs took 1.38 seconds

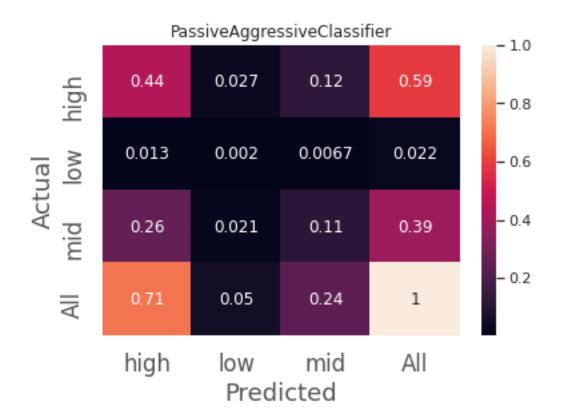
[Parallel(n\_jobs=-1)]: Done 3 out of 3 | elapsed: 2.5s finished

Evaluating model on X\_test: (72914, 6) y\_test: (72914,)

# ${\tt Classification\ Report\ for\ PassiveAggressiveClassifier}$

	precision	recall	f1-score	support
low	0.04	0.09	0.05	1603
mid	0.47	0.29	0.36	28483
high	0.62	0.75	0.68	42828
accuracy			0.55	72914
macro avg	0.38	0.37	0.36	72914
weighted avg	0.55	0.55	0.54	72914

Accuracy Score: 0.5531722302987081



Trained PassiveAggressiveClassifier in 8.378920316696167s

```
[21]: clf
[21]: Pipeline(steps=[('preprocessor',
                      ColumnTransformer(transformers=[('num',
                                                        Pipeline(steps=[('imputer',
     SimpleImputer(strategy='median')),
                                                                         ('scaler',
     StandardScaler())]),
                                                        ['age',
                                                          'year_of_publication']),
                                                       ('cat',
     OneHotEncoder(handle_unknown='ignore'),
                                                        ['book_title', 'book_author',
                                                         'publisher', 'country'])])),
                     ('classifier',
                      PassiveAggressiveClassifier(early_stopping=True, n_jobs=-1,
                                                   verbose=1))])
```

# 1.5 Ridge Classifier

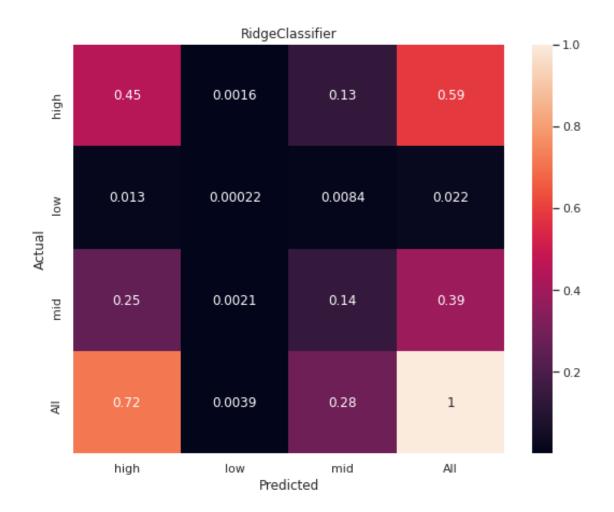
Started Training RidgeClassifier on X\_train: (291656, 6) y\_train: (291656,)

Evaluating model on X\_test: (72914, 6) y\_test: (72914,)

Classification Report for RidgeClassifier

	precision	recall	f1-score	support
low	0.06	0.01	0.02	1603
mid	0.49	0.35	0.41	28483
high	0.63	0.77	0.69	42828
accuracy			0.59	72914
macro avg	0.39	0.38	0.37	72914
weighted avg	0.56	0.59	0.57	72914

Accuracy Score: 0.5888855363853307



# Trained RidgeClassifier in 22.259966611862183s

# 1.6 AdaBoost Classifier

```
[29]: abc = AdaBoostClassifier()
```

[30]: clf = fit\_model(algorithm=abc, data=(X\_train, X\_test, y\_train, y\_test), \_\_\_\_\_
preprocessor=preprocessor)

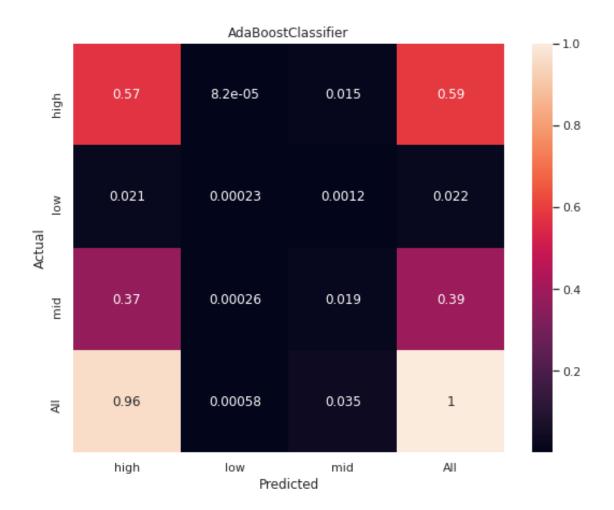
Started Training AdaBoostClassifier on X\_train: (291656, 6) y\_train: (291656,)

Evaluating model on X\_test: (72914, 6) y\_test: (72914,)

Classification Report for AdaBoostClassifier

	precision	recall	f1-score	support
low mid high	0.40 0.55 0.59	0.01 0.05 0.97	0.02 0.09 0.74	1603 28483 42828
accuracy macro avg weighted avg	0.51 0.57	0.34 0.59	0.59 0.28 0.47	72914 72914 72914

Accuracy Score: 0.5919165043750172



# Trained AdaBoostClassifier in 50.97685408592224s

# 1.7 Linear SVC

```
[28]: | lsvc = LinearSVC(verbose=1)
```

[29]: clf = fit\_model(algorithm=lsvc, data=(X\_train, X\_test, y\_train, y\_test), 
→preprocessor=preprocessor)

Started Training LinearSVC on X\_train: (291656, 6) y\_train: (291656,)

/usr/local/lib/python3.6/dist-packages/sklearn/svm/\_base.py:986: ConvergenceWarning: Liblinear failed to converge, increase the number of iterations.

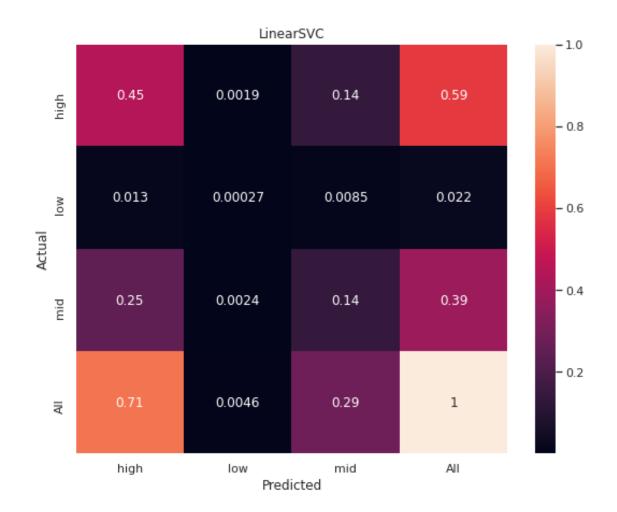
"the number of iterations.", ConvergenceWarning)

Evaluating model on X\_test: (72914, 6) y\_test: (72914,)

Classification Report for LinearSVC

	precision	recall	f1-score	support
low	0.06	0.01	0.02	1603
mid	0.49	0.35	0.41	28483
high	0.63	0.76	0.69	42828
accuracy			0.59	72914
macro avg	0.39	0.38	0.37	72914
weighted avg	0.56	0.59	0.57	72914

Accuracy Score: 0.5862660120141536



# Trained LinearSVC in 199.0021414756775s

# 1.8 MLP Classifier

```
[26]: mlp = MLPClassifier(alpha=0.001, max_iter=1, verbose=1)
```

[27]: clf = fit\_model(algorithm=mlp, data=(X\_train, X\_test, y\_train, y\_test),\_\_
preprocessor=preprocessor)

Started Training MLPClassifier on X\_train: (291656, 6) y\_train: (291656,) Iteration 1, loss = 0.75835927

Evaluating model on X\_test: (72914, 6) y\_test: (72914,)

/usr/local/lib/python3.6/dist-

packages/sklearn/neural\_network/\_multilayer\_perceptron.py:617:

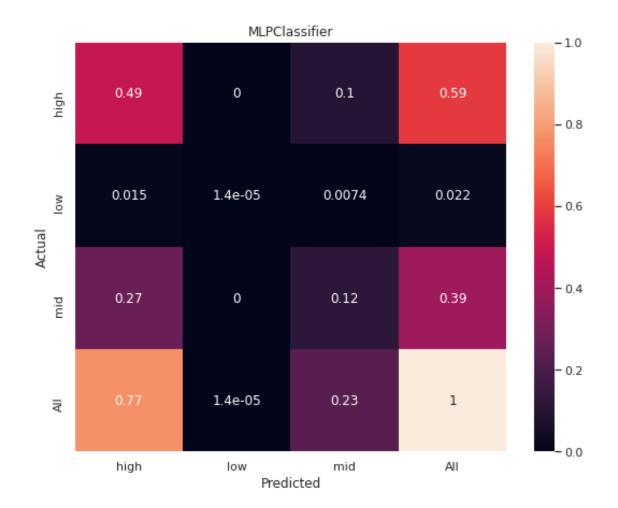
ConvergenceWarning: Stochastic Optimizer: Maximum iterations (1) reached and the optimization hasn't converged yet.

% self.max\_iter, ConvergenceWarning)

Classification Report for MLPClassifier

	precision	recall	f1-score	support
low	1.00	0.00	0.00	1603
mid	0.53	0.31	0.39	28483
high	0.63	0.83	0.72	42828
accuracy			0.61	72914
macro avg	0.72	0.38	0.37	72914
weighted avg	0.60	0.61	0.57	72914

Accuracy Score: 0.6084153934772472



# Trained MLPClassifier in 630.3555474281311s

## 1.9 RandomForest Classifier

```
[23]: rfc = RandomForestClassifier(max_depth=50, verbose=1, n_jobs=-1)

[24]: clf = fit_model(algorithm=rfc, data=(X_train, X_test, y_train, y_test), □

preprocessor=preprocessor)
```

Started Training RandomForestClassifier on X\_train: (291656, 6) y\_train: (291656,)

[Parallel(n\_jobs=-1)]: Using backend ThreadingBackend with 2 concurrent workers.

[Parallel(n\_jobs=-1)]: Done 46 tasks | elapsed: 4.1min

[Parallel(n\_jobs=-1)]: Done 100 out of 100 | elapsed: 8.8min finished

Evaluating model on X\_test: (72914, 6) y\_test: (72914,)

 $[Parallel(n\_jobs=2)]: \ Using \ backend \ Threading Backend \ with \ 2 \ concurrent \ workers.$ 

[Parallel(n\_jobs=2)]: Done 46 tasks | elapsed: 0.6s

[Parallel(n\_jobs=2)]: Done 100 out of 100 | elapsed: 1.2s finished

Classification Report for RandomForestClassifier

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/\_classification.py:1245: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

\_warn\_prf(average, modifier, msg\_start, len(result))

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/\_classification.py:1245: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

\_warn\_prf(average, modifier, msg\_start, len(result))

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/\_classification.py:1245: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

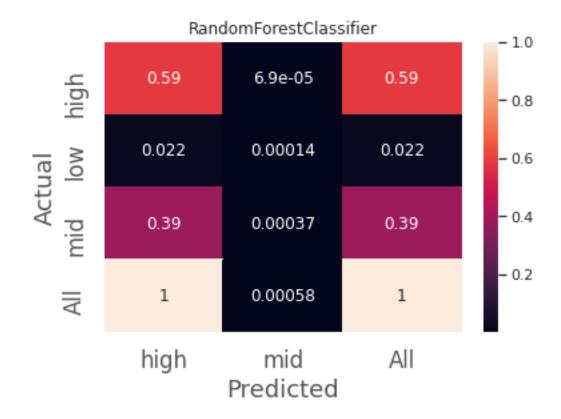
\_warn\_prf(average, modifier, msg\_start, len(result))

	precision	recall	f1-score	support
low	0.00	0.00	0.00	1603
mid	0.64	0.00	0.00	28483
high	0.59	1.00	0.74	42828

accuracy			0.59	72914
macro avg	0.41	0.33	0.25	72914
weighted avg	0.60	0.59	0.44	72914

Accuracy Score: 0.5876786351043696

Balanced Accuracy Score: 0.33361039593376624



Trained RandomForestClassifier in 537.0094470977783s

# 1.10 Comparing Scalers

- StandardScaler
- MinMaxScaler
- RobustScaler
- MaxAbsScaler
- PowerTransformer
- OuantileTransformer
- Normalizer

```
[43]: from time import time
     def fit_model_unscaled(algorithm, data):
         categorical_transformer = OneHotEncoder(handle_unknown='ignore')
         preprocessor = ColumnTransformer(
             transformers=[
                 ('cat', categorical_transformer, categorical_features)
         )
         t1 = time()
         X_train, X_test, y_train, y_test = data
         clf = Pipeline(steps=[('preprocessor', preprocessor),
                               ('classifier', algorithm)])
         # train the model
         clf.fit(X_train, y_train)
         # test the model
         y_true = y_test.copy()
         y_pred = clf.predict(X_test)
         # get the classification report
         print(f"\nClassification Report for {algorithm._class_._name_}\_

→Unscaled")
```

```
print(classification_report(y_true, y_pred, target_names=target_names,_
      →labels=target_names))
         acc_score = accuracy_score(y_true, y_pred)
         bal_score = balanced_accuracy_score(y_true, y_pred)
         print(f"\nAccuracy Score: {acc_score}")
         print(f"Balanced Accuracy Score: {bal_score}")
         print()
        t2 = time()
         print(f'Trained {algorithm.__class__.__name__} in {(t2 - t1)}s')
         return clf
[25]: from time import time
     def fit_model_scaler(algorithm, data, scaler):
         numeric_transformer = Pipeline(
             steps=[
                 ('imputer', SimpleImputer(strategy='median')),
                 ('scaler', scaler)]
             )
         categorical_transformer = OneHotEncoder(handle_unknown='ignore')
         preprocessor = ColumnTransformer(
             transformers=[
                 ('num', numeric_transformer, numeric_features),
                 ('cat', categorical_transformer, categorical_features)
         )
         t1 = time()
         X_train, X_test, y_train, y_test = data
         clf = Pipeline(steps=[('preprocessor', preprocessor),
                               ('classifier', algorithm)])
         # train the model
         clf.fit(X_train, y_train)
```

```
# test the model
         y_true = y_test.copy()
         y_pred = clf.predict(X_test)
         # get the classification report
         print(f"\nClassification Report for {algorithm.__class__.__name__} with⊔
      →{scaler.__class__.__name__}")
         print(classification_report(y_true, y_pred, target_names=target_names,_
      →labels=target_names))
         acc_score = accuracy_score(y_true, y_pred)
         bal_score = balanced_accuracy_score(y_true, y_pred)
         print(f"\nAccuracy Score: {acc_score}")
         print(f"Balanced Accuracy Score: {bal_score}")
         print()
         t2 = time()
         print(f'Trained {algorithm.__class__.__name__} in {(t2 - t1)}s')
         return clf
[26]: algo = PassiveAggressiveClassifier(early_stopping=True, n_jobs=-1)
```

## 1.10.1 Standard Scaler

Classification Report for PassiveAggressiveClassifier with StandardScaler precision recall f1-score support

low	0.06	0.06	0.06	1603
mid	0.46	0.36	0.40	28483
high	0.62	0.72	0.67	42828
accuracy			0.56	72914
macro avg	0.38	0.38	0.38	72914
weighted avg	0.55	0.56	0.55	72914

Accuracy Score: 0.5615656801163014

Trained PassiveAggressiveClassifier in 7.8669610023498535s

## 1.10.2 MinMax Scaler

Classification Report for PassiveAggressiveClassifier with MinMaxScaler precision recall f1-score support

	P-00-0-0-1			Dupper o	
low	0.05	0.00	0.01	1603	
mid	0.44	0.55	0.49	28483	
high	0.64	0.55	0.59	42828	
accuracy			0.54	72914	
macro avg	0.37	0.37	0.36	72914	
weighted avg	0.55	0.54	0.54	72914	

Accuracy Score: 0.5409249252544093

Balanced Accuracy Score: 0.3694473636392875

Trained PassiveAggressiveClassifier in 7.548543214797974s

### 1.10.3 MaxAbs Scaler

Classification Report for PassiveAggressiveClassifier with MaxAbsScaler precision recall f1-score support

	precision	recarr	II BCOLE	Support
low	0.07	0.01	0.02	1603
mid	0.49	0.19	0.27	28483
high	0.61	0.87	0.72	42828
accuracy			0.59	72914
macro avg	0.39	0.36	0.34	72914
weighted avg	0.55	0.59	0.53	72914

Accuracy Score: 0.5882820857448501

Balanced Accuracy Score: 0.3598257153815126

Trained PassiveAggressiveClassifier in 7.605406761169434s

```
'publisher', 'country'])])), ('classifier',
PassiveAggressiveClassifier(early_stopping=True, n_jobs=-1))])
```

#### 1.10.4 Robust Scaler

```
[33]: robust_scaler = RobustScaler(quantile_range=(25, 75))

[34]: fit_model_scaler(algo, data=(X_train, X_test, y_train, y_test),__

scaler=robust_scaler)
```

Classification Report for PassiveAggressiveClassifier with RobustScaler precision recall f1-score support

1				11
low	0.08	0.03	0.05	1603
mid	0.47	0.26	0.33	28483
high	0.61	0.81	0.70	42828
accuracy			0.58	72914
macro avg	0.39	0.37	0.36	72914
weighted avg	0.54	0.58	0.54	72914

Accuracy Score: 0.5769811010231232

Balanced Accuracy Score: 0.36651556542661007

Trained PassiveAggressiveClassifier in 7.649526119232178s

# 1.10.5 PowerTransform (Yeo-Johnson)

```
[35]: power_trans = PowerTransformer(method='yeo-johnson')
[36]: fit_model_scaler(algo, data=(X_train, X_test, y_train, y_test),
      →scaler=power_trans)
    /usr/local/lib/python3.6/dist-packages/numpy/core/ methods.py:205:
    RuntimeWarning: overflow encountered in multiply
      x = um.multiply(x, x, out=x)
    /usr/local/lib/python3.6/dist-packages/numpy/core/_methods.py:216:
    RuntimeWarning: overflow encountered in reduce
      ret = umr_sum(x, axis, dtype, out, keepdims)
    Classification Report for PassiveAggressiveClassifier with PowerTransformer
                               recall f1-score
                  precision
                                                   support
             low
                       0.06
                                 0.04
                                            0.05
                                                      1603
                       0.43
                                 0.58
                                            0.50
                                                     28483
             mid
            high
                       0.64
                                 0.51
                                            0.57
                                                     42828
                                            0.53
                                                     72914
        accuracy
                       0.38
                                 0.37
                                            0.37
                                                     72914
       macro avg
                                                     72914
    weighted avg
                       0.55
                                 0.53
                                            0.53
    Accuracy Score: 0.5252214938146309
    Balanced Accuracy Score: 0.37488500156296717
    Trained PassiveAggressiveClassifier in 10.008253574371338s
[36]: Pipeline(steps=[('preprocessor',
                      ColumnTransformer(transformers=[('num',
                                                        Pipeline(steps=[('imputer',
     SimpleImputer(strategy='median')),
                                                                        ('scaler',
     PowerTransformer())]),
                                                        ['age',
                                                         'year_of_publication']),
                                                       ('cat',
     OneHotEncoder(handle_unknown='ignore'),
                                                        ['book_title', 'book_author',
                                                         'publisher', 'country'])])),
                     ('classifier',
                      PassiveAggressiveClassifier(early_stopping=True, n_jobs=-1))])
```

## 1.10.6 Quantile Transform

```
[37]: quant_trans = QuantileTransformer(output_distribution='normal')

[38]: fit_model_scaler(algo, data=(X_train, X_test, y_train, y_test),___

scaler=quant_trans)
```

```
{\tt Classification}\ {\tt Report}\ {\tt for}\ {\tt PassiveAggressiveClassifier}\ {\tt with}\ {\tt QuantileTransformer}
                 precision
                                 recall f1-score
                                                         support
           low
                       0.08
                                    0.03
                                                0.05
                                                             1603
           mid
                       0.43
                                    0.59
                                                0.50
                                                            28483
          high
                       0.64
                                    0.49
                                                0.56
                                                            42828
```

accuracy 0.52 72914 macro avg 0.38 0.37 0.37 72914 weighted avg 0.54 0.52 0.52 72914

Accuracy Score: 0.5197492936884549

Balanced Accuracy Score: 0.37149925184802246

Trained PassiveAggressiveClassifier in 8.048646688461304s

### 1.10.7 Normalizer (L2)

```
[39]: norm_scaler = Normalizer()

[40]: fit_model_scaler(algo, data=(X_train, X_test, y_train, y_test),__

scaler=norm_scaler)
```

Classification Report for PassiveAggressiveClassifier with Normalizer precision recall f1-score support

	precision	recarr	11 20016	Support	
low	0.05	0.03	0.04	1603	
mid	0.46	0.30	0.36	28483	
high	0.62	0.77	0.69	42828	
accuracy			0.57	72914	
macro avg	0.38	0.37	0.36	72914	
weighted avg	0.54	0.57	0.55	72914	

Accuracy Score: 0.5706722988726445

Balanced Accuracy Score: 0.36646952962846085

Trained PassiveAggressiveClassifier in 8.04029631614685s

# 1.10.8 Unscaled

```
[44]: fit_model_unscaled(algo, data=(X_train, X_test, y_train, y_test))
```

Classification Report for PassiveAggressiveClassifier Unscaled precision recall f1-score support

low	0.09	0.02	0.03	1603
mid	0.48	0.24	0.32	28483
high	0.61	0.83	0.70	42828
accuracy			0.58	72914
macro avg	0.39	0.36	0.35	72914

```
weighted avg
                       0.55
                                 0.58
                                           0.54
                                                     72914
    Accuracy Score: 0.5831938996626163
    Balanced Accuracy Score: 0.36462720452376884
    Trained PassiveAggressiveClassifier in 7.465895652770996s
[44]: Pipeline(steps=[('preprocessor',
                      ColumnTransformer(transformers=[('cat',
     OneHotEncoder(handle_unknown='ignore'),
                                                        ['book_title', 'book_author',
                                                         'publisher', 'country'])])),
                     ('classifier',
                      PassiveAggressiveClassifier(early_stopping=True, n_jobs=-1))])
 []:
```