### **Exercise - Ensemble**

In this exercise, we will focus on underage drinking. The data set contains data about high school students. Each row represents a single student. The columns include the characteristics of deidentified students. This is a binary classification task: predict whether a student drinks alcohol or not (this is the **alc** column: 1=Yes, 0=No). This is an important prediction task to detect underage drinking and deploy intervention techniques.

## **Description of Variables**

The description of variables are provided in "Alcohol - Data Dictionary.docx"

#### Goal

Use the alcohol.csv data set and build a model to predict alc.

## **Read and Prepare the Data**

```
In [188]: ▶ # Common imports
    import pandas as pd
    import numpy as np
    np.random.seed(42)
```

### Get the data

#### Out[189]:

	age	Medu	Fedu	traveltime	studytime	failures	famrel	freetime	goout	health	absences
0	18	2	1	4	2	0	5	4	2	5	2
1	18	4	3	1	0	0	4	4	2	3	9
2	15	4	3	2	3	0	5	3	4	5	0
3	15	3	3	1	4	0	4	3	3	3	10
4	17	3	2	1	2	0	5	3	5	5	2
4											<b>&gt;</b>

# Split data (train/test)

# **Data Prep**

### Separate the target variable

## Feature Engineering: Derive a new column

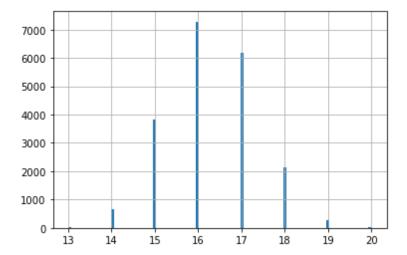
#### Examples:

- · Ratio of study time to travel time
- Student is younger than 18 or not
- Average of father's and mother's level of education
- (etc.)

```
In [217]:
           | train_inputs['age'].describe()
   Out[217]: count
                       20400.000000
              mean
                           16.301765
              std
                            1.047393
                          13.000000
              min
              25%
                          16.000000
              50%
                           16.000000
              75%
                           17.000000
              max
                           20.000000
              Name: age, dtype: float64
```

#### 

#### Out[218]: <AxesSubplot:>



```
In [219]: | def new_col(df):
    #Create a copy so that we don't overwrite the existing dataframe
    df1 = df.copy()

# Use the formula, though fill in 0s when the value is 0/0 (because 0/0 ge
    df1['ave'] = ((df1['Medu'] + df1['Fedu'])/2).fillna(0)

# Replace the infinity values with 1 (because a value divided by 0 general
    df1['ave'].replace(np.inf, 1, inplace=True)

return df1[['ave']]

# You can use this to check whether the calculation is made correctly:
    #return df1
```

```
In [220]:
             ▶ new col(train set)
    Out[220]:
                        ave
                 12759
                        4.5
                  4374
                        1.5
                  8561
                        3.5
                 10697
                        4.0
                 19424
                        5.0
                         ...
                 16850
                        1.5
                  6265
                        2.0
                 11284 0.0
                   860
                        2.0
                 15795 1.5
                23800 rows × 1 columns
```

## Identify the numeric, binary, and categorical columns

```
In [221]:
             # Identify the numerical columns
             numeric_columns = train_inputs.select_dtypes(include=[np.number]).columns.to_
             # Identify the categorical columns
             categorical_columns = train_inputs.select_dtypes('object').columns.to_list()
In [222]:
             numeric_columns
   Out[222]: ['age',
               'Medu',
               'Fedu',
               'traveltime',
               'studytime',
               'failures',
               'famrel',
               'freetime',
               'goout',
               'health',
               'absences']
           In [223]:
   Out[223]: ['gender']
```

## **Pipeline**

```
In [225]:
        numeric_transformer = Pipeline(steps=[
                         ('imputer', SimpleImputer(strategy='median')),
                         ('scaler', StandardScaler())])
In [226]:
         ('imputer', SimpleImputer(strategy='constant', fill_value=0)),
               ('onehot', OneHotEncoder(handle unknown='ignore'))])
my_new = Pipeline(steps=[('my_new', FunctionTransformer(new_col)),
                                      ('scaler', StandardScaler())])
         preprocessor = ColumnTransformer([
In [231]:
                   ('num', numeric transformer, numeric columns),
                   ('cat', categorical_transformer, categorical_columns),
                   ('trans', my_new, feat_eng_columns)],
                   remainder='drop')
            #passtrough is an optional step. You don't have to use it.
```

# Transform: fit\_transform() for TRAIN

```
In [232]:
           #Fit and transform the train data
              train_x = preprocessor.fit_transform(train_inputs)
             train x
   Out[232]: array([[-0.28811749, -0.30630919, -0.82135848, ...,
                                , -0.56468129],
                     [ 0.66665783, -0.93973112, 0.90150543, ...,
                               , -0.17173893],
                     [-1.24289281, 0.32711275, 0.90150543, ...,
                               , 0.6141458 ],
                       0.
                     [ 0.66665783, -2.20657499, -2.54422238, ...,
                                , -2.52939313],
                     [ 1.62143314, -0.30630919, -1.68279043, ...,
                                , -0.95762366],
                     [ 1.62143314, -0.30630919, -2.54422238, ..., 0.
                                , -1.35056603]])
```

```
In [233]: ▶ train_x.shape
Out[233]: (20400, 14)
```

# **Tranform: transform() for TEST**

```
# Transform the test data
In [234]:
             test x = preprocessor.transform(test inputs)
             test_x
   Out[234]: array([[-1.24289281, 0.32711275, 1.76293738, ...,
                                                             1.
                              , 1.00708817],
                   [-1.24289281, -0.30630919, 0.04007348, ...,
                             , -0.17173893],
                   [-0.28811749, 0.32711275, 0.04007348, ...,
                             , 0.22120344],
                   [-0.28811749, -1.57315306, 0.04007348, ...,
                              , -0.95762366],
                   [0.66665783, -2.20657499, -1.68279043, ..., 0.
                              , -2.13645076],
                   [1.62143314, -0.30630919, -1.68279043, ..., 0.
                              , -0.95762366]])
Out[235]: (13600, 14)
```

### Calculate the Baseline

# Train a voting classifier

★ from sklearn.metrics import accuracy\_score

```
from sklearn.tree import DecisionTreeClassifier
In [240]:
              from sklearn.linear model import SGDClassifier
              from sklearn.linear_model import LogisticRegression
              from sklearn.ensemble import VotingClassifier
              dtree_clf = DecisionTreeClassifier(max_depth=30)
              log clf = LogisticRegression(multi class='multinomial', solver = 'lbfgs', C=1
              sgd_clf = SGDClassifier(max_iter=10000, tol=1e-3)
              voting clf = VotingClassifier(
                          estimators=[('dt', dtree_clf),
                                       ('lr', log_clf),
                                       ('sgd', sgd_clf)],
                          voting='hard')
              voting_clf.fit(train_x, train_y)
   Out[240]:
                                        VotingClassifier
                                                   1r
                                                                      sgd
                                                                 SGDClassifier
                ▶ DecisionTreeClassifier
                                          ▶ LogisticRegression
```

In [241]:

```
In [242]:
           ▶ #Train accuracy
              train_y_pred = voting_clf.predict(train_x)
              train_acc = accuracy_score(train_y, train_y_pred)
              print('Train acc: {}' .format(train_acc))
              Train acc: 0.847843137254902
In [243]:
           #Test accuracy
              test_y_pred = voting_clf.predict(test_x)
              test_acc = accuracy_score(test_y, test_y_pred)
              print('Test acc: {}' .format(test_acc))
              Test acc: 0.8191911764705883
In [251]:
           dtree_clf = DecisionTreeClassifier(max_depth=20)
              log clf = LogisticRegression(multi class='multinomial', solver = 'lbfgs', C=10
              sgd clf = SGDClassifier(max iter=10000, tol=1e-3)
              voting_clf = VotingClassifier(
                          estimators=[('dt', dtree_clf),
                                      ('lr', log_clf),
                                       ('sgd', sgd_clf)],
                          voting='hard',
                          weights=[0.5, 0.25, 0.25])
              voting clf.fit(train x, train y)
   Out[251]:
                                        VotingClassifier
                                                   1r
                           dt
                                                                      sgd
                ▶ DecisionTreeClassifier
                                                                ▶ SGDClassifier
                                          ▶ LogisticRegression
In [252]:
           ▶ #Train accuracy
              train y pred = voting clf.predict(train x)
              train_acc = accuracy_score(train_y, train_y_pred)
              print('Train acc: {}' .format(train_acc))
```

Train acc: 0.908921568627451

```
In [253]: #Test accuracy

test_y_pred = voting_clf.predict(test_x)

test_acc = accuracy_score(test_y, test_y_pred)

print('Test acc: {}' .format(test_acc))
```

Test acc: 0.7868382352941177

# Train a bagging classifier

```
In [254]:
             #If you want to do pasting, change "bootstrap=False"
             #n jobs=-1 means use all CPU cores
             #bagging automatically performs soft voting
             bag_clf = BaggingClassifier(
                        SGDClassifier(), n_estimators=100,
                        max samples=1000, bootstrap=True, n jobs=-1)
             bag_clf.fit(train_x, train_y)
   Out[254]:
                     BaggingClassifier
              ▶ base_estimator: SGDClassifier
                      ▶ SGDClassifier
In [255]:
             train_y_pred = bag_clf.predict(train_x)
             train_acc = accuracy_score(train_y, train_y_pred)
             print('Train acc: {}' .format(train_acc))
             Train acc: 0.8215686274509804
In [256]:
         #Test accuracy
             test_y_pred = bag_clf.predict(test_x)
             test_acc = accuracy_score(test_y, test_y_pred)
             print('Test acc: {}' .format(test_acc))
```

Test acc: 0.8196323529411764

### Train a random forest classifier

```
In [257]:
          ★ from sklearn.ensemble import RandomForestClassifier
              rnd_clf = RandomForestClassifier(n_estimators=1000, max_depth=15, n_jobs=-1)
              rnd_clf.fit(train_x, train_y)
   Out[257]:
                                     RandomForestClassifier
              RandomForestClassifier(max_depth=15, n_estimators=1000, n_jobs=-1)
In [258]:
          #Train accuracy
              train_y_pred = rnd_clf.predict(train_x)
             train_acc = accuracy_score(train_y, train_y_pred)
             print('Train acc: {}' .format(train_acc))
              Train acc: 0.9312745098039216
In [259]:
          test_y_pred = rnd_clf.predict(test_x)
             test_acc = accuracy_score(test_y, test_y_pred)
              print('Test acc: {}' .format(test_acc))
              Test acc: 0.8152205882352941
In [260]:
         rnd_clf.feature_importances_
   Out[260]: array([0.08939691, 0.14837393, 0.05420347, 0.08227138, 0.15398757,
                    0.00337793, 0.05872083, 0.03952672, 0.05564999, 0.05959961,
                    0.087997 , 0.02044607, 0.02047596, 0.12597265])
In [261]:
           ▶ | np.round(rnd_clf.feature_importances_,2)
   Out[261]: array([0.09, 0.15, 0.05, 0.08, 0.15, 0. , 0.06, 0.04, 0.06, 0.06, 0.09,
                    0.02, 0.02, 0.13])
```

```
In [262]: ▶ from sklearn.ensemble import ExtraTreesClassifier
              ext_clf = ExtraTreesClassifier(n_estimators=1000, max_depth=10, n_jobs=-1)
              ext_clf.fit(train_x, train_y)
   Out[262]:
                                     ExtraTreesClassifier
              ExtraTreesClassifier(max_depth=10, n_estimators=1000, n_jobs=-1)
           ▶ #Train accuracy
In [263]:
              train_y_pred = ext_clf.predict(train_x)
              train_acc = accuracy_score(train_y, train_y_pred)
              print('Train acc: {}' .format(train_acc))
              Train acc: 0.8111274509803922
In [264]: 

#Test accuracy
              test_y_pred = ext_clf.predict(test_x)
              test_acc = accuracy_score(test_y, test_y_pred)
              print('Test acc: {}' .format(test_acc))
```

### Test acc: 0.7918382352941177

## Train an adaboost classifier

```
In [266]:
          ▶ #Train accuracy
             train_y_pred = ada_clf.predict(train_x)
             train_acc = accuracy_score(train_y, train_y_pred)
             print('Train acc: {}' .format(train_acc))
             Train acc: 0.9844607843137255
In [267]:
         test_y_pred = ada_clf.predict(test_x)
             test_acc = accuracy_score(test_y, test_y_pred)
             print('Test acc: {}' .format(test_acc))
             Test acc: 0.7879411764705883
         Train a gradient boosting classifier
In [268]:  ▶ #Use GradientBoosting
             from sklearn.ensemble import GradientBoostingClassifier
             gbclf = GradientBoostingClassifier(max_depth=2, n_estimators=100, learning_ra
             gbclf.fit(train_x, train_y)
   Out[268]:
                    GradientBoostingClassifier
              GradientBoostingClassifier(max_depth=2)
```

Train acc: 0.8192647058823529

```
In [270]:
          test_y_pred = gbclf.predict(test_x)
             test_acc = accuracy_score(test_y, test_y_pred)
             print('Test acc: {}' .format(test_acc))
             Test acc: 0.8119852941176471
In [271]:
         ▶ #Train on 75% of the sample only
             gbclf = GradientBoostingClassifier(max_depth=2, n_estimators=100,
                                             learning_rate=0.1, subsample=0.75)
             gbclf.fit(train_x, train_y)
   Out[271]:
                           GradientBoostingClassifier
             GradientBoostingClassifier(max depth=2, subsample=0.75)
train_y_pred = gbclf.predict(train_x)
             train_acc = accuracy_score(train_y, train_y_pred)
             print('Train acc: {}' .format(train_acc))
             Train acc: 0.8188235294117647
In [273]:
         test_y_pred = gbclf.predict(test_x)
             test_acc = accuracy_score(test_y, test_y_pred)
             print('Test acc: {}' .format(test_acc))
             Test acc: 0.8120588235294117
 In [ ]:
```