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Project No. 1: Mercedes-Benz Greener Manufacturing
# Step1: Import the required libraries
# linear algebra
import numpy as np
# data processing, CSV file I/O (e.g. pd.read_csv)
import pandas as pd
# for dimensionality reduction
from sklearn.decomposition import PCA
# Step2: Read the data from train.csv
df train = pd.read csv('train.csv')
# let us understand the data
print('Size of training set: {} rows and {} columns'
   .format(*df train.shape))
# print few rows and see how the data looks like
df train.head()
# Step3: Collect the Y values into an array
# seperate the y from the data as we will use this to learn as
# the prediction output
y_train = df_train['y'].values
# Step4: Understand the data types we have
# iterate through all the columns which has X in the name of the column
cols = [c for c in df train.columns if 'X' in c]
print('Number of features: {}'.format(len(cols)))
print('Feature types:')
df train[cols].dtypes.value counts()
# Step5: Count the data in each of the columns
counts = [[], [], []]
for c in cols:
 typ = df_train[c].dtype
  uniq = len(np.unique(df train[c]))
  if uniq == 1:
   counts[0].append(c)
 elif uniq == 2 and typ == np.int64:
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counts[1].append(c)

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else:
    counts[2].append(c)
print('Constant features: {} Binary features: {} Categorical features: {}\n'
   .format(*[len(c) for c in counts]))
print('Constant features:', counts[0])
print('Categorical features:', counts[2])
# Step6: Read the test.csv data
df test = pd.read csv('test.csv')
# remove columns ID and Y from the data as they are not used for learning
usable columns = list(set(df train.columns) - set(['ID', 'y']))
y train = df train['y'].values
id_test = df_test['ID'].values
x_train = df_train[usable_columns]
x_test = df_test[usable_columns]
# Step7: Check for null and unique values for test and train sets
def check_missing_values(df):
  if df.isnull().any().any():
    print("There are missing values in the dataframe")
  else:
    print("There are no missing values in the dataframe")
check missing values(x train)
check missing values(x test)
# Step8: If for any column(s), the variance is equal to zero,
# then you need to remove those variable(s).
# Apply label encoder
for column in usable columns:
  cardinality = len(np.unique(x_train[column]))
  if cardinality == 1:
    x train.drop(column, axis=1) # Column with only one
    # value is useless so we drop it
    x test.drop(column, axis=1)
  if cardinality > 2: # Column is categorical
    mapper = lambda x: sum([ord(digit) for digit in x])
    x train[column] = x train[column].apply(mapper)
    x_test[column] = x_test[column].apply(mapper)
x train.head()
# Step9: Make sure the data is now changed into numericals
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print('Feature types:')
x_train[cols].dtypes.value_counts()
# Step10: Perform dimensionality reduction
# Linear dimensionality reduction using Singular Value Decomposition of
# the data to project it to a lower dimensional space.
n comp = 12
pca = PCA(n_components=n_comp, random_state=420)
pca2_results_train = pca.fit_transform(x_train)
pca2 results test = pca.transform(x test)
# Step11: Training using xgboost
import xgboost as xgb
from sklearn.metrics import r2_score
from sklearn.model selection import train test split
x_train, x_valid, y_train, y_valid = train_test_split(
    pca2 results train,
    y train, test size=0.2,
    random_state=4242)
d_train = xgb.DMatrix(x_train, label=y_train)
d_valid = xgb.DMatrix(x_valid, label=y_valid)
#d test = xgb.DMatrix(x test)
d_test = xgb.DMatrix(pca2_results_test)
params = {}
params['objective'] = 'reg:linear'
params['eta'] = 0.02
params['max depth'] = 4
def xgb r2 score(preds, dtrain):
  labels = dtrain.get label()
  return 'r2', r2_score(labels, preds)
watchlist = [(d_train, 'train'), (d_valid, 'valid')]
clf = xgb.train(params, d train,
        1000, watchlist, early stopping rounds=50,
        feval=xgb_r2_score, maximize=True, verbose_eval=10)
# Step12: Predict your test df values using xgboost
p test = clf.predict(d test)
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