models

October 28, 2018

— title: Technical test for Junior Data Scientist position at Datrik Intelligence author: Antonio Ortega date: October 28, 2018—

1 Load libraries

```
In [25]: import pandas as pd
         import matplotlib.pyplot as plt
         import numpy as np
         import os.path
         from sklearn.model_selection import GridSearchCV
         from sklearn.decomposition import PCA
         from sklearn.model_selection import train_test_split
         from sklearn.linear_model import Ridge
         from sklearn.ensemble import GradientBoostingClassifier
         from sklearn.ensemble import AdaBoostClassifier
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.metrics import accuracy_score
         from sklearn.metrics import roc_auc_score
         from sklearn.metrics import roc_curve
         from sklearn.metrics import auc
         from sklearn.preprocessing import StandardScaler
```

2 Load datasets

```
In [29]: x_train_mca = pd.read_csv(os.path.join("proc_data", "x_train_mca.csv"))
         x_train = pd.read_csv(os.path.join("proc_data", "x_train.csv"))
         x_train = pd.concat([x_train, x_train_mca.filter(like='MCA_')], axis=1)
         # x_test_mca = pd.read_csv(os.path.join("proc_data", "x_test_mca.csv"))
         y_train = pd.read_csv(os.path.join("proc_data", "y_train.csv"))
         X_train, X_test, y_train, y_test = train_test_split(x_train, y_train, test_size=0.33, random_s

In [30]: print("{} features loaded".format(X_train.columns.values.shape[0]))
         feature_count = np.unique([e.replace(".", "_").split("_")[0] for e in X_train.columns.values.t
         pd.DataFrame({"feature":feature_count[0], "count": feature_count[1]})

815 features loaded

Out[30]:    feature  count
         0      MCA      5
         1   binary      3
         2  counter      7
         3     edad      1
         4       et    719
```

```
5  farmaco     21
6  nominal     51
7  ordinal      2
8     raza      5
9     sexo      1
```

In [4]:
```
# plt.scatter(X_train["Dim 1"], X_train["Dim 2"], c=y_train.Y)
# plt.show()
```

In [31]:
```
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)
```

```
/home/antortjim/anaconda3/envs/ML/lib/python3.6/site-packages/sklearn/preprocessing/data.py:617: DataCo
  return self.partial_fit(X, y)
/home/antortjim/anaconda3/envs/ML/lib/python3.6/site-packages/sklearn/base.py:462: DataConversionWarnin
  return self.fit(X, **fit_params).transform(X)
/home/antortjim/anaconda3/envs/ML/lib/python3.6/site-packages/ipykernel_launcher.py:3: DataConversionWa
  This is separate from the ipykernel package so we can avoid doing imports until
```

In [32]:
```
# pca = PCA(n_components=2, svd_solver='full')
# pca.fit(X_train_scaled)
# X_train_transform = pd.DataFrame(data=pca.transform(X_train_scaled), columns=["PC1","PC2"])
# X_train_transform.head()
```

Out[32]:

	PC1	PC2
0	-1.035688	-1.141891
1	-1.088252	1.600154
2	1.883554	2.612355
3	-0.372338	-1.955333
4	-0.039655	3.380544

In [37]:
```
ridge = Ridge(normalize=True, max_iter=2000)
parameters = {"alpha": [0, 1, 5, 7, 9, 10]}
ridge_cv = GridSearchCV(estimator=ridge, param_grid=parameters, cv=5)
ridge_cv.fit(X_train, y_train)
```

Out[37]:
```
GridSearchCV(cv=5, error_score='raise-deprecating',
       estimator=Ridge(alpha=1.0, copy_X=True, fit_intercept=True, max_iter=2000,
   normalize=True, random_state=None, solver='auto', tol=0.001),
       fit_params=None, iid='warn', n_jobs=None,
       param_grid={'alpha': [0, 1, 5, 7, 9, 10]}, pre_dispatch='2*n_jobs',
       refit=True, return_train_score='warn', scoring=None, verbose=0)
```

In [38]:
```
y_predict=np.round(ridge_cv.predict(X_train))
```

In [39]:
```
ridge_cv.cv_results_["params"]
ridge_cv.best_estimator_
```

Out[39]:
```
Ridge(alpha=1, copy_X=True, fit_intercept=True, max_iter=2000, normalize=True,
   random_state=None, solver='auto', tol=0.001)
```

In [40]:
```
ridge_cv.score(X_train, y_train)
```

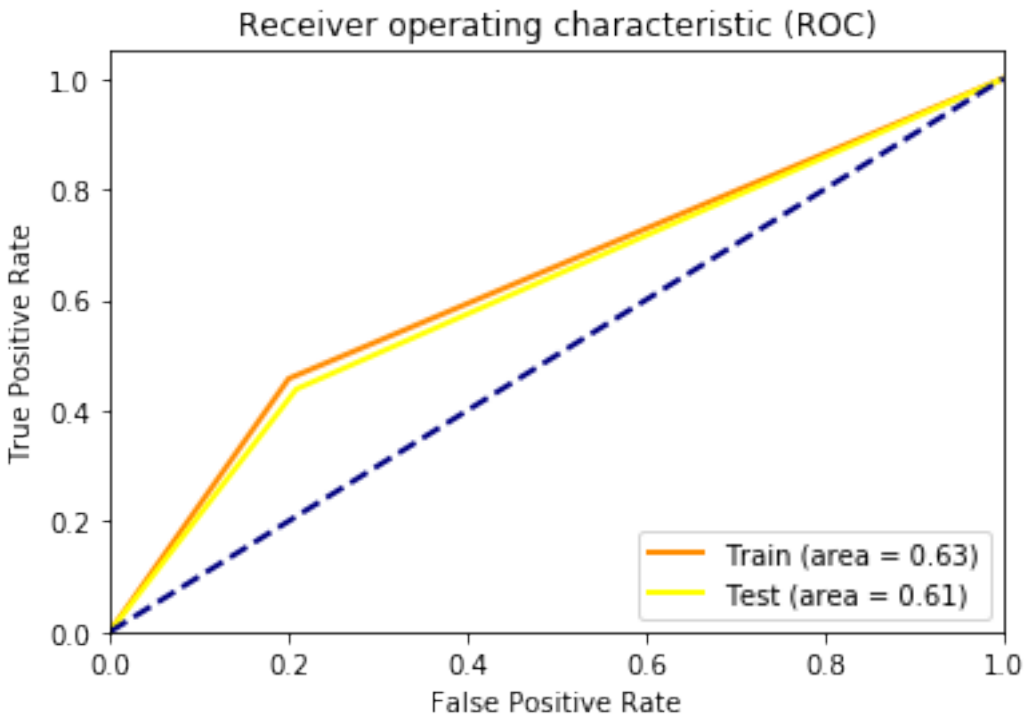
Out[40]:
```
0.10634049318208794
```

In [41]:
```
y_predict_train = ridge_cv.predict(X_train)
y_predict_train_bin = np.round(y_predict_train)
print("Accuracy score (train): {0:.3f}".format(accuracy_score(y_train, y_predict_train_bin, no
print("AUC score (train): {0:.3f}".format(roc_auc_score(y_train, y_predict_train_bin)))

y_predict_test = ridge_cv.predict(X_test)
y_predict_test_bin = np.round(y_predict_test)
print("Accuracy score (test): {0:.3f}".format(accuracy_score(y_test, y_predict_test_bin, normal
print("AUC score (test): {0:.3f}".format(roc_auc_score(y_test, y_predict_test_bin)))
```

```
Accuracy score (train): 0.643
AUC score (train): 0.629
Accuracy score (test): 0.627
AUC score (test): 0.615
```

In [42]:
```
def plot_roc(y_train, y_predict_train_bin, y_test, y_predict_test_bin):
    lw=2
    fpr = dict()
    tpr = dict()
    roc_auc = dict()
    fpr[0], tpr[0], _ = roc_curve(y_true=y_train, y_score = y_predict_train_bin)
    roc_auc[0] = auc(fpr[0], tpr[0])
    fpr[1], tpr[1], _ = roc_curve(y_true=y_test, y_score = y_predict_test_bin)
    roc_auc[1] = auc(fpr[1], tpr[1])
    cols = ['darkorange', "yellow"]
    datasets = ["Train", "Test"]
    plt.figure()
    for i in range(2):
      plt.plot(fpr[i], tpr[i], color=cols[i],
               lw=lw, label='%s (area = %0.2f)' % (datasets[i], roc_auc[i]))

    plt.plot([0, 1], [0, 1], color='navy', lw=lw, linestyle='--')
    plt.xlim([0.0, 1.0])
    plt.ylim([0.0, 1.05])
    plt.xlabel('False Positive Rate')
    plt.ylabel('True Positive Rate')
    plt.title('Receiver operating characteristic (ROC)')
    plt.legend(loc="lower right")
    plt.show()
```

In [43]:
```
plot_roc(y_train, y_predict_train_bin, y_test, y_predict_test_bin)
```




```
In [44]: learning_rates = [0.05, 0.1, 0.3, 0.5]
         gb = GradientBoostingClassifier(random_state = 0, subsample = 0.9, n_estimators=500)
         param_grid = dict(learning_rate = learning_rates,
                           #n_estimators = [10,50,100],
                           max_features=[1,5,10], max_leaf_nodes = [2,3,4])
         gb_cv = GridSearchCV(estimator=gb, param_grid=param_grid, cv=5, verbose=True,n_jobs=2)
         gb_cv.fit(X_train_scaled, y_train.values.reshape((-1,)))
```

```
Fitting 5 folds for each of 36 candidates, totalling 180 fits

[Parallel(n_jobs=2)]: Using backend LokyBackend with 2 concurrent workers.
[Parallel(n_jobs=2)]: Done  46 tasks      | elapsed:  5.4min
[Parallel(n_jobs=2)]: Done 180 out of 180 | elapsed: 20.6min finished
```

```
Out[44]: GridSearchCV(cv=5, error_score='raise-deprecating',
                estimator=GradientBoostingClassifier(criterion='friedman_mse', init=None,
                       learning_rate=0.1, loss='deviance', max_depth=3,
                       max_features=None, max_leaf_nodes=None,
                       min_impurity_decrease=0.0, min_impurity_split=None,
                       min_samples_leaf=1, min_sampl...      subsample=0.9, tol=0.0001, validation_fracti
                       verbose=0, warm_start=False),
                fit_params=None, iid='warn', n_jobs=2,
                param_grid={'learning_rate': [0.05, 0.1, 0.3, 0.5], 'max_features': [1, 5, 10], 'max_leaf
                pre_dispatch='2*n_jobs', refit=True, return_train_score='warn',
                scoring=None, verbose=True)
```

```
In [46]: y_predict_train = gb_cv.predict(X_train_scaled)
         y_predict_train_bin = np.round(y_predict_train)
```

```
        print("Accuracy score (training): {0:.3f}".format(accuracy_score(y_train, y_predict_train_bin,
        print("AUC score (training): {0:.3f}".format(roc_auc_score(y_train, y_predict_train_bin)))

        y_predict_test = gb_cv.predict(X_test_scaled)
        y_predict_test_bin = np.round(y_predict_test)
        print("Accuracy score (test): {0:.3f}".format(accuracy_score(y_test, y_predict_test_bin, norma
        print("AUC score (test): {0:.3f}".format(roc_auc_score(y_test, y_predict_test_bin)))
        # print("Learning rate: ", learning_rate)
        # print("Accuracy score (training): {0:.3f}".format(gb.score(X_train_scaled, y_train)))
        # print("Accuracy score (validation): {0:.3f}".format(gb.score(X_test_scaled, y_test)))
```

```
Accuracy score (training): 0.652
AUC score (training): 0.641
Accuracy score (test): 0.642
AUC score (test): 0.632
```

```
In [48]: plot_roc(y_train, y_predict_train_bin, y_test, y_predict_test_bin)
```