Machine Learning Cheat Sheet - Python

Regression

X = df["features columns"]
y = df["label column"]
import statsmodels.formula.api as sm
model = sm.ols(formula="y ~ X1,X2", data=df)
fitted = model.fit()
fitted.summary2()
from sklearn.linear_model import LinearRegression
lr = LinearRegression()
lr.fit(X, y)
predictions = lr.predict(X)

Logistic Regression

from statsmodels.formula.api import ols
model = sm.ols(formula="Bought ~ Age", data=sales)
fitted = model.fit()
fitted.summary()
Other Option
from sklearn.linear_model import LogisticRegression
logistic = LogisticRegression()
logistic.fit(X,y)
predictions=logistic.predict(X_test)

Decision Trees

Dietting the T

Plotting the Trees install pydot: conda install -c anaconda pydot Also Install graphviz in the system

from IPython.display import Image

from sklearn.externals.six import StringIO

import pydot

dot_data = StringIO()

tree.export_graphviz(clf,out_file = dot_data,feature_names = features,filled=True, rounded=True,impurity=False) graph = pydot.graph_from_dot_data(dot_data.getvalue()) Image(graph.create_png())

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Training and R&D - Data Science and Deep Learning



GRN

from sklearn.ensemble import GradientBoostingClassifier
clf=GradientBoostingClassifier(
 loss ="exponential", #for AdaBoost : "exponential"
 learning_rate = 0.1, #shrinkage
 n_estimators = 100, #boosting stages to perform
 max_depth = 4, #number of nodes in tree *Important*
 criterion = "friedman_mse",#function: quality of a split
 min_samples_split : 3, #min samples for each split
 max_features = "sqrt",#max feature in tree, sqrt of total
 verbose = 1, #to print progress, 0: don't print)
clf.fit(X,y)
y_pred = clf.predict(X_test)
accuracy = clf.score(X_test, y_test)

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Model Validation Metrics
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from sklearn, metrics import confusion matrix
cm = confusion_matrix(y, y_pred)
total1=sum(sum(cm))
from confusion matrix calculate accuracy
accuracy=(cm [0,0]+cm [1,1])/total1
Sensitivity=cm[0,0]/(cm[0,0]+cm[0,1])
specificity=cm[1,1]/(cm[1,0]+cm[1,1])
                        ROC Curve and AUC
from sklearn.metrics import roc_curve, auc
import matplotlib.pyplot as plt
false_positive_rate, true_positive_rate, thresholds = roc_curve(y,
y_pred)
plt.title(Receiver Operating Characteristic)
plt.plot(false_positive_rate, true_positive_rate)
plt.plot([0,1],[0,1],r--)
plt.ylabel(True Positive Rate(Sensitivity))
plt.xlabel(False Positive Rate(Specificity))
plt.show()
                      Area Under Curve - AUC
roc_auc = auc(false_positive_rate, true_positive_rate)
roc_auc
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X,y, test_size =
                      K-fold Cross Validation
```

from sklearn import cross_validation
kfold = cross_validation.KFold(len(X), n_folds=10)
score = cross_validation.cross_val_score(clf, X, y, cv=kfold)
Mean Kfold accuracy
score.mean()

SVM

from sklearn import svm
clf = svm.SVC(kernel="linear")
model =clf.fit(X,y)
y_pred = model.predict(X_test)
Accuracy = model.score(X_test, y_test)

Neural Network

pip install neurolab import neurolab as nl import numpy as np #Defining Network: in the example below 1st argument is a list of min-max values of predictor variables: 2 input with range [0,1] each 2nd argument is list of num of nodes in each layer 4 noded hidden layer and 1 noded outlayer #Transf: list of transfer function applied in each layer in order net = nl.net.newff([[0, 1],[0,1]],[4,1],transf=[nl.trans.LogSig()] * 2) net.trainf = nl.train.train_rprop #Training Network error = [] error.append(net.train(X, y, epochs = 100, goal=0.001)) #Simulate Network (predicting) predicted_values = net.sim(X_test)

Random Forest

from sklearn.ensemble import RandomForestClassifier
clf=RandomForestClassifier(
 n_estimators =100, #number of trees in forest
 criterion ="entropy", #tree splitting criterion
 max_features = "sqrt", #max feature in tree, sqrt of total
 max_depth = 4, #number of nodes in the tree *Important*
 min_samples_split : 3, #min samples for each split
 Bootstrap = True, #if samples are bootstrapped
 class_weight = "balanced") #to handle class imbalance
clf.fit(X,y)
y_pred = clf.predict(X_test)
accuracy = clf.score(X_test, y_test)