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# ANOMALY DETECTION
import time
import numpy as np
import matplotlib
import matplotlib.pyplot as plt
from sklearn import svm
from sklearn.datasets import make_moons, make_blobs
from sklearn.covariance import EllipticEnvelope
from sklearn.ensemble import IsolationForest
from sklearn.neighbors import LocalOutlierFactor
matplotlib.rcParams['contour.negative_linestyle'] = 'solid'
n_samples = 300
outliers fraction = 0.15
n_outliers = int(outliers_fraction * n_samples)
n_inliers = n_samples - n_outliers
anomaly_algorithms = [
  ("Robust covariance", EllipticEnvelope(contamination=outliers_fraction)),
  ("One-Class SVM", svm.OneClassSVM(nu=outliers_fraction, kernel="rbf",
                       gamma=0.1)),
  ("Isolation Forest", IsolationForest(contamination=outliers_fraction,
                        random_state=42)),
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("Local Outlier Factor", LocalOutlierFactor(
    n neighbors=35, contamination=outliers fraction))]
# Define datasets
blobs_params = dict(random_state=0, n_samples=n_inliers, n_features=2)
datasets = [
  make_blobs(centers=[[0, 0], [0, 0]], cluster_std=0.5,
          **blobs_params)[0],
  make_blobs(centers=[[2, 2], [-2, -2]], cluster_std=[0.5, 0.5],
          *blobs params)[0],
  make_blobs(centers=[[2, 2], [-2, -2]], cluster_std=[1.5, .3],
          *blobs params)[0].
  4. * (make_moons(n_samples=n_samples, noise=.05, random_state=0)[0] -
      np.array([0.5, 0.25])),
  14. * (np.random.RandomState(42).rand(n_samples, 2) - 0.5)]
# Compare given classifiers under given settings
xx, yy = np.meshgrid(np.linspace(-7, 7, 150),
             np.linspace(-7, 7, 150))
plt.figure(figsize=(len(anomaly_algorithms) * 2 + 4, 12.5))
plt_subplots_adjust(left=.02, right=.98, bottom=.001, top=.96, wspace=.05,
            hspace=.01)
plot_num = 1
rng = np.random.RandomState(42)
for i_dataset, X in enumerate(datasets):
  # Add outliers
  X = np concatenate([X, rng uniform(low=-6, high=6, size=(n_outliers, 2))],
              axis=0)
  for name, algorithm in anomaly_algorithms:
     t0 = time.time()
     algorithm.fit(X)
     t1 = time.time()
     plt_subplot(len(datasets), len(anomaly_algorithms), plot_num)
     if i dataset == 0:
       plt.title(name, size=18)
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# fit the data and tag outliers
     if name == "Local Outlier Factor":
       y_pred = algorithm.fit_predict(X)
     else:
       y_pred = algorithm.fit(X).predict(X)
     # plot the levels lines and the points
     if name != "Local Outlier Factor": # LOF does not implement predict
       Z = algorithm.predict(np.c_[xx.ravel(), yy.ravel()])
       Z = Z.reshape(xx.shape)
       plt.contour(xx, yy, Z, levels=[0], linewidths=2, colors='black')
     colors = np.array(['#377eb8', '#ff7f00'])
     plt.scatter(X[:, 0], X[:, 1], s=10, color=colors[(y_pred + 1) // 2])
     plt.xlim(-7, 7)
     plt.ylim(-7, 7)
     plt_xticks(())
     plt.yticks(())
     plt.text(.99, .01, ('%.2fs' % (t1 - t0)).lstrip('0'),
          transform=plt.gca().transAxes, size=15,
          horizontalalignment='right')
     plot_num += 1
plt.show()
# THE END
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