

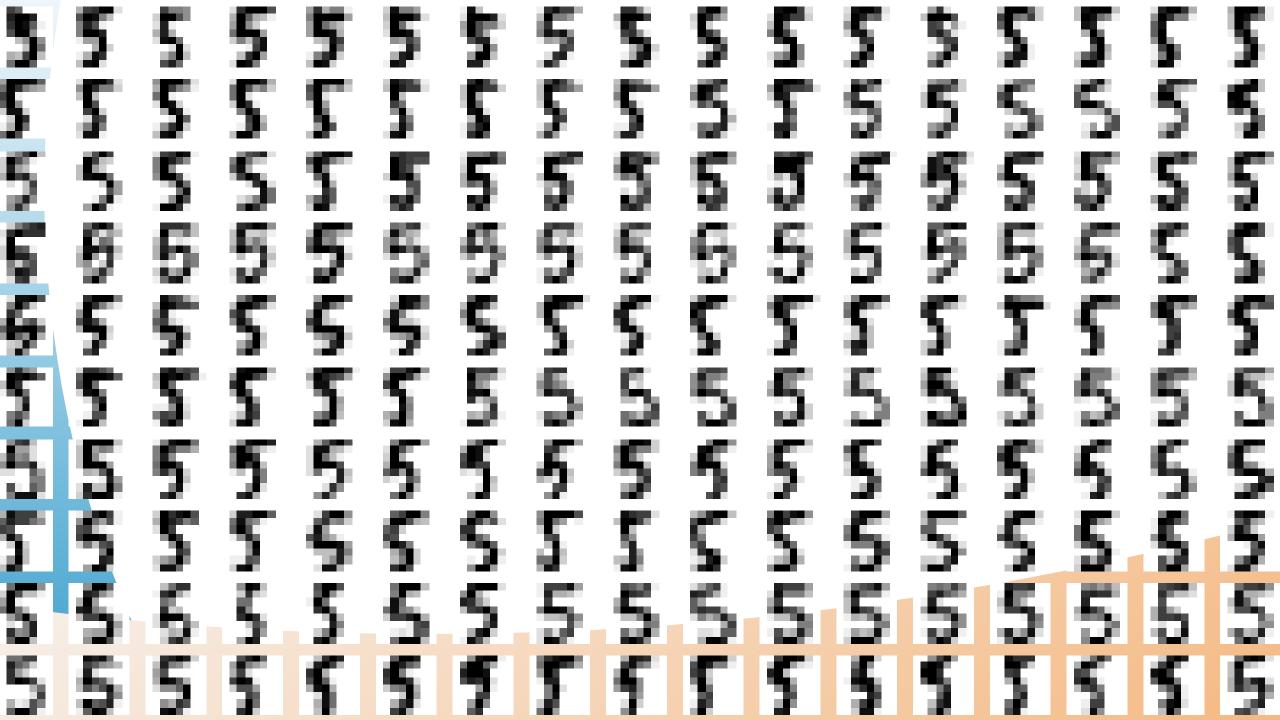


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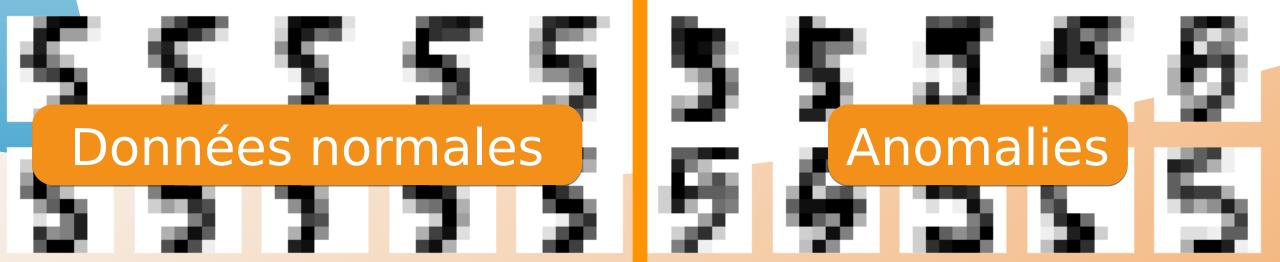
Capport de scikitlearn à la détection d'anomalies





Données

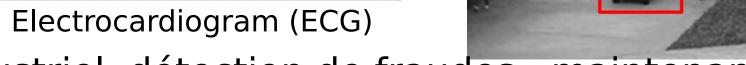
- >>> from sklearn.ensemble import IsolationForest
- >>> iforest = IsolationForest(contamination=0.05)
- >>> iforest = iforest.fit(X)
 - permet de distinguer les:



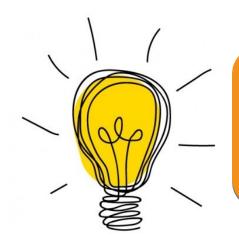
Applications:







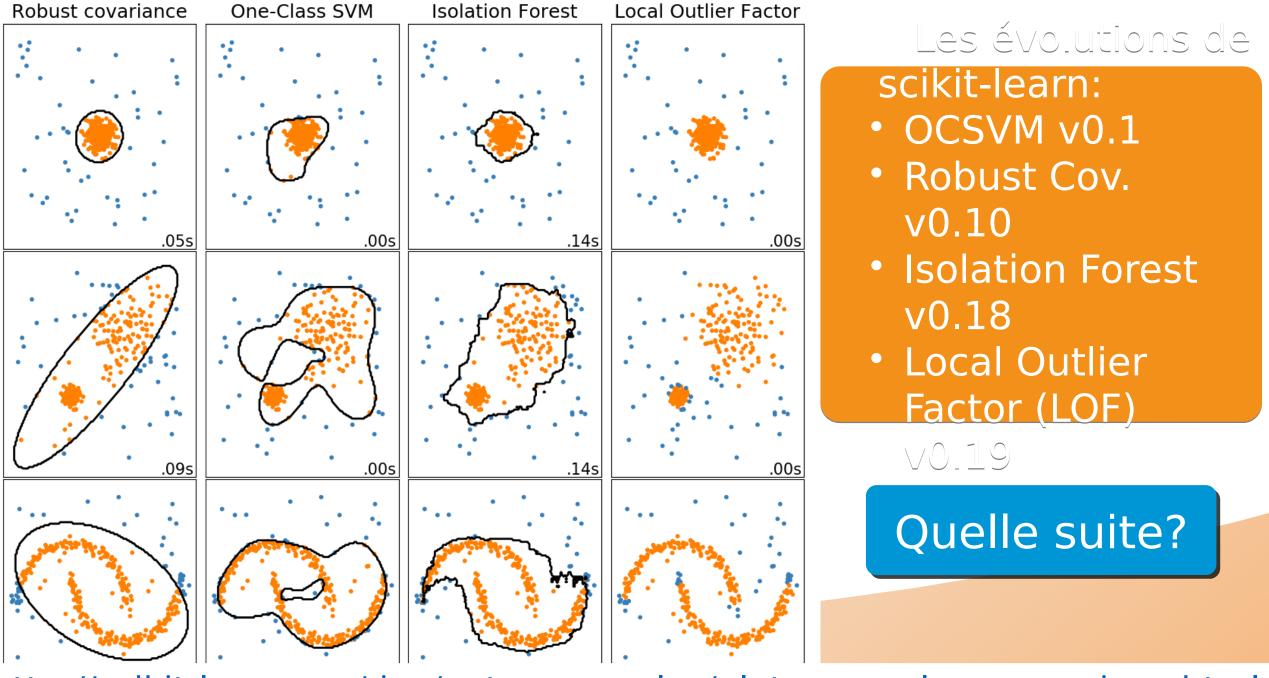
Suivi de processus industriel, détection de fraudes, maintenance pré



évènements statistiquement

Trouver es

rares



http://scikit-learn.org/dev/auto_examples/plot_anomaly_comparison.html

http://scikit-learn.org/stable/modules/outlier_detection.html



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scikit-learn v0.19.2
Other versions

Please **cite us** if you use the software.

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2.7.1. Novelty Detection

2.7.2. Outlier Detection

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 Elliptic Envelope versus Isolation
 Forest versus LOF

2.7. Novelty and Outlier Detection

Many applications require being able to decide whether a new observation belongs to the same distribution as existing observations (it is an inlier), or should be considered as different (it is an outlier). Often, this ability is used to clean real data sets. Two important distinction must be made:

novelty detection:

The training data is not polluted by outliers, and we are interested in detecting anomalies in new observations.

outlier detection:

The training data contains outliers, and we need to fit the central mode of the training data, ignoring the deviant observations.

The scikit-learn project provides a set of machine learning tools that can be used both for novelty or outliers detection. This strategy is implemented with objects learning in an unsupervised way from the data:

estimator.fit(X_train)

new observations can then be sorted as inliers or outliers with a predict method:

estimator.predict(X_test)

Inliers are labeled 1, while outliers are labeled -1.

2.7.1. Novelty Detection