ANOMALY DETECTION

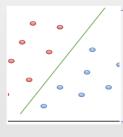
Rodwel Mupambirei Consulting Actuary

PyData Bristol 18 July 2019

OUTLINE

- Why I care about anomalies in my models.
- What methods are currently used to detect anomalies and their effectiveness (or lack of).
- Deep dive into a method that actually works.
- Incorporating anomaly detection in pipelines

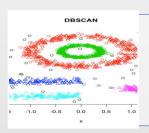
ANOMALY DETECTION METHODS



Model Based

SVM

Statistical



Density Based

DBSCAN

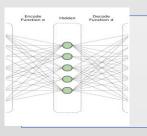
LOF



Distance based

K-means

KNN

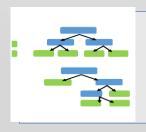


Network based

Autoencoder



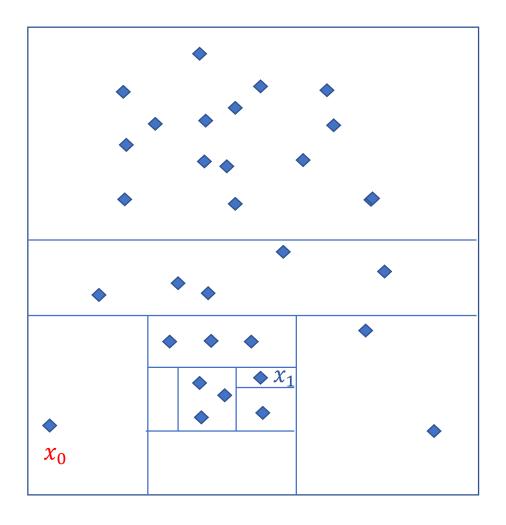
Rules Based

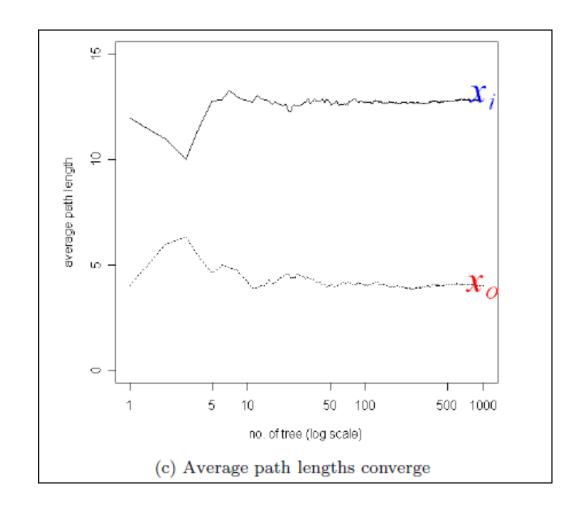


Trees



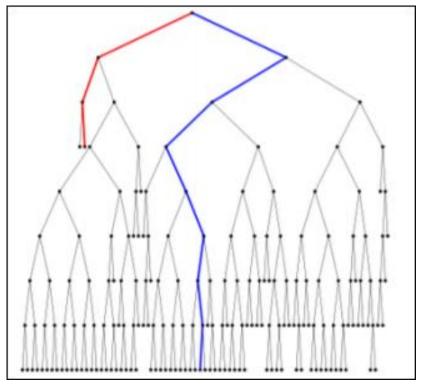
GROWING ISOLATION FORESTS



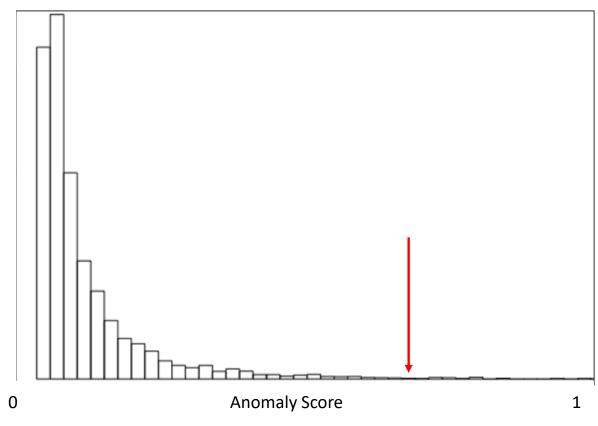


PREDICTION ISOLATION PROCESS

Anomaly Score Distribution

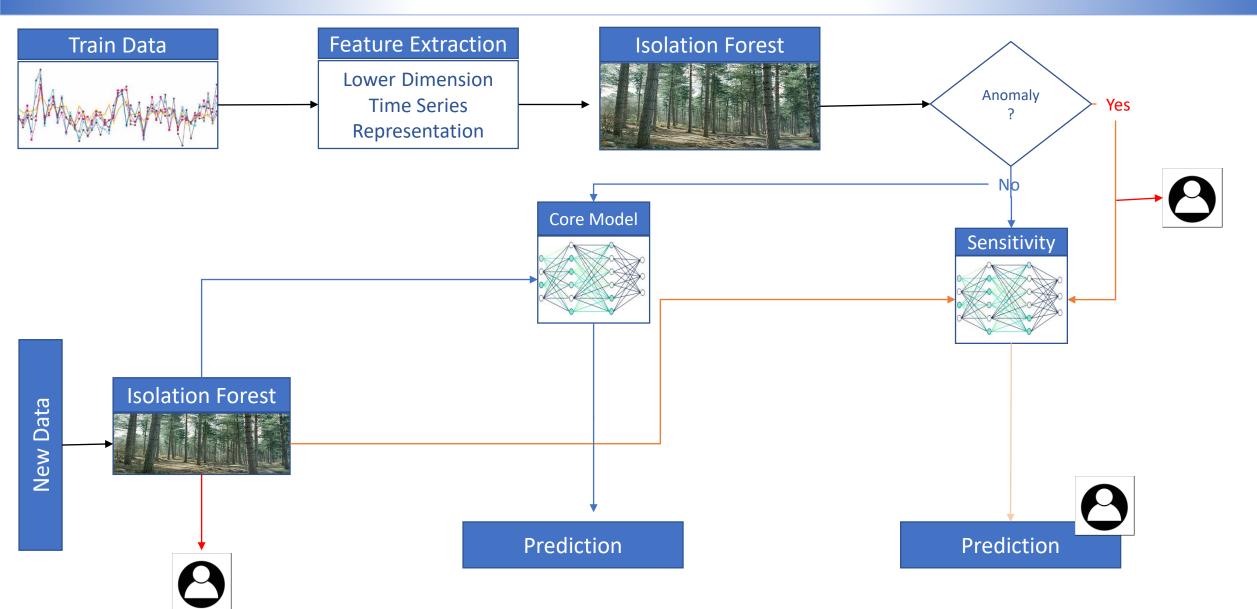


Anomaly Score Distribution



Sahand Hariri and Matias Carrasco Kind Extended Isolation Forest for
Anomaly Detection
https://github.com/sahandha/eif

PIPELINE

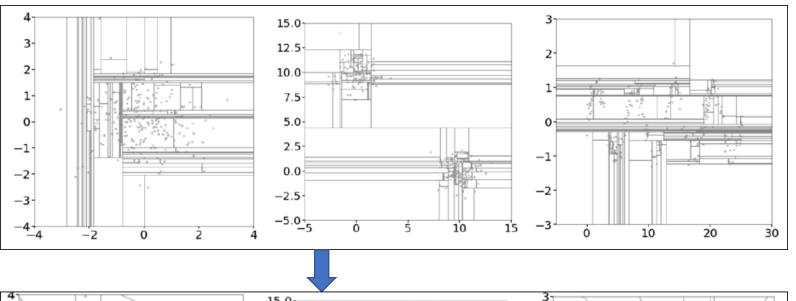


IMPLEMENTATION

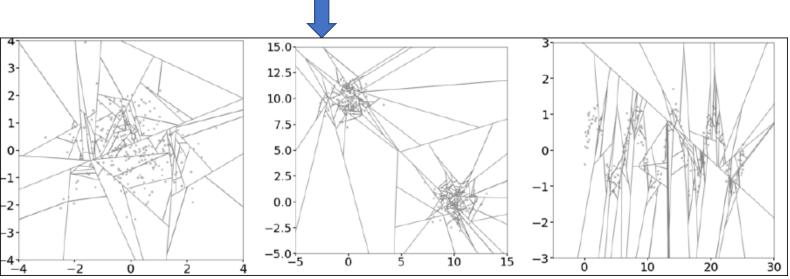
scikit-learn implementation of Isolation Forest

```
In [44]: # Isolation Forest ----
         from sklearn.ensemble import IsolationForest
         import pandas as pd
         # training the model
         Model = IsolationForest(behaviour='new', max_samples=100, contamination=0.053)
         Model.fit(df)
In [45]: Anomaly Scores
         -Model.score samples(df)
Out[45]: array([0.54313713, 0.59857059, 0.59667779, ..., 0.35795571, 0.35758752,
                0.3568004 ])
         #predictions
In [47]:
         Model.predict(df)
Out[47]: array([ 1, -1, -1, ..., 1, 1, 1])
```

EXTENDED ISOLATION FOREST (eIF)



IF splits the data using a single feature at each split



EIF uses vectors rather that a single feature in the partitioning

PIPELINE

I have hopefully convinced you that:

You should be using isolation forests to make your models more robust

Isolation Forest are an efficient method for identifying anomalies

Isolation Forest based anomaly detection can be integrated in Machine Leaning Pipelines