Developing Spidey Senses

Anomaly Detection for Javascript

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Spidey Sense?

- tingling sensation on the back of Peter Parker's skull
- ability to sense / react to danger

Uses

- Increases his ability to detect evil (and even clones)
- Helps him navigate if he is impaired (disoriented or unable to see/hear)
- Aids him in discovering secret passageways and find hidden/lost objects
- Helps fire his Web Shooters and swing instinctively



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Real Spider Sense

- "hyper-awareness"
- long, thin hairs, *trichobothria*
- low-level vibrations through their web
- can detect the vibrations of faint sounds
- small insects moving up to 3 meters away





Any new web developers here?

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Spidey Sense?

Gut feeling Vibe Feeling Intuition Discover Blind Spots

Learning from the past



Agenda

What is Anomaly Detection?

Time Series Anomaly Detection

Demo

Takeaways

Anomaly Detection

- Identifying unexpected items or events in data sets, which differ from the norm
- An Outlier
- Assumptions:
 - Anomalies only occur very rarely in the data.
 - Their features differ from the normal instances significantly.



Causes of Outliers





Artificial (Error) / Non-natural

Natural

Causes of Outliers

- Data Entry Errors: 100,000 vs 1,000,000 fat fingered
- Measurement Error: common
- Experimental Error: start late in sprint
- Intentional Outlier: underreporting alcohol consumption
- Data Processing Error: extraction errors
- Sampling Error: reporting height for all athletes and included most basketball players
- Natural Outlier: When it's not artificial





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Needle in a haystack

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Rule-based Systems

Methods



Statistical Techniques



Machine Learning

Rule-based Systems



Statistical Techniques



flags the data points => deviate from common statistical properties (mean, median, mode, quantiles)



U

a rolling average or a moving average

n-period simple moving average "low pass filter." e.g. Kalman Filters



More Interpretable and sometimes more useful than ML methods

 $\overline{\mathbb{Q}}$

Supervised (e.g. Decision Tree, SVM, LSTM Forecasting)



Unsupervised (e.g. K-Means, Hierarchical Clustering, DBSCAN) Machine Learning Methods



Self-Supervised (e.g. LSTM Autoencoder)

Anomaly Detection

- Very small number of positive examples
- Large number of negative examples
- Many different "types" of anomalies. Hard to learn from positive examples
- Future anomalies may not be discovered yet.

Supervised Learning

- Large number of positive and negative examples
- Enough positive examples for algorithm to learn.
- Future positive examples likely to be similar to training set

Anomaly Detection

- Fraud Detection
- Manufacturing (engines/machineries)
- Monitoring Data Center
- Internet of Things

Supervised Learning

- Email spam classification
- Weather prediction
- Cancer classification

Machine Learning

Density-Based Anomaly Detection

- based on the k-nearest neighbors' algorithm.
- Assumption: Normal data points occur around a dense neighborhood and abnormalities are far away.

Clustering-Based Anomaly Detection

- Assumption: Data points that are similar tend to belong to clusters --> distance from local centroids.
- K-means







Gaussian Distribution

- Gaussian Distribution and given a new data-point,
- Compute the probability of the data-point
- If the probability is below a threshold => outlier or anomalous.

Machine Learning

Support Vector Machine-Based Anomaly Detection

- OneClassSVM
- >100 features, aggressive boundary
- find a function that is positive for regions with high density of points, and negative for small densities

PCA-Based Anomaly Detection

- analyzing available features to determine what constitutes a "normal" class
- applying distance metrics
- Fast training





Simple Anomaly Detection DEMO

Time Series Data

- Series of data points indexed in time order
- Examples:
- - Logs
 - Stock Market
 - Sales Data
 - Sensors
 - Any data captured with Time Stamp



Internet of Things







Increasing Data Volume (sensors are cheaper)

Increased Data Speed (improved networking)

Risk environment that are moving very fast but failures are not tolerated.

Internet of Broken Things



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Time Series Anomaly Types





Outlier



Spike and Level Shift



Pattern Change



Seasonality

```
Intelligent Kiosk Sample
```

Anomaly Detector

Bike rental Manufacturing Telecom Live sound



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Time Series Anomalies



Multivariate



Azure Cognitive Services

- Al for every developer— w/o requirement ML expertise.
- Just an API call

Decision	Make sma
Language	Anomaly D
Speech	Identify pot Content Ma
Vision	Detect pote
Web search	Personalize Create rich,

rter decisions faster

etector

tential problems early on.

oderator

entially offensive or unwanted content.

personalized experiences for every user.





Detect anomalies as they occur in real-time.

Anomaly Detector Features



Detect anomalies as a batch.

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Automatically adapts and learns from new data



Fine Tune Sensitivity



Anomaly Detector Features



No machine learning expertise needed



Eliminate need for labeled training data



Automatically identify and apply best-fitting model



Gallery of Algorithms Fourier Transformation **Extreme Studentized** Deviate (ESD) **STL Decomposition** Dynamic Threshold Z-score detector **SR-CNN**

Limitations

- Data Granularity Daily, Hourly, Minutely, Monthly, Weekly, Yearly
- Series Data Points 12 to 8640 entries

```
JSON
{
    "granularity": "daily",
    "series": [
        "timestamp": "2018-03-01T00:00:00Z",
        "value": 32858923
        },
        {
        "timestamp": "2018-03-02T00:00:00Z",
        "value": 29615278
        },
        ]
      }
```

Calling the Anomaly Detector API





Client SDK

C#, Python, Node

REST API

Any language supporting HTTP calls

Anomaly Detector Demo



Where can you use this?

C#, Javascript, Python

Docker Containers

Power BI

Azure Databricks for streaming data

Anomaly Detector Demo



Where can you use this?

Javascript, Python, C#

https://docs.microsoft.com/en-us/azure/cognitive-services/anomaly-detector/quickstarts/client-libraries?pivots=programming-languagecsharp&tabs=linux

Docker Containers

https://docs.microsoft.com/en-us/azure/cognitive-services/anomaly-detector/anomaly-detector-container-howto

- Power Bl
- <u>https://docs.microsoft.com/en-us/azure/cognitive-services/anomaly-detector/tutorials/batch-anomaly-detection-powerbi</u>
- Azure Databricks for streaming data
- <u>https://docs.microsoft.com/en-us/azure/cognitive-services/anomaly-detector/tutorials/anomaly-detection-streaming-databricks</u>

Metrics Advisor

- Part of Azure Cognitive Services
- Performs data monitoring, anomaly detection in time series data
- Automates applying models
- Analyze multi-dimensional data from multiple data sources
- Identify and correlate anomalies
- Configure and fine-tune the anomaly detection model
- Diagnose anomalies and help with root cause analysis
- REST API and Web Portal
- Currently in preview





The best superpower you can give to your project is a "spidey-sense".



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https://github.com/rondagdag/spidey-sense-js

Recap

 ✓ What is Anomaly Detection?
 process of identifying unexpected items or events in data sets

 What is Time Series Data series of data points indexed in time order

 Anomaly Detector
 API to detect anomalies automatically adapts
 learns from new data

About Me

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