# Streaming Outlier Analysis for Fun and Scalability

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# Hi, I'm Casey Stella!

- The future involves non-trivial analytics done on streaming data
- It's not just IoT
- There is a need for insights to keep pace with the velocity of your data

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- The Good: There is no shortage of computational frameworks to handle streaming
- The Bad: There are not an overabundance of high-quality outlier analysis frameworks

#### **Outlier Analysis**

Outlier analysis or anomaly detection is the analytical technique by which "interesting" points are differentiated from "normal" points. Often "interesting" implies some sort of error or state which should be researched further.

<sup>&</sup>lt;sup>1</sup>http://arxiv.org/pdf/1603.00567v1.pdf

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Macrobase<sup>1</sup>, an outlier analysis system built for IoT by MIT and Stanford and Cambridge Mobile Telematics, noted several properties of IoT data:

- Data produced by IoT applications often have come from some "ordinary" distribution
- IoT anomalies are often systemic
- They are often fairly rare

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This becomes a data filter which can be attached to a timeseries data stream within a distributed computational framework (i.e. Storm, Spark, Flink, NiFi) to detect outliers.

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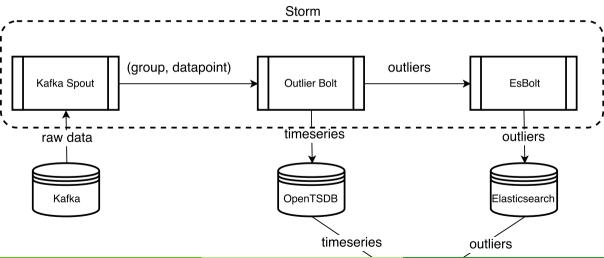
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tl;dr: A formal way to encode our intuition: If a point is far away from the "central" point of our window, then it's likely an outlier.

#### Architecture



# **Demos**

#### Questions

#### Thanks for your attention! Questions?

- Code & scripts for this talk available at http://github.com/cestella/streaming\_outliers
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