

**Anomaly Detection** 

Ingeniería Electrónica







### **Definition**

Anomaly detection is any process that finds the outliers of a dataset; those items that don't belong. These anomalies might point to unusual network traffic, uncover a sensor on the fritz, or simply identify <u>data for cleaning</u>, before analysis.

In today's world of distributed systems, managing and monitoring the system's performance is crucial. With hundreds or thousands of items to watch, anomaly detection can help point out where an error is occurring, enhancing root cause analysis and quickly getting tech support on the issue. Anomaly detection helps the monitoring cause of <a href="mailto:chaos engineering">chaos engineering</a> by detecting outliers, and informing the responsible parties to act.



### **Applications**

In enterprise IT, anomaly detection is commonly used for:

- Data cleaning
- Intrusion detection
- Fraud detection
- Systems health monitoring
- Event detection in sensor networks
- Ecosystem disturbances



### **Anomaly Detection with ML**

"Anomaly detection (AD) systems are either manually built by experts setting thresholds on data or constructed automatically by learning from the available data through machine learning (ML)." Bram Steenwinckel

Machine learning, then, suits the engineer's purpose to create an AD system that:

- Works better
- Is adaptive and on time
- Handles large datasets

Despite these benefits, anomaly detection with machine learning can only work under certain conditions.



### **Anonaly Detection Settings**

In a 2018 lecture, Dr. Thomas Dietterich and his team at Oregon State University explain how anomaly detection will occur under three different settings. They all depend on the condition of the data. The three settings are:

- Supervised
- Clean
- Unsupervised



### **Anonaly Detection - Supervised**





#### Popular ML algorithms for structured data:

- Support vector machine learning
- k-nearest neighbors (KNN)
- Bayesian networks
- Decision trees



### **Anonaly Detection - Clean**



In the Clean setting, all data are assumed to be "nominal", and it is contaminated with "anomaly" points.

The clean setting is a less-ideal case where a bunch of data is presented to the modeler, and it is clean and complete, but all data are presumed to be nominal data points. Then, it is up to the modeler to detect the anomalies inside of this dataset.



### **Anonaly Detection - Clean**



Nomimal and Anomaly datapoints

The datasets in the unsupervised case do not have their parts labeled as nominal or anomalous. There is no ground truth from which to expect the outcome to be. The model must show the modeler what is anomalous and what is nominal.



### K-means

# **Unlabelled Data Labelled Clusters** K-means 🗶 = Centroid



### K-means

Original image



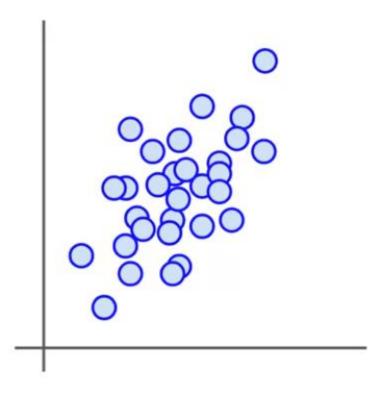
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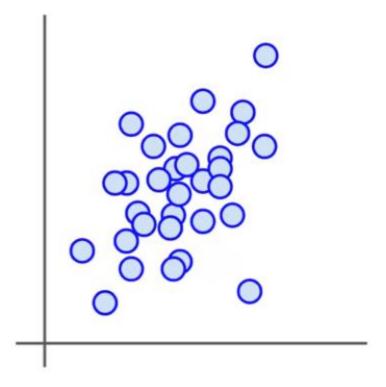
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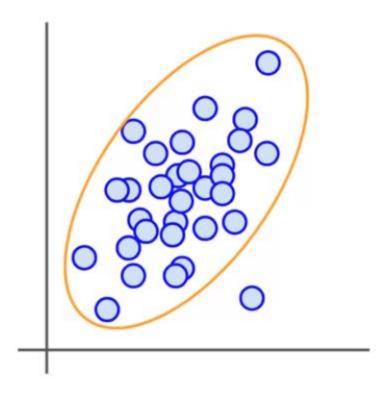




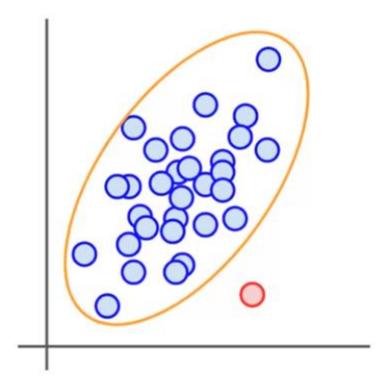








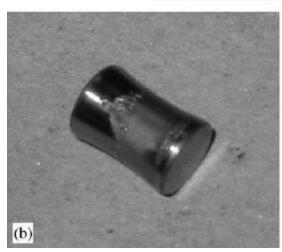






### **Bearing Anomaly**







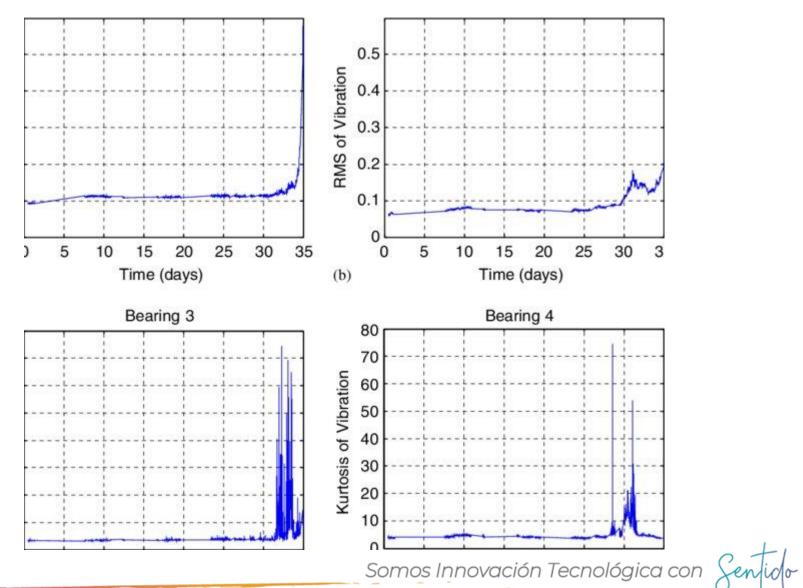
https://www.researchgate.net/publication/223556476 Wavelet filter-

based weak signature detection method and its application on rolling element bearing prognostics/figures}

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# **Bearing Anomaly**



Viailada Mineducación



# 1 Gracias!



