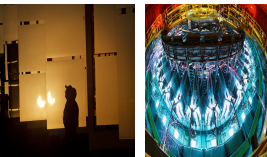


5/10/17



PV Monitoring and Modeling

a machine learning perspective

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Overview

Introduction

What is Machine Learning?

Why use Machine Learning?

How can Machine Learning Provide Value?

Data Monitoring Quality

Rule-based versus Outlier Detection

Performance Validation

... Based on Extensive Data Sets

... Based on Limited Data Sets

Fault Detection & Diagnostics (FDD)

Novelty Detection

Multi-Class Classification

What is Machine Learning?

What?

Machine Learning is the science of programming a computer so that it can learn from data

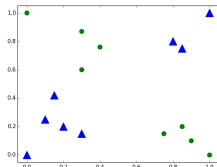
Two main types:

1. Supervised learning:
Inputs w/ desired outputs
2. Unsupervised learning:
Inputs w/out desired outputs

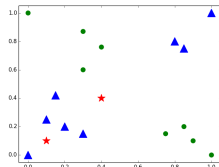
Supervised Learning Example

Classification using Support Vector Machine

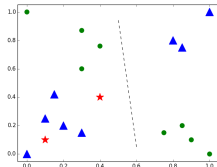
XOR data



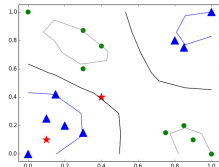
Classify new data



Not linearly separable



Non-linear classification



Why use Machine Learning?

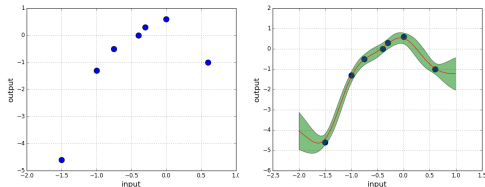
Why?

1. Regression
2. Classification
3. Density estimation
4. Others

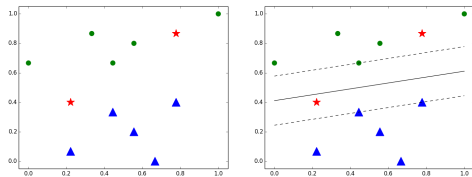
Approaches:

1. Artificial Neural Networks
2. Deep Learning
3. Support Vector Machines
4. Clustering
5. Many others

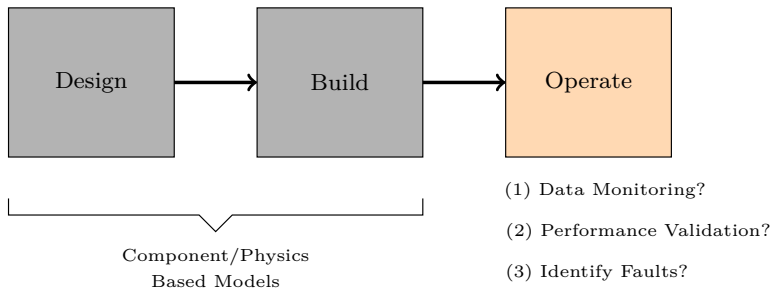
Regression (Gaussian Process)



Classification (Support Vector Machine)



How can Machine Learning Provide Value?



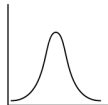
Machine Learning Applications?

Data Monitoring Quality:

Rule-based

```
if True then
  Report
else
  No Report
end if
```

Outlier Detection



Validate Performance:

Extensive Data

Inputs = {E, T_{module}}

Outputs = {Power}

Limited Data

InputsA = {P_{nearby}}

InputsB = {E_{forecast}}

Outputs = {Power}

Fault Analysis:

Fault Detection

Anomaly/Novelty

Detection

Fault Diagnostics

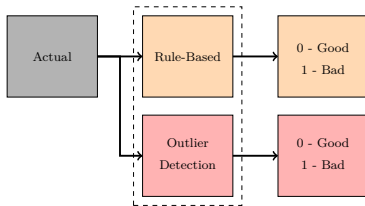
Multi-Class Classification

Data Monitoring: Rule-Based versus Outlier Detection

Problem Statement

Machine learning can use polluted data that contains incorrect or corrupt data records to identify outliers by assuming a distribution.

Process



Results

Rule-Based

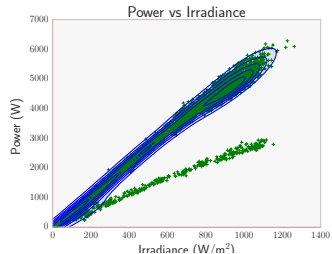
Rules were not violated

```

    if Power > Pmpp × Numbermod. × 1.2 = 7776
    then
      Alarm
    else if Power < 0 then
      Alarm
    end if
  
```

Kernel Density Function

Outliers were detected

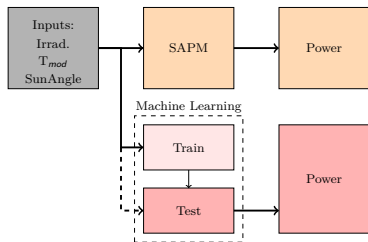


Validate Performance: Extensive Data

Problem Statement

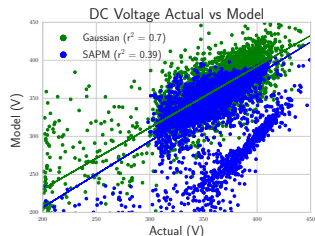
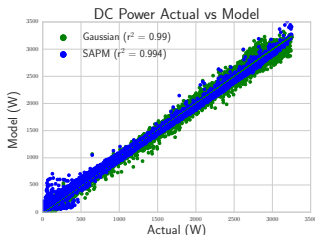
Machine learning can model existing PV systems using collected weather and performance data.

Process



Train: 01/16 to 12/16 -> Test: 01/17 to 04/17

Results

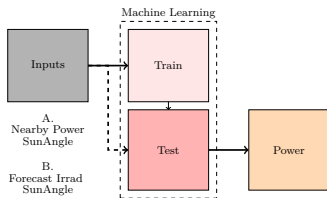


Validate Performance: Limited Data

Problem Statement

Machine learning can model existing PV systems where power is the only monitored value. The algorithm can associate PV power with nearby system outputs and forecasted irradiance.

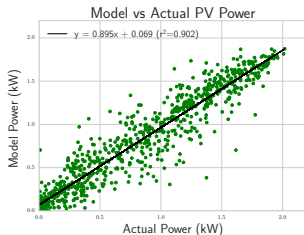
Process



Results

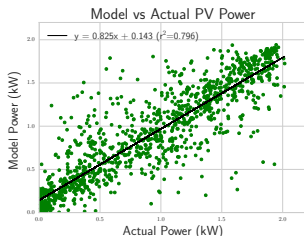
Inputs A:

1. Nearby Power (17km)
2. Sun Angle



Inputs B:

1. Forecast Irrad. (NOAA)
2. Sun Angle



FDD: Novelty Detection (I-V Data)

Problem Statement

Machine learning can be used to perform binary classification of I-V curve data.

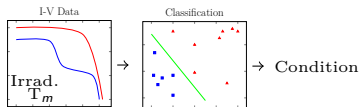
Process

Train & Test

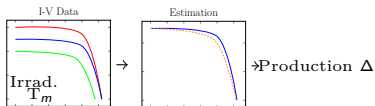
Training: 86 normal and 21 abnormal curves

Testing: 242 normal and 82 abnormal curves

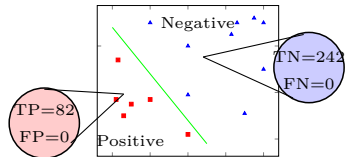
Classification



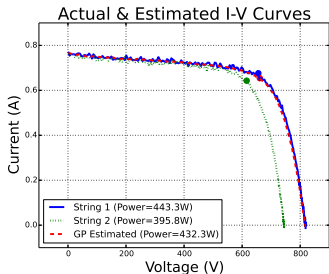
Estimation



Classification



Estimate Loss

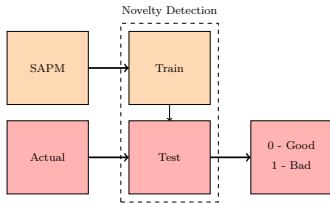


FDD: Novelty Detection (Max Power Point Data)

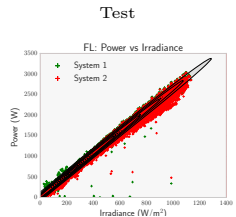
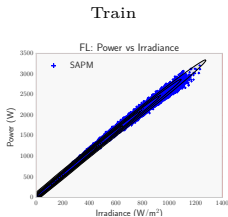
Problem Statement

Machine learning can be used to identify anomalies automatically by training on “clean” data and testing on new, possibly polluted, observations.

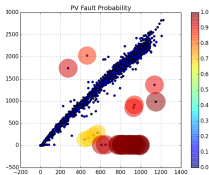
Process



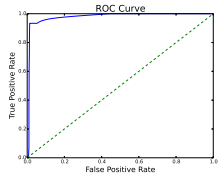
Train & Test Results



Estimate Probabilities



Evaluate Accuracy



FDD: Classification using LAPART

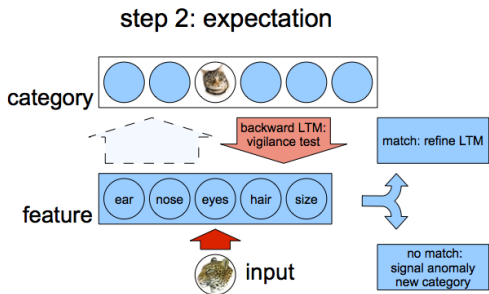
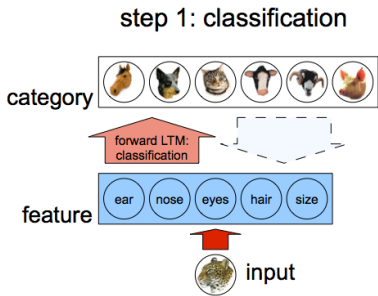
What?

1. Unsupervised
2. Supervised

How?

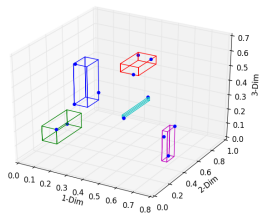
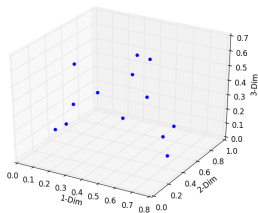
Why?

1. Classification
2. Regression

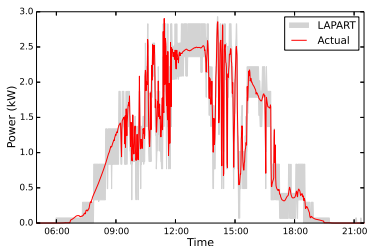


FDD: Classification using LAPART

How does it learn data?



PV Power:

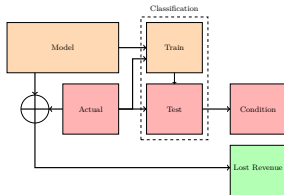


FDD: Classification (Max Power Point Data)

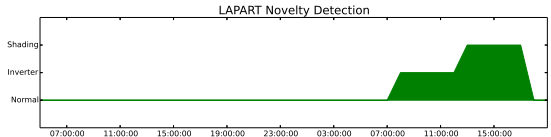
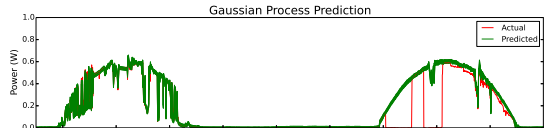
Problem Statement

Machine learning can be used to classify fault conditions and estimate lost revenue.

Process



Results



Summary

Machine Learning

1. What?
 - 1.1 machines can learn
 - 1.2 two main types of learning
2. Why?
 - 2.1 Regression
 - 2.2 Classification
3. How?
 - 3.1 Monitor
 - 3.2 Validation
 - 3.3 Faults

Examples

1. Data Quality
 - 1.1 Kernel Density Function
(python: scipy, scikit)
2. Performance Validation
 - 2.1 Gaussian, SVM, etc
(python: scikit)
3. Novelty
Detection/Classification
 - 3.1 Gaussian, SVM, etc
(python: scikit)
 - 3.2 LAPART (python: coming soon)

Questions

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