



SAND2017-4969C

# 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 they can learn from data

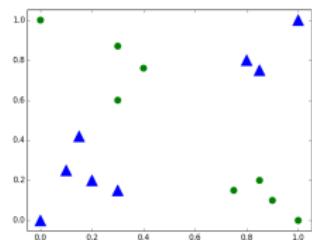
## Two main types:

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

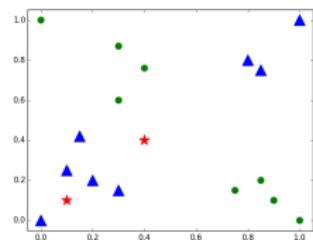
## Supervised Learning Example

Classification using Support Vector Machine

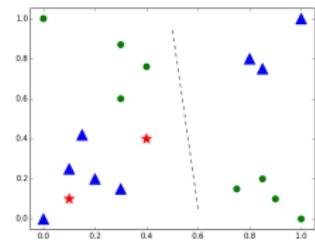
XOR data



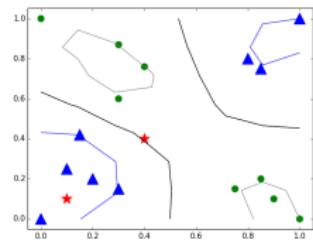
Classify new data



Not linearly separable



Non-linear classification



# Why use Machine Learning?

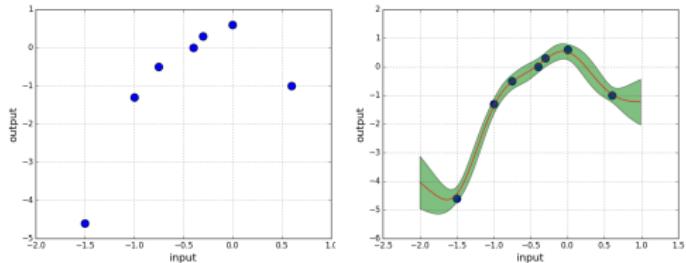
## Why?

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

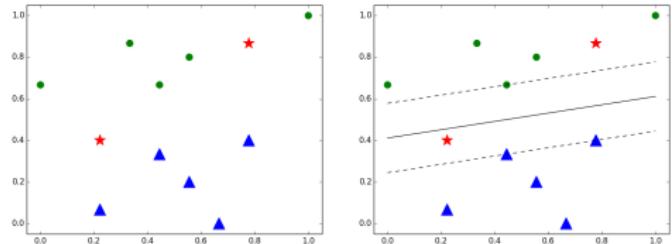
## Approaches:

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?

## Design

1. Specify system
2. Estimate production
3. Confirm budget/financing

## Model Types:

1. SAPM
2. Single Diode
3. Others

## Build

1. System matches design
2. Correct Equipment/Connections
3. Reliable data monitoring

## Model Types:

1. N/A

## Operate

1. Monitor
2. Validate performance
3. Identify faults

## Model Types:

1. Model-based
2. Model and ML
3. ML

# Machine Learning Applications?

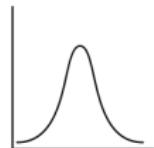
## Data Monitoring Quality:

### 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

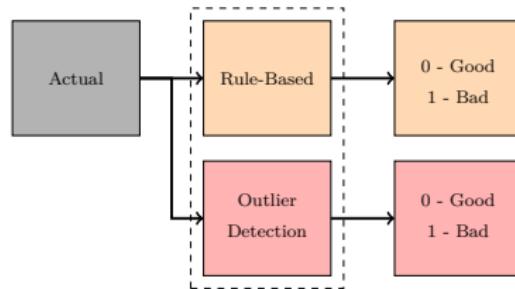
Multi-Class Classification

# 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

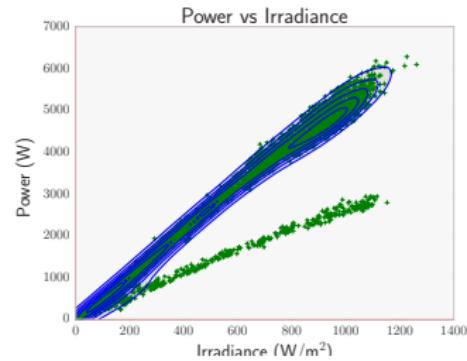
Rules were not violated

```

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

### Kernel Density Function

Outliers were detected

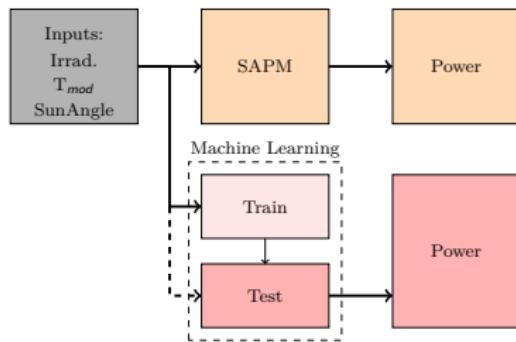


# Validate Performance: Extensive Data

## Problem Statement

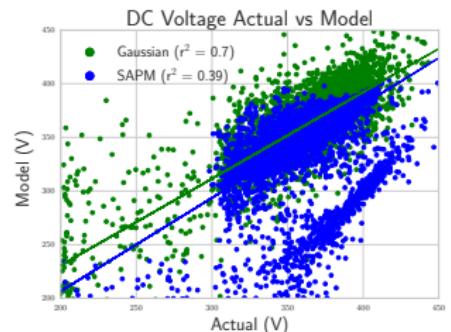
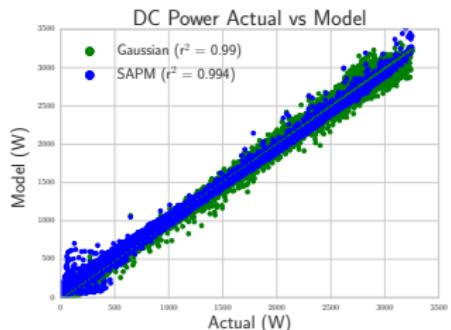
Machine learning can model existing PV systems using collected weather and performance data.

## Process



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

## Results

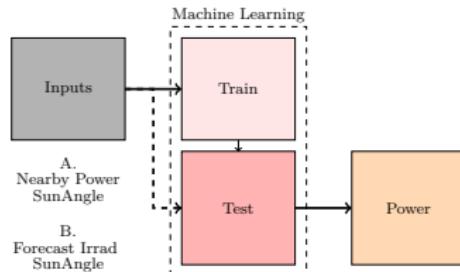


# Validate Performance: Limited Data

## 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.

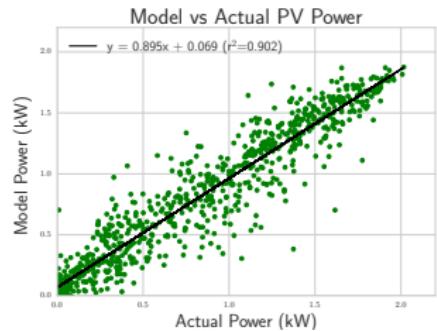
## Process



## Results

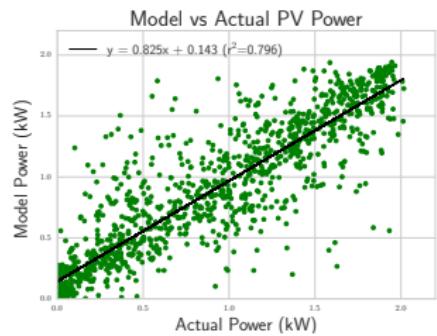
### Inputs A:

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



### Inputs B:

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



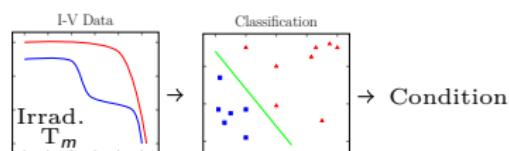
# FDD: Novelty Detection (I-V Data)

## Problem Statement

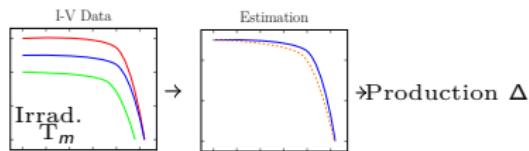
Machine learning can be used to perform binary classification of I-V curve data.

## Process

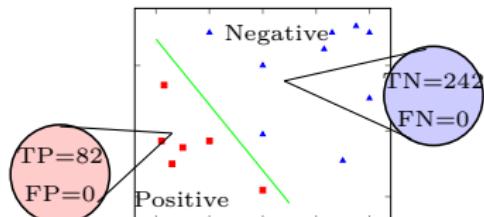
### Classification



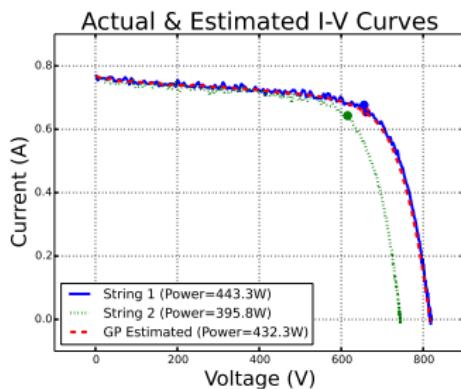
### Estimation



## Classification



## Estimate Loss

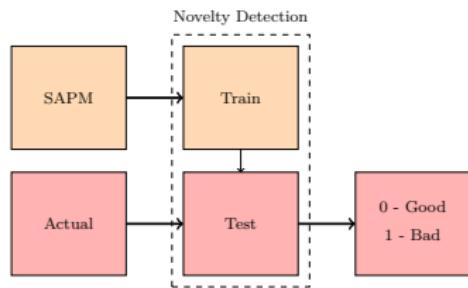


# FDD: Novelty Detection (Max Power Point Data)

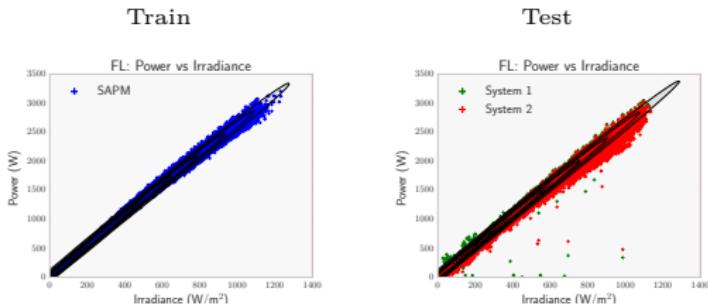
## Problem Statement

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

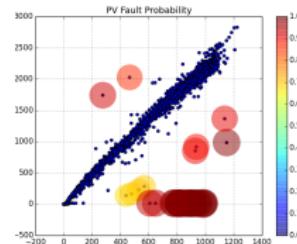
## Process



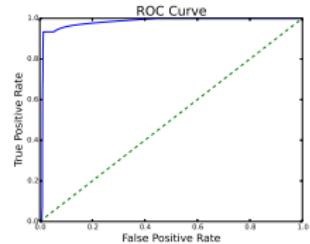
## Train & Test Results



## Estimate Probabilities



## Evaluate Accuracy

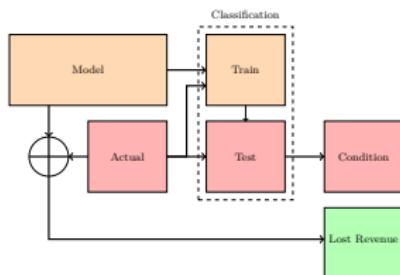


# FDD: Classification (Max Power Point Data)

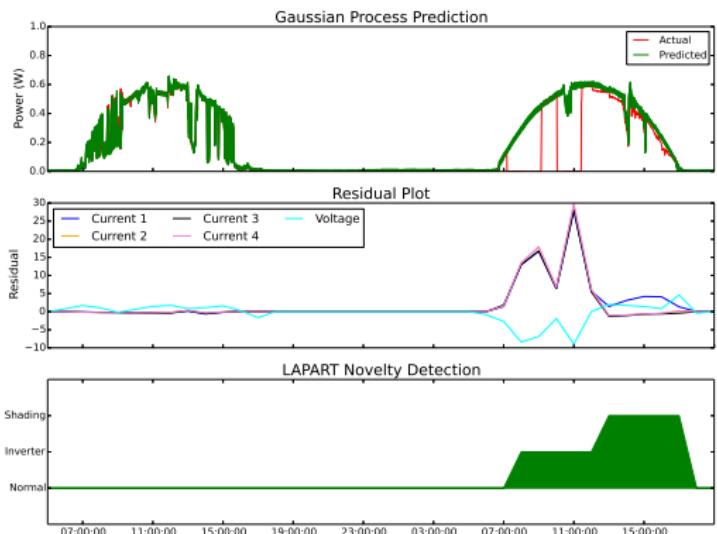
## Problem Statement

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 Outlier Detection provides detailed review of collected data sets

### 2. Performance Validation

- 2.1 ML can model PV power, current, and voltage given various inputs

### 3. Fault Detection

- 3.1 Classify I-V curve and MPP data

# Questions

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