

# Open Source Machine Learning Applied to PV Monitoring Applications

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# Introduction

## Motivation

1. Advanced PV monitoring
2. Embed most accurate Machine Learning Techniques into monitoring systems
3. Low cost embedded devices can be deployed easily
4. Open-source code for multiple applications

## Machine Learning Advantages

1. Rapid adaptable learning, adaptable
2. Design information not needed
3. Accurate results

## Presentation

1. Experiment Setup
2. Machine Learning Techniques
3. Experiment Results
4. Demonstration of embedded system description
5. Conclusions

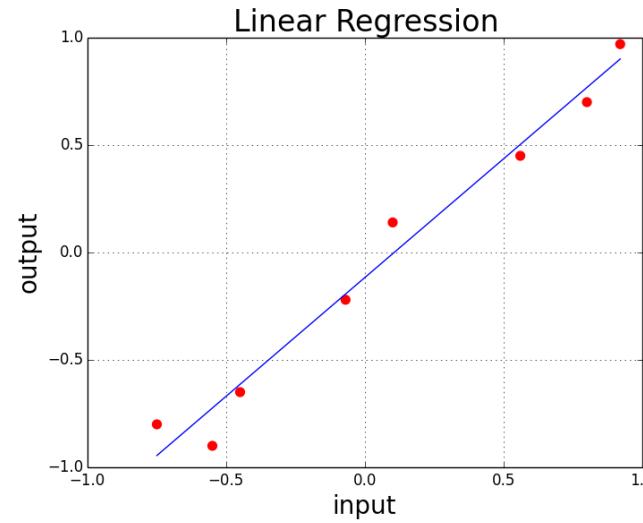


# Experiment Set-up

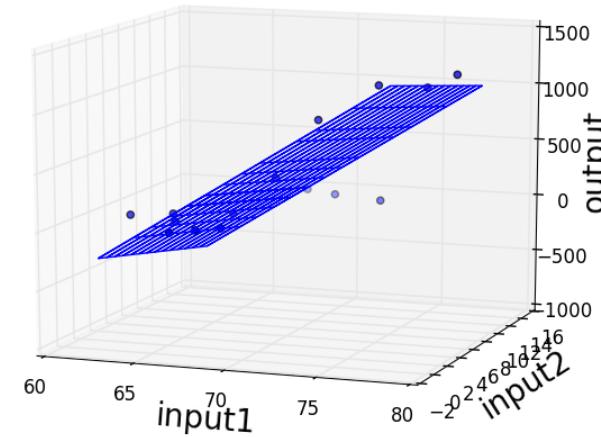
- Purpose
  - Analyze PV data (voltage, current, and power)
  - Apply algorithms to estimate outputs
- Methodology
  - Compare Two Algorithms
    - Least Squares Regression
    - Gaussian Process Regression
  - Training Data
    - Variable test days from 1 – 15 days
    - POA irradiance (pyranometer) and module temperature
    - Data Set Types
      - Complete set (1 minute data)
      - 33% of data randomly selected
      - Histogram bins set based on irradiance and cell temperature statistics (IEC 61853-1)
  - Testing Data is constant
    - 10 days at one minute intervals

# Least-Squares Linear Regression

- Common form of Machine Learning
  - Relationship between:
    - One **dependent** variable and
    - One or more **independent** variables
  - Solution minimizes:
    - $\sum(\hat{y} - y)^2$
  - Gradient Descent that discovers the best:
    - $B_0, \beta_1$
  - Goal fit a function to the data set
    - $f(x, \beta) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots$
- Applied to PVUSA Rating Methodology
  - Present study only uses POA irradiance and module temp (no wind speed)

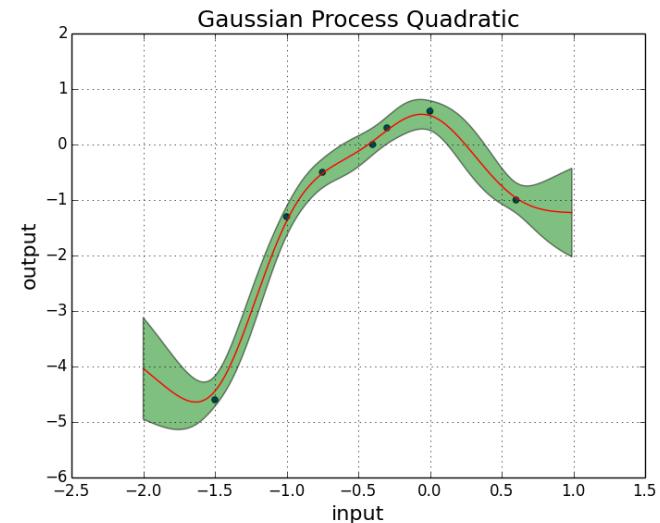
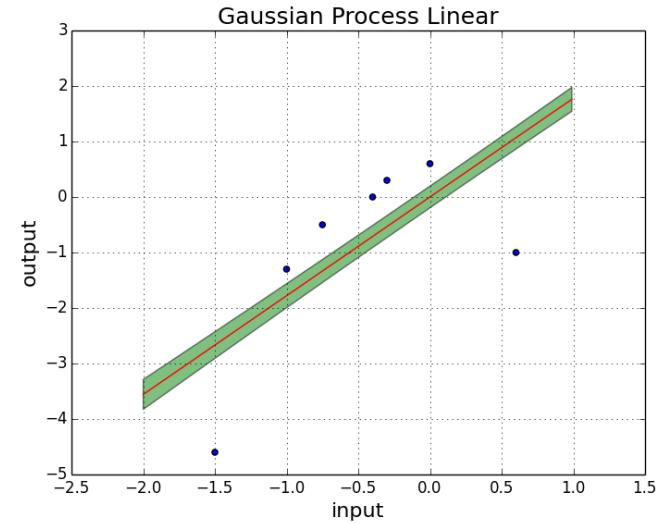


Multivariant Linear Regression



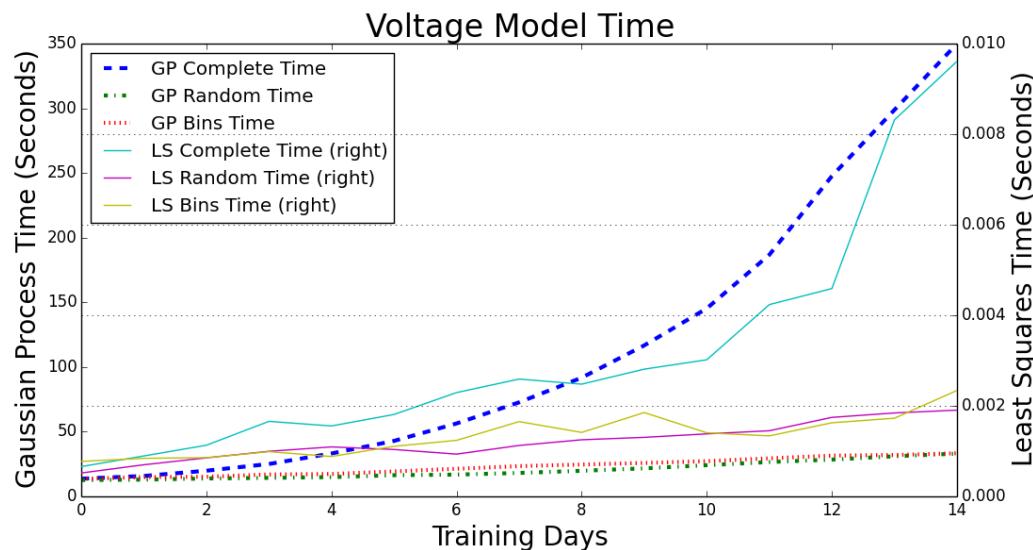
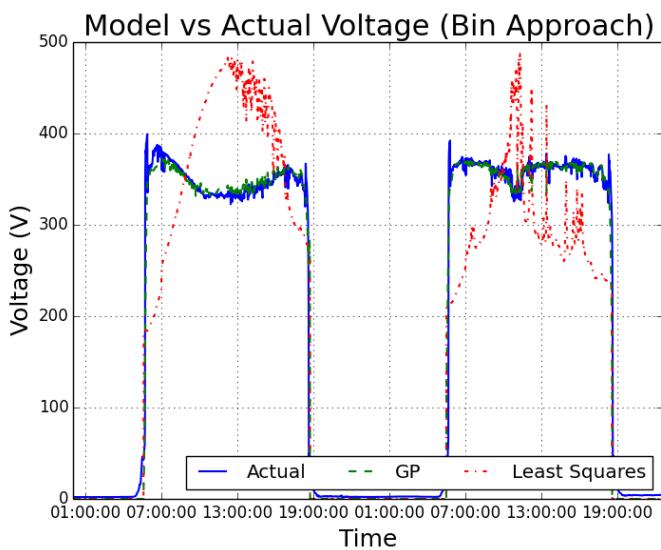
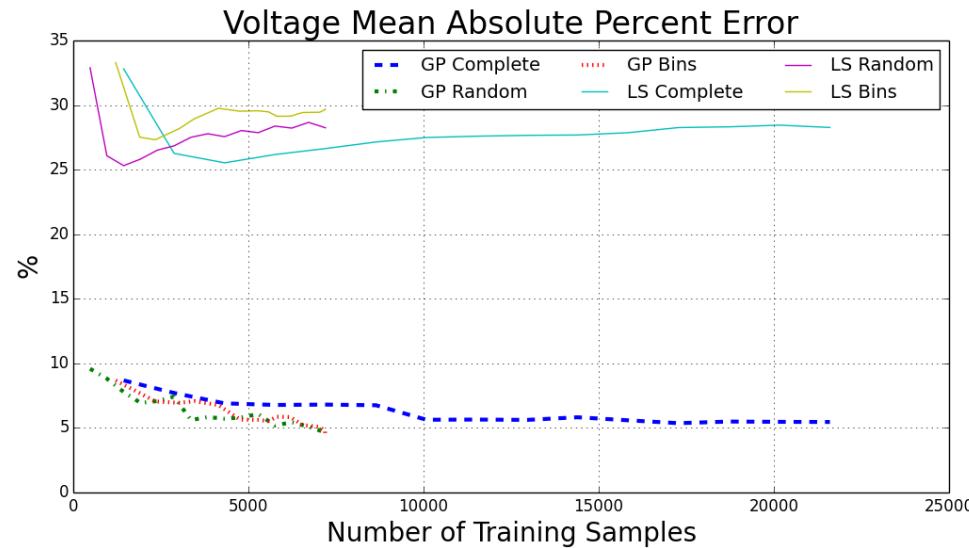
# Gaussian Process Regression

- *GP is the generalization of the Gaussian probability distribution*
- *Parameterized by:*
  - *Mean function –  $\mu(x)$*
  - *Covariance function (kernel) -  $K(x, x')$*
  - $f(x) \sim \mathcal{GP}(\mu(x), K(x, x'))$
- *Gaussian Process Regression*
  - *Predictive distribution*
    - $p(y_*|x_*X, y) = \mathcal{N}(\mu_*, \sigma_*^2)$
  - *where*
    - $\mu_* = K_{*N}(K_N + \sigma^2 I)^{-1}y$
    - $\sigma_*^2 = K_{**} - K_{*N}(K_N + \sigma^2 I)^{-1}K_{*N} + \sigma^2$
    - *and*
      - $K_N$  (*covariance*)
      - $K_{*N}$  (*training set covariance*)
      - $K_{**}$  (*training-test set covariance*)



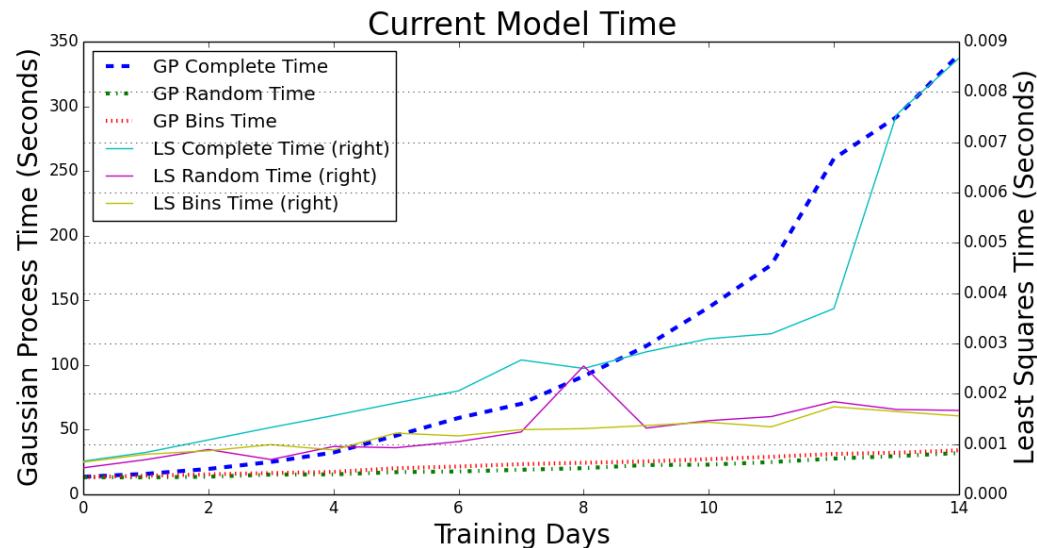
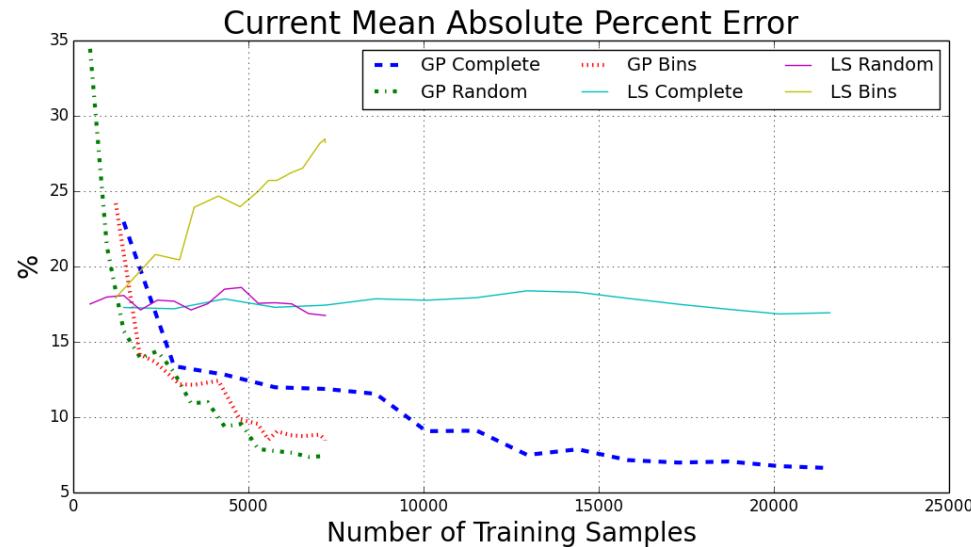
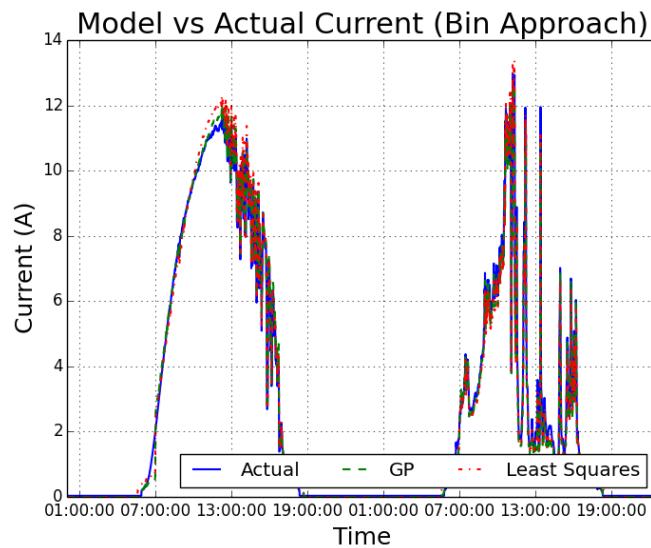
# Voltage Model Results

- GP converges to 5% Mean Absolute Percent Error (MAPE)
- LS MAPE over 25%
- Random and statistical bin sampling have similar results



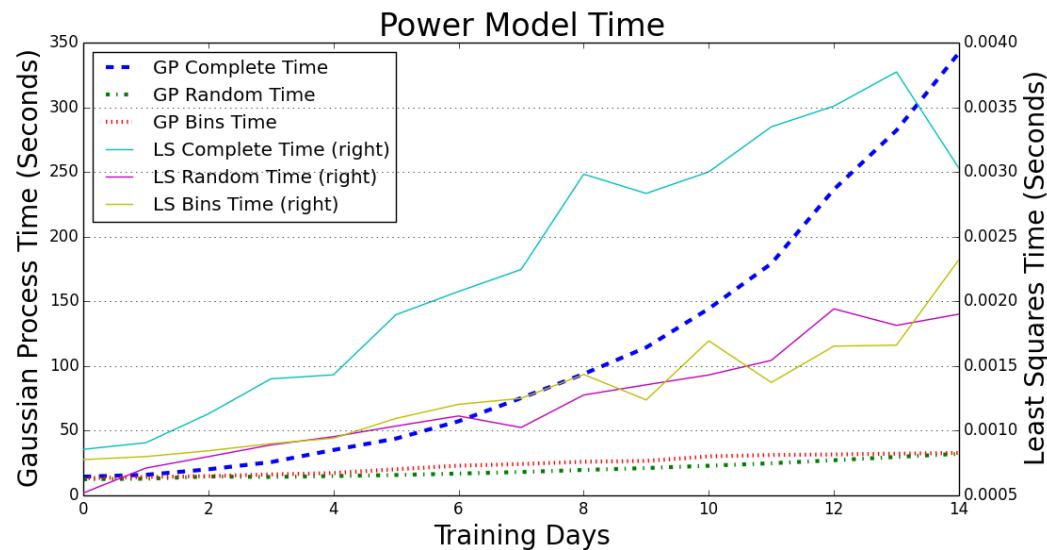
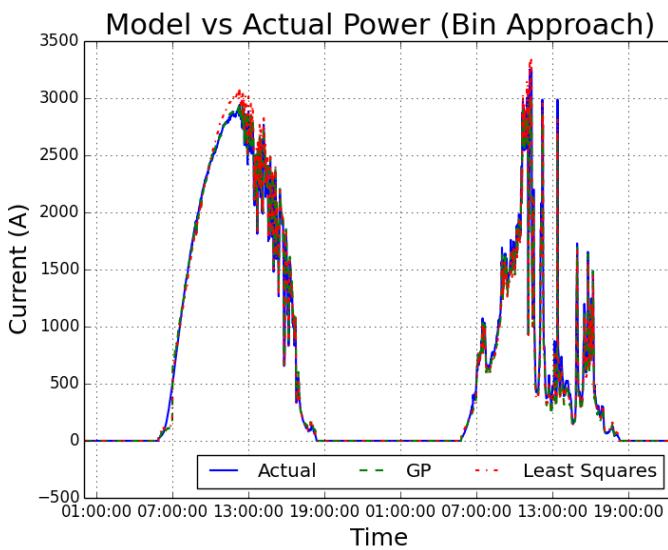
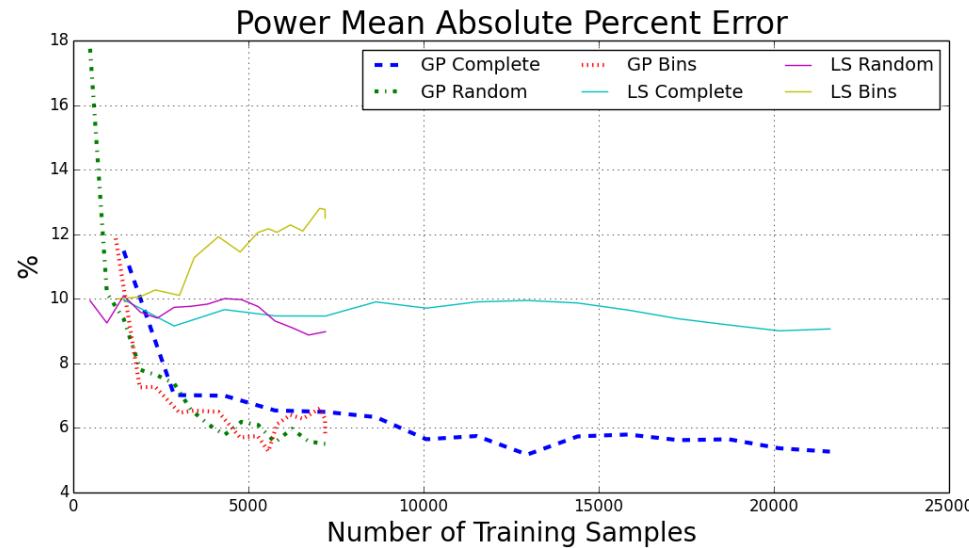
# Current Model Results

- GP converges to 6-7% MAPE
- LS MAPE over 15%
- Random and statistical bin sampling have similar results



# Power Model Results

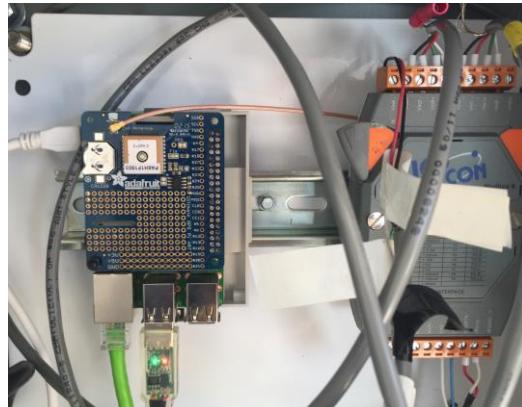
- GP converges to about 5% MAPE
- LS MAPE over 8%
- Random and statistical bin sampling have similar results



# Demonstration

## Embedded Device

- Raspberry Pi
- Modbus connection (Python code)
- MySQL database
- Gaussian Process Estimate (Python code)
- Web-based interface
- Material Cost: < \$300



Query real-time results through web-based interface



# Conclusions & Future Work



## Conclusion

- Gaussian Process provides an accurate and reliable model of PV parameters (Voltage, Current, & Power)
- The algorithm can be embedded into monitoring systems
- Inputs are flexible

## Future Work

- PVSC presentation & paper
- Open source distribution
- Interested partnerships??

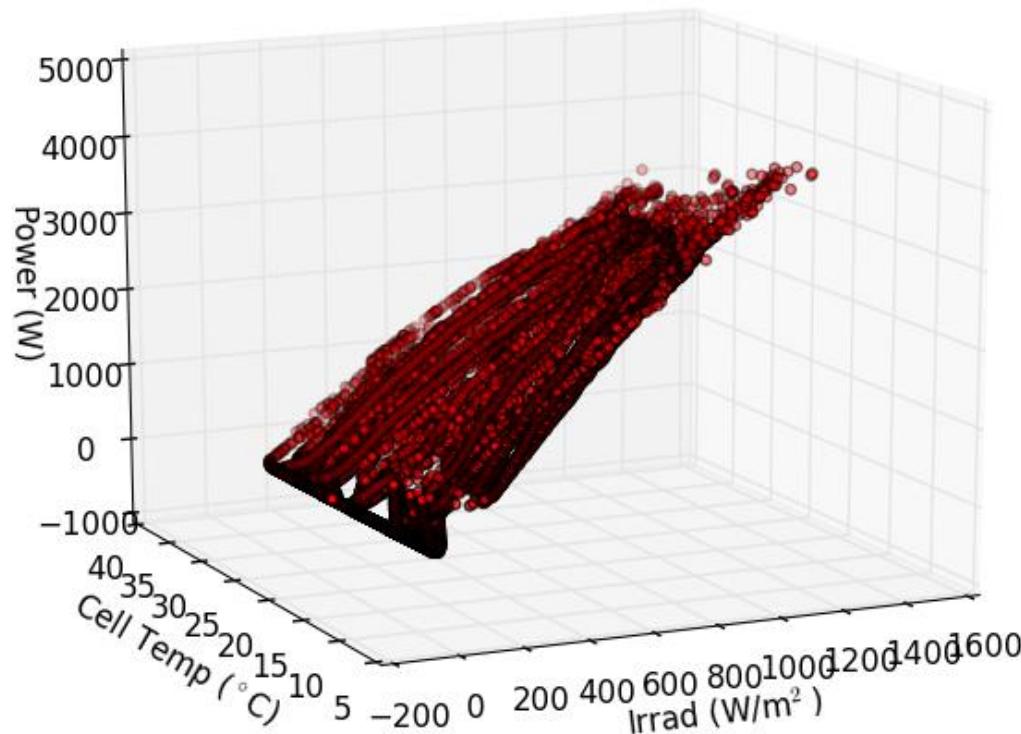
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# Extra



# Extra

