



Sandia
National
Laboratories

Solar Transposition Modeling via Deep Neural Networks with Sky Images



PRESENTED BY

Benjamin G. Pierce, Jennifer L. Braid, Joshua S. Stein, Jim Augustyn, Daniel Riley



Sandia National Laboratories is a multimission laboratory managed and operated by National Technology & Engineering Solutions of Sandia, LLC, a wholly owned subsidiary of Honeywell International Inc., for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA0003525.

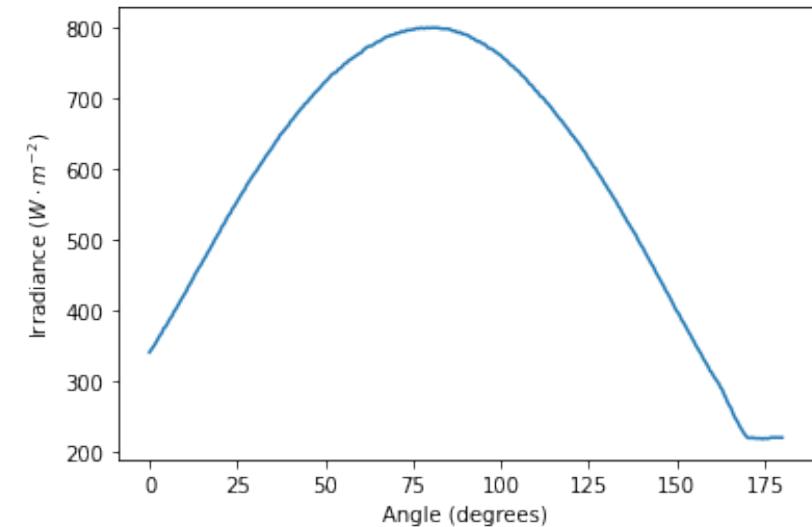
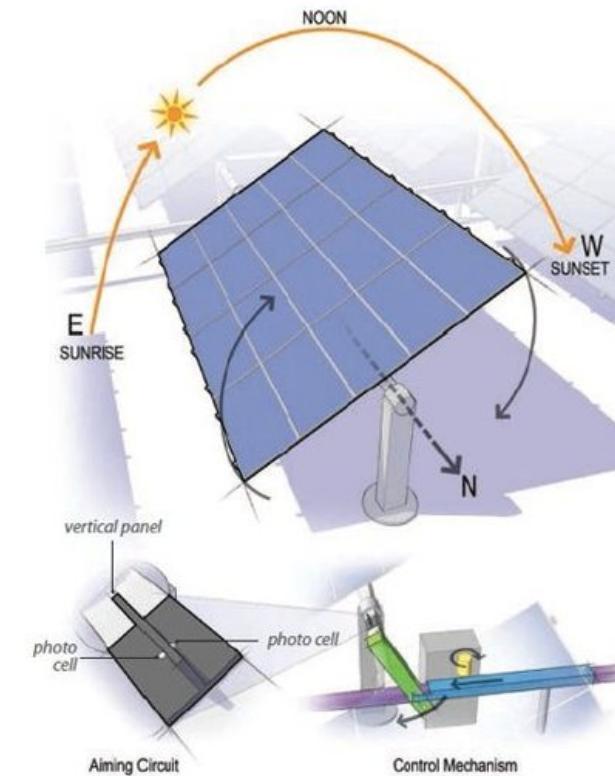
Introduction/premise



- ❖ Objective of the project: make single axis tracking more efficient
- ❖ Align solar panels with Sun
- ❖ Current approach merely follows where Sun should be, not current conditions
- ❖ When clouds are in front of the Sun, panel should be more horizontal

What models need to do:

1. **Estimate angular irradiance profiles**
2. Find optimal *movement* != *angle of max. irrad.* of the tracking system



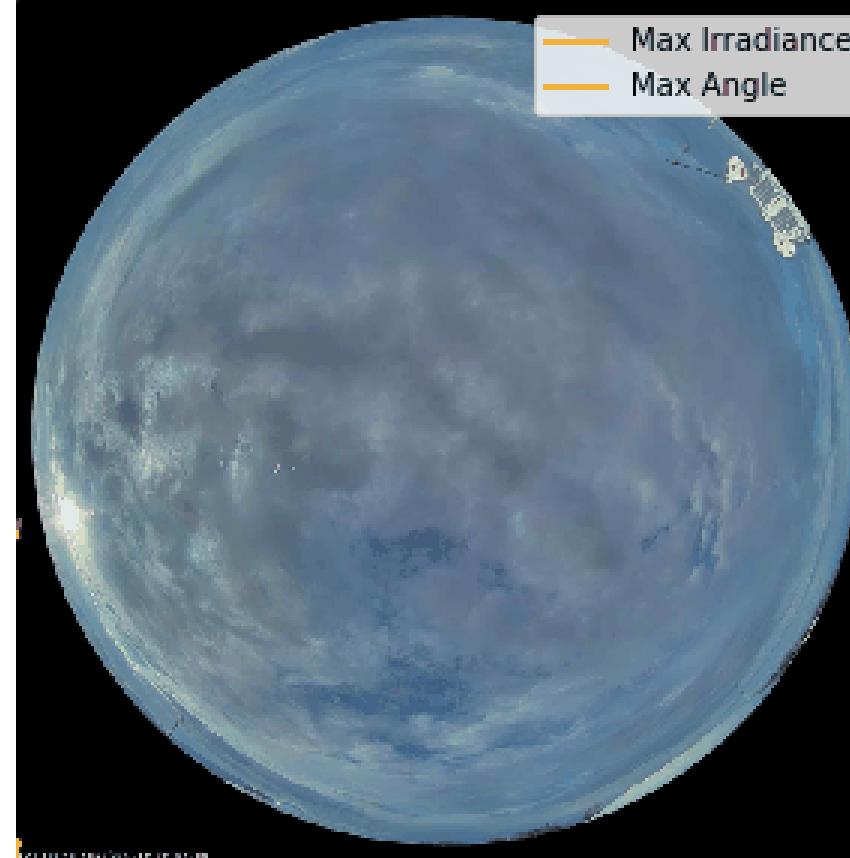
Data Sources



The (custom) MPIS sensor



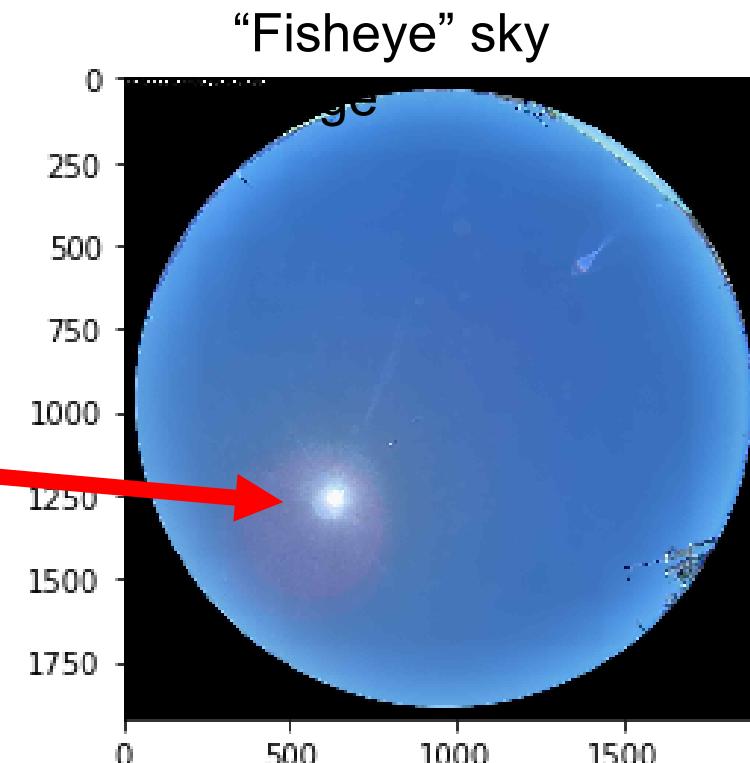
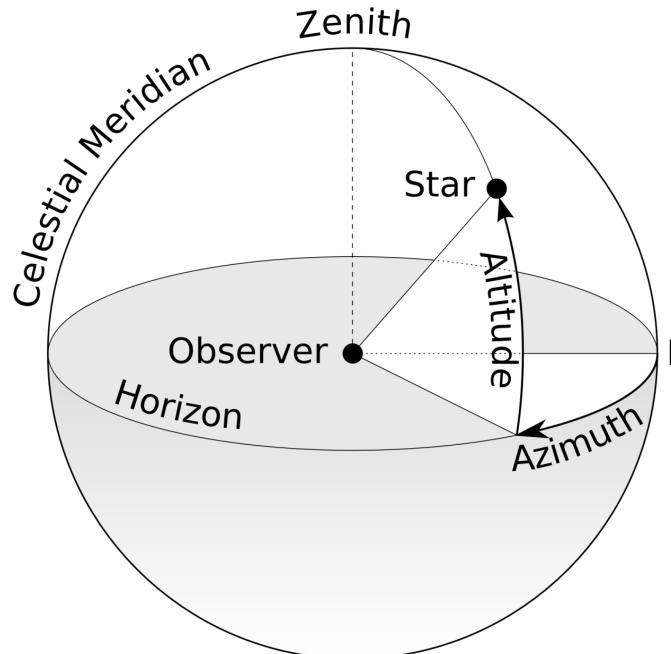
ASI-16 Sky Camera



Model Input: Solar Position + Sky Images



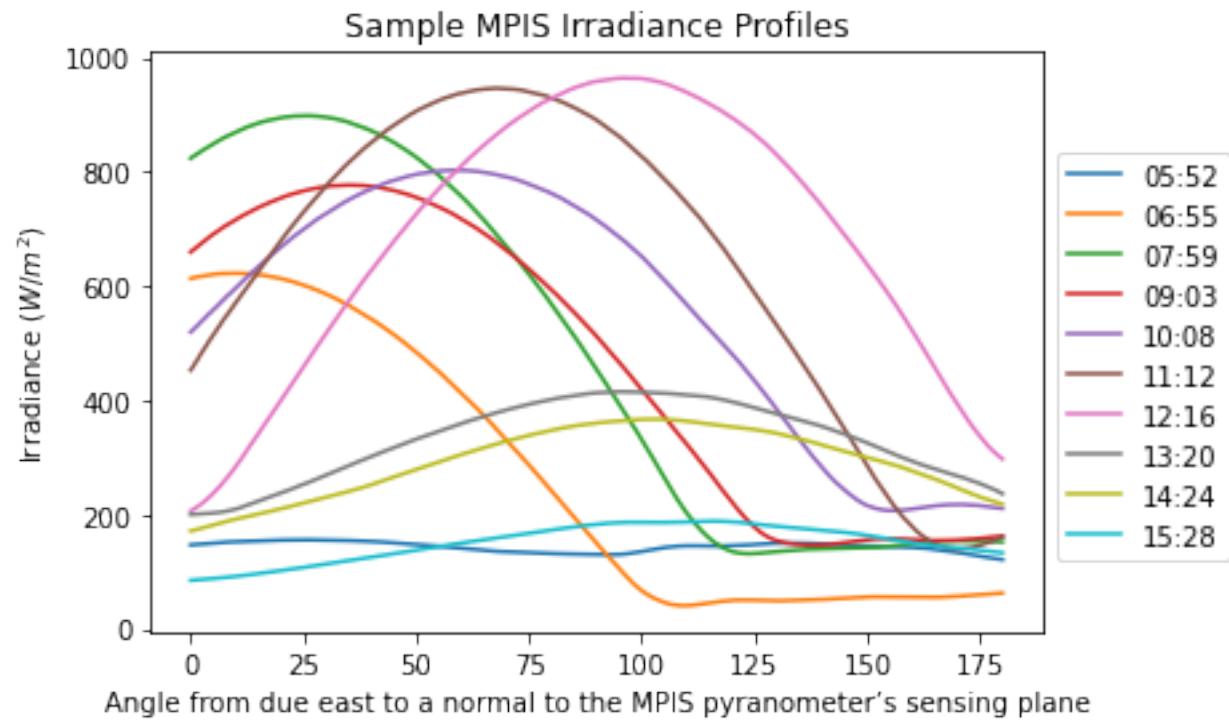
- ❖ The Sun moves **predictably**, and its position can be easily **calculated**
 - ❖ This is useful to know when the Sun is obstructed by clouds
- ❖ This calculation is already implemented in **pvlib**
 - ❖ But not with respect to the **image**
 - ❖ This may be part of what the model learns
 - ❖ Or just a simple irradiance calculation



Model Output: Multi Planar Irradiance Sensor (MPIS)



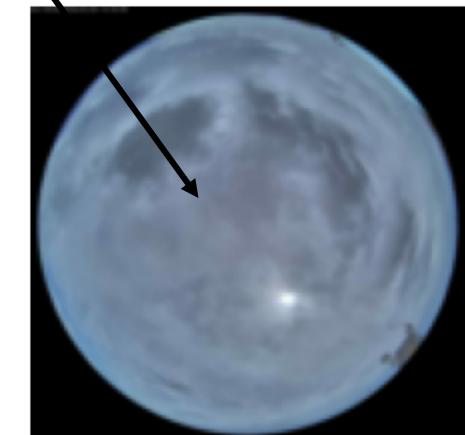
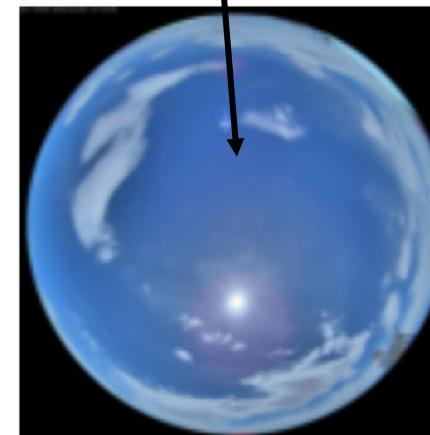
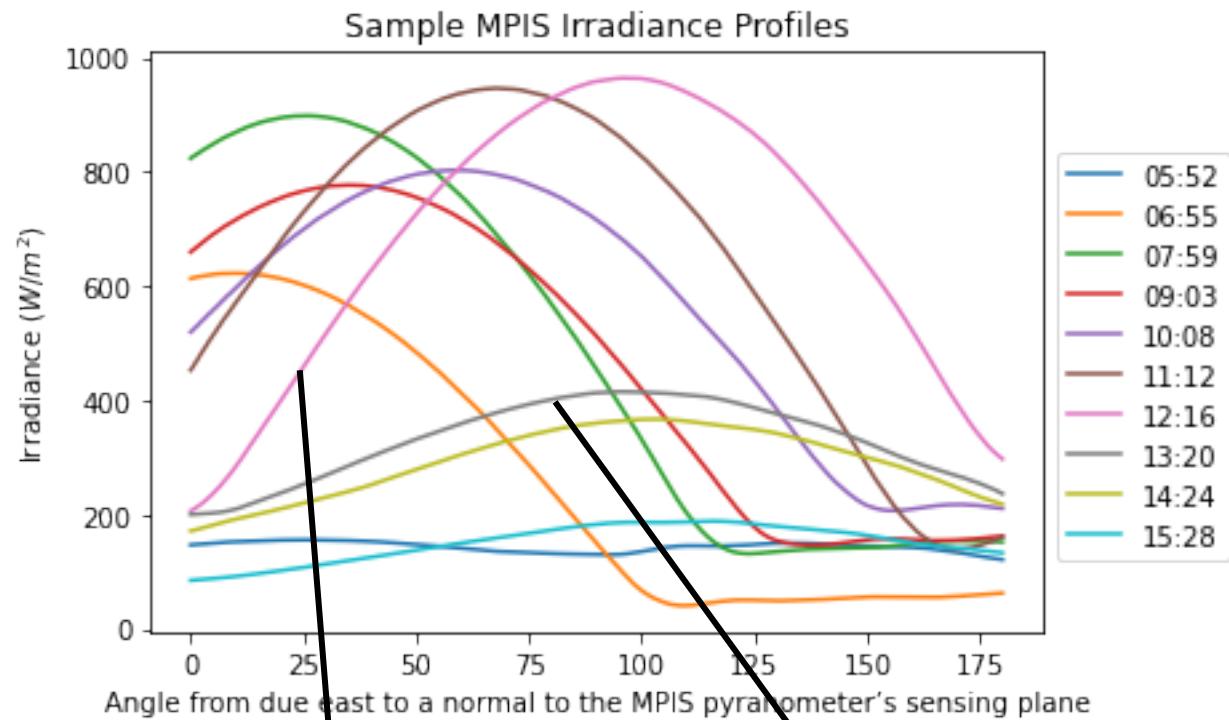
- ❖ Sensor rotates from east to west over about a second
- ❖ Measures in half-degree increments from horizon to horizon
 - ❖ 0-180 degrees
- ❖ Developed by Augustyn & Company, Berkeley, California
- ❖ US Patent #10281552



Model Output: Multi Planar Irradiance Sensor (MPIS)



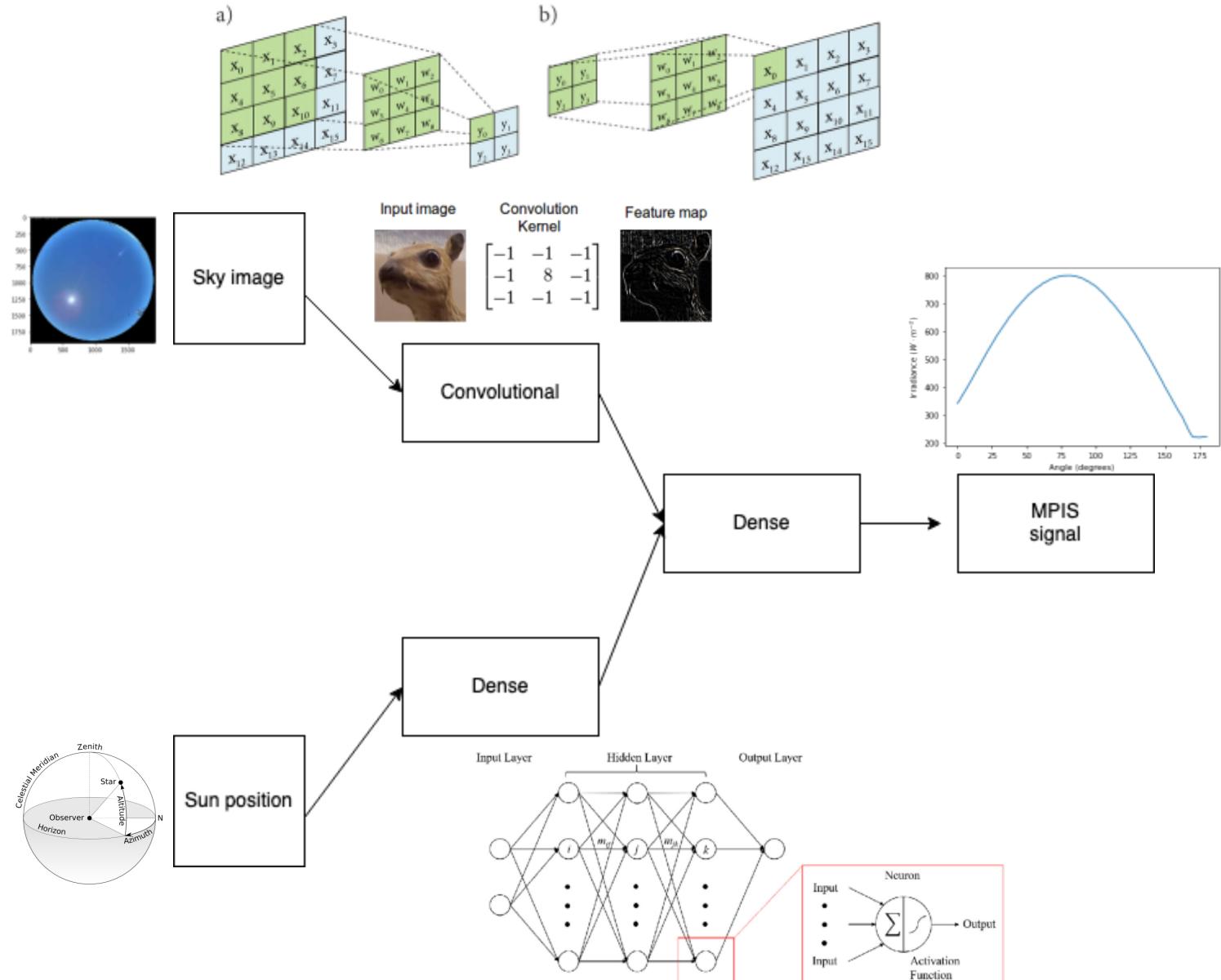
- ❖ Sensor rotates from east to west over about a second
- ❖ Measures in half-degree increments from horizon to horizon
 - ❖ 0-180 degrees
- ❖ Developed by Augustyn & Company, Berkeley, California
- ❖ US Patent #10281552



Multi-Input Convolutional Neural Network

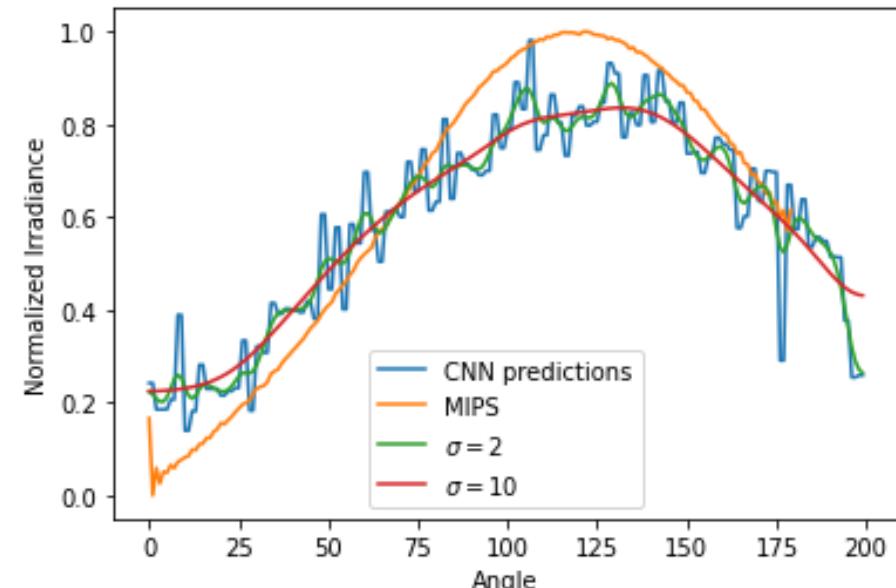
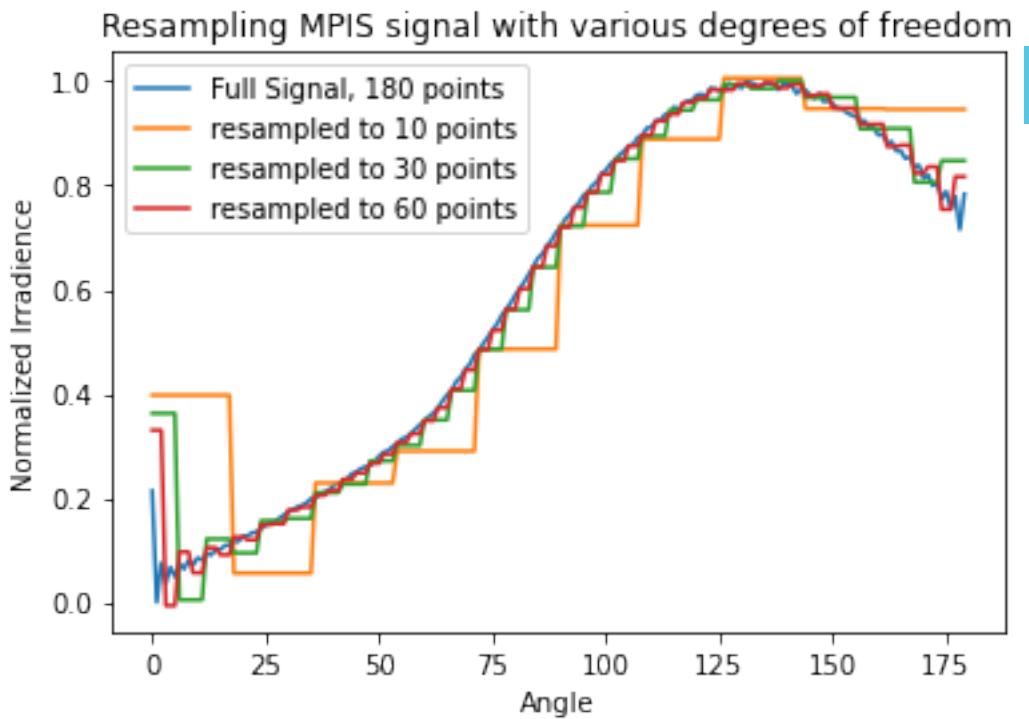


- ❖ This neural network takes **two** inputs
 - ❖ Sky image from **ASI 16**
 - ❖ Solar position (vector of length 6) from **pvlib**
- ❖ And **vector** output (**regression**)
 - ❖ Trained on the **MPIS** sensor readings as the ground truth
- ❖ Full model structure shown in paper
- ❖ Primarily consists of convolutional and dense layers
- ❖ ANN: learns arbitrary function given data



Resampling and Smoothing

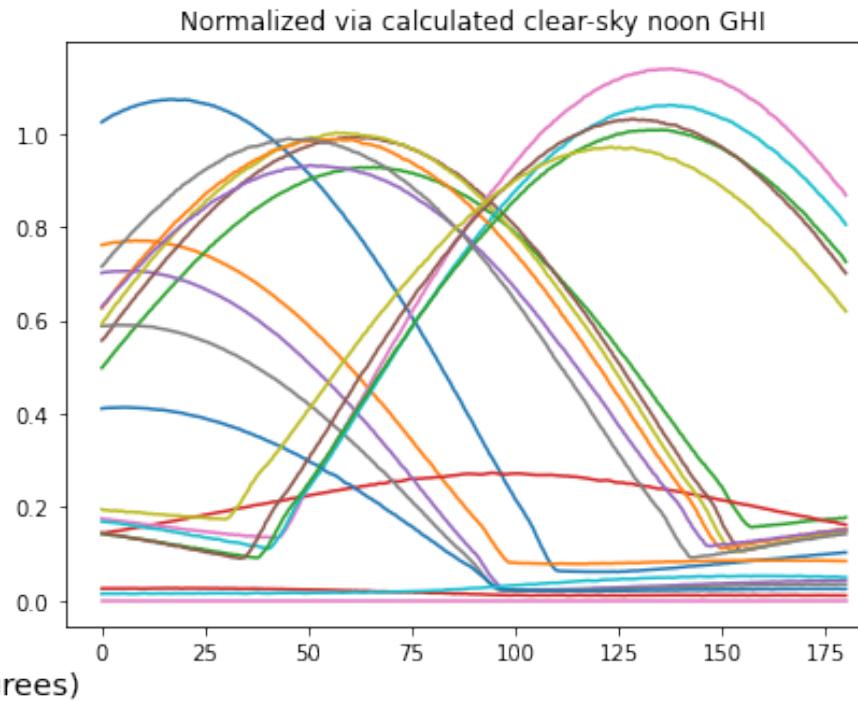
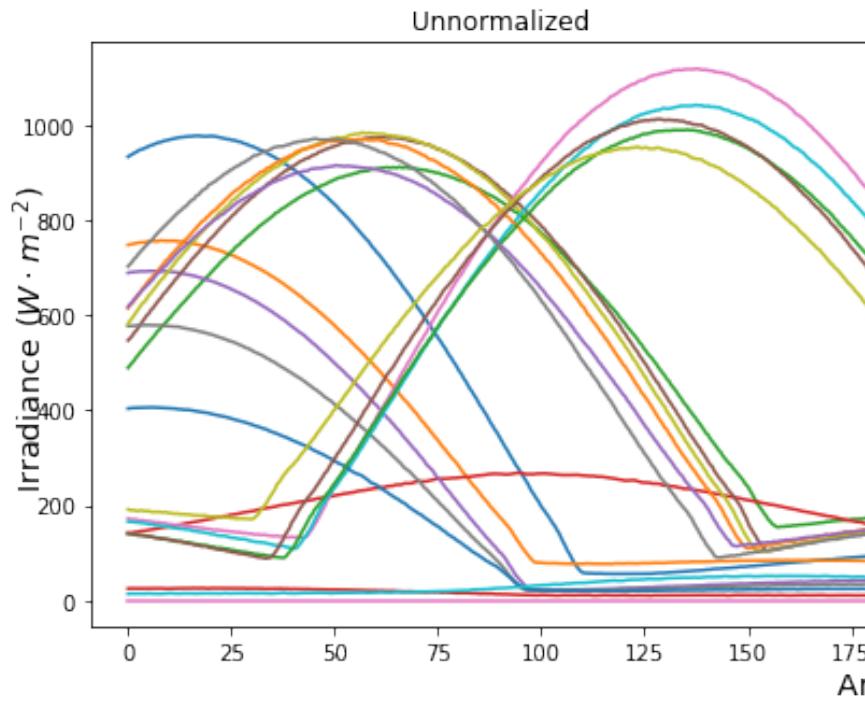
- ❖ It is difficult to model a high dimensional regression problem
 - ❖ Curse of dimensionality
- ❖ However, the MPIS signal is very smooth
 - ❖ Variance between angles is locally small
- ❖ Solution: downsample signal
 - ❖ Using Fourier method
 - ❖ Remove mirrored middle band frequencies
- ❖ This introduces additional source of error/jaggedness
 - ❖ That can be smoothed with Gaussian kernel
 - ❖ Also reduces natural error in model



Normalization



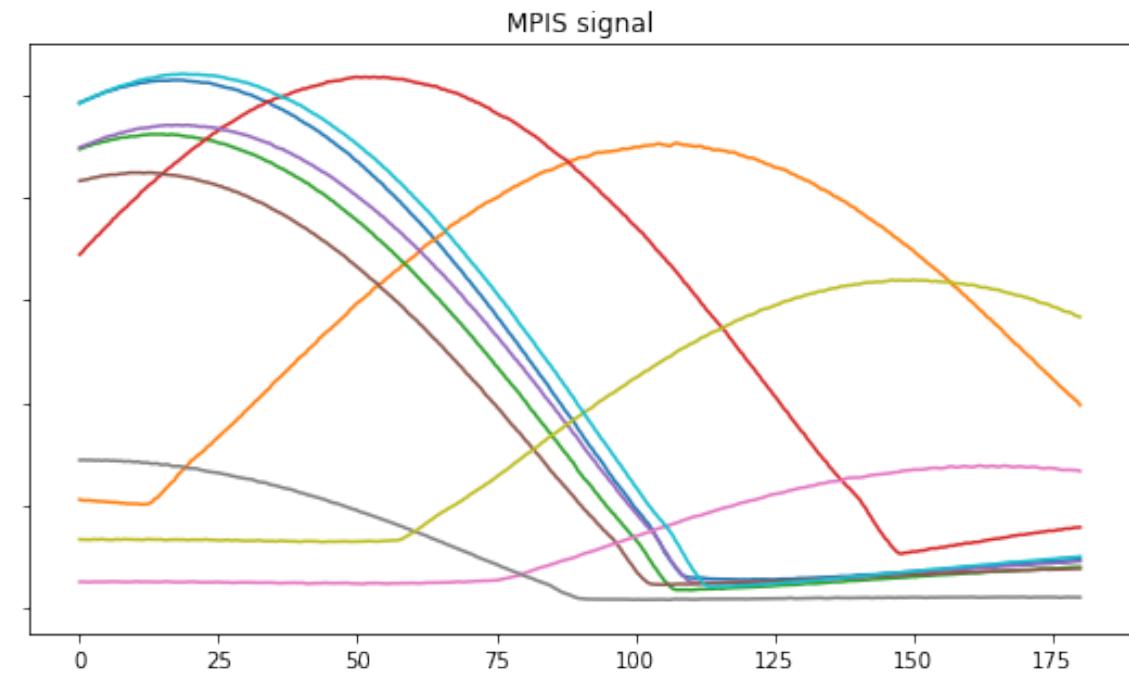
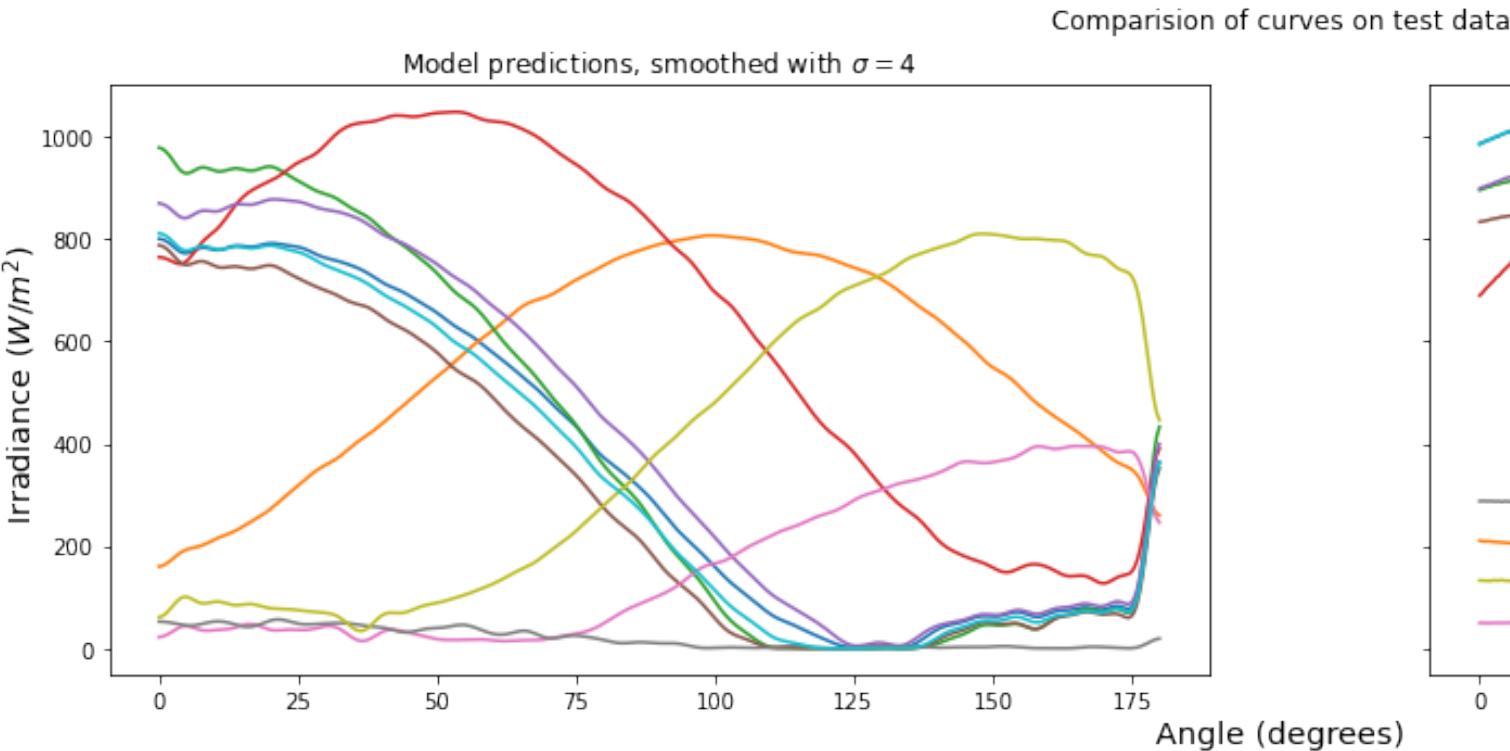
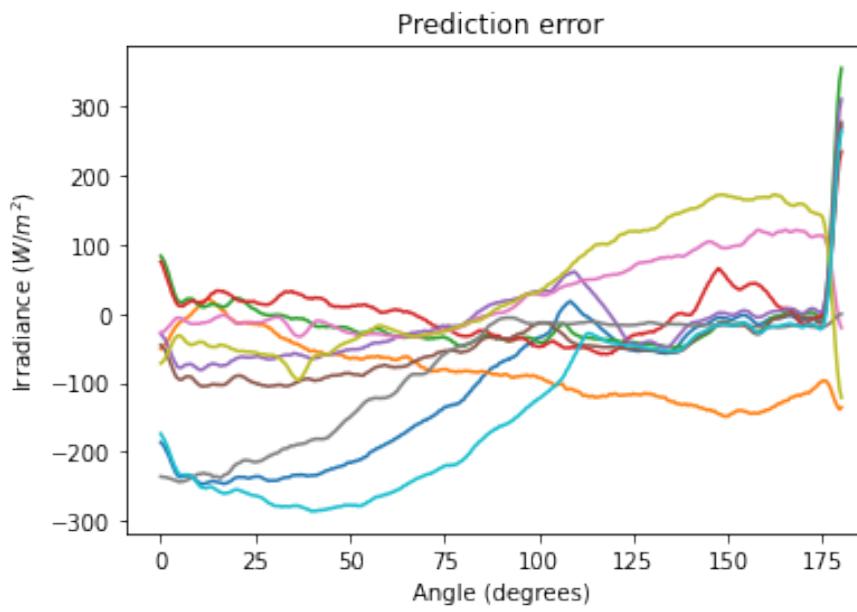
- ❖ The MPIS signal must be normalized in some way to account for seasonality/camera fluctuations
- ❖ Sky image is too qualitative to predict entire irradiance profile directly
 - ❖ But can predict *shape* of the curve
- ❖ Traditional min-max normalization fails to preserve relative magnitudes
- ❖ Solution: normalize by ideal clear-sky GHI at solar noon
 - ❖ Easily calculable with pvlib
 - ❖ Simple rescale operation for days within a week
 - ❖ Still preserves magnitudes as opposed to using true GHI at time t



Results



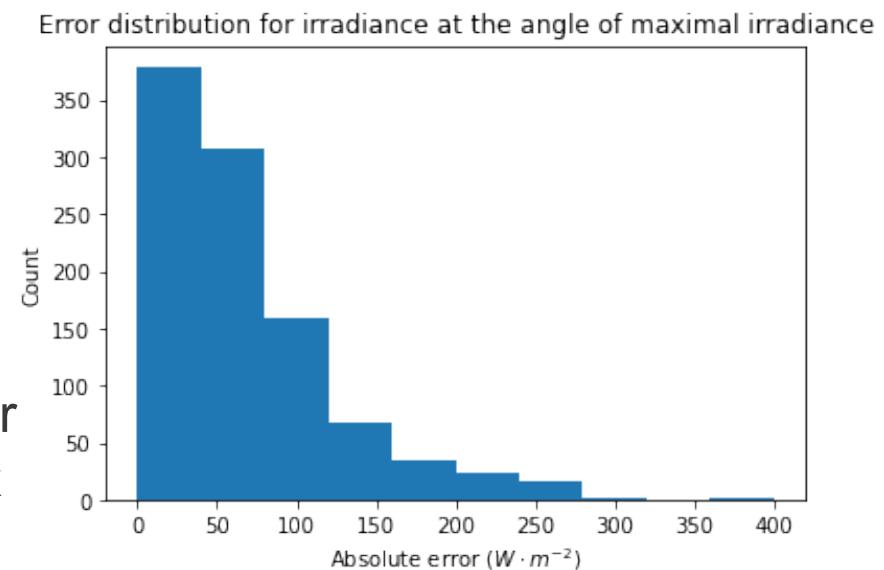
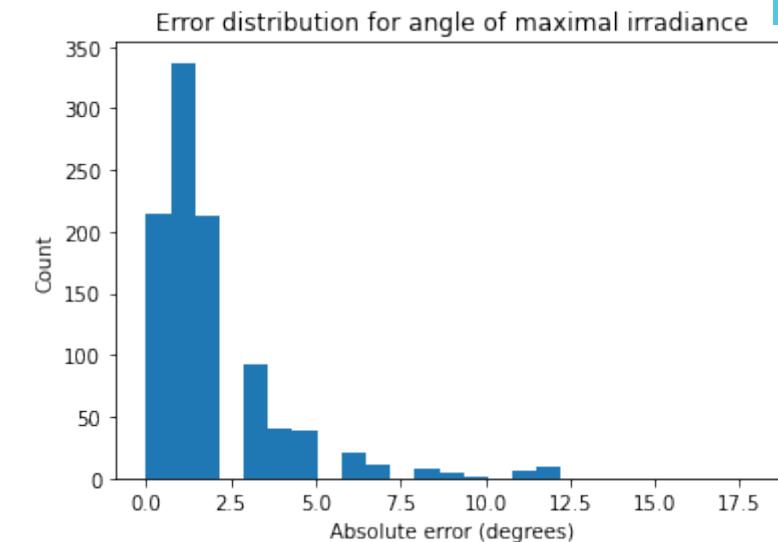
- ❖ Predictions for test samples shown below
- ❖ Model was not trained on these exact samples
- ❖ Most error occurs in less important regions



Conclusion/Future Work



- ❖ Through use of resampling, nonstandard normalization, and other techniques, this high dimensional regression problem becomes feasible
 - ❖ Model correctly learns shape of MPIS irradiance profile and can rescale to absolute irradiance values
- ❖ Overarching project goal: inform movement of single axis tracker
 - ❖ Most important quantity: angle of maximal irradiance
- ❖ Model accurately predicts (within error of 2.5 degrees) over 70% of 1000-sample test set
 - ❖ The remaining samples all fall within ~10 degrees
- ❖ In the future, we are developing algorithms to move a tracker
 - ❖ One prototype: move based on projected change of angle of max irradiance



Helping the neural network



- ❖ Values will be constrained in range $[0, \epsilon]$
 - ❖ Due to normalization process
 - ❖ ϵ constant for dataset, calculated with max
- ❖ Use custom activation function to limit values to this range
 - ❖ Modified ReLu
 - ❖ Linear in range $[0, \epsilon]$
- ❖ Gradient clipping
 - ❖ Truncate gradient when it gets too large
- ❖ Weight decay
 - ❖ “Complexity penalty”
- ❖ Dropout
 - ❖ Viewed as ensembling

