

Duke University Energy Data Analytics Lab

# **Distributed energy resource and infrastructure assessment using overhead imagery**

Kyle Bradbury, Jordan Malof, Bohao Huang, Artem Streltsov, and Leslie Collins

Duke



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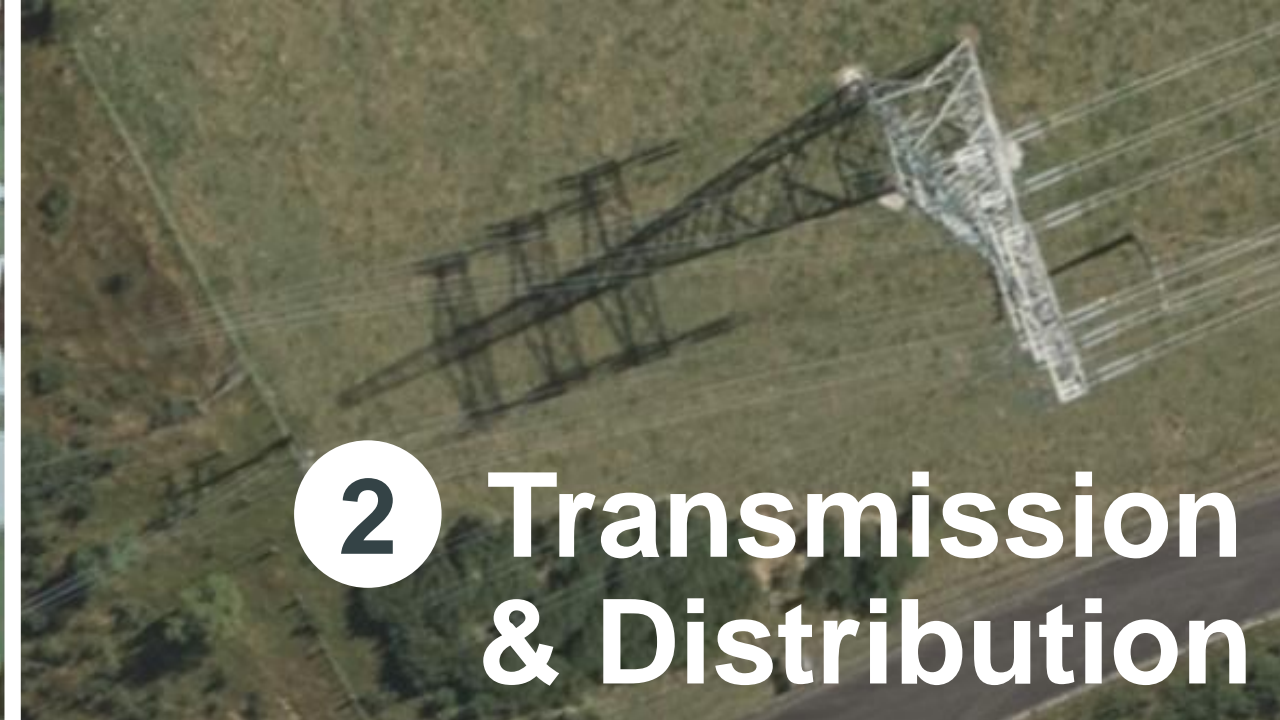
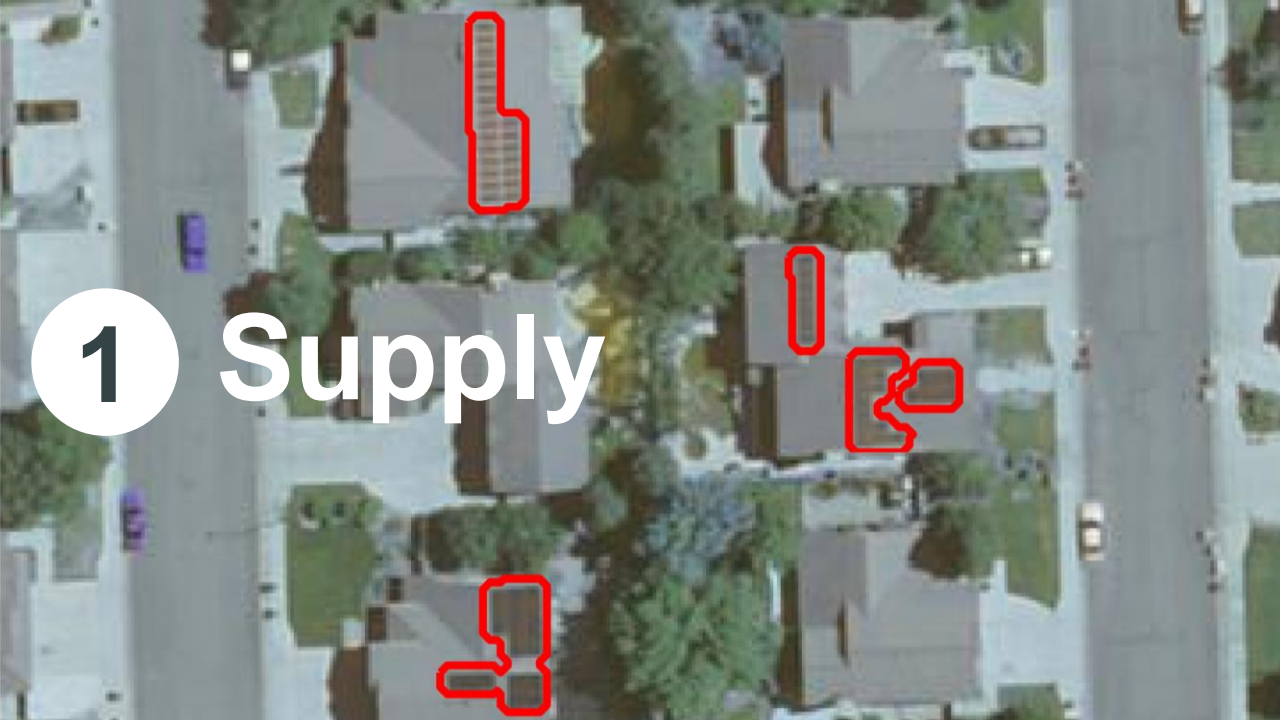
PRATT SCHOOL *of*  
ENGINEERING

# Research **Vision**

**Global**, automated assessment of **energy infrastructure**...

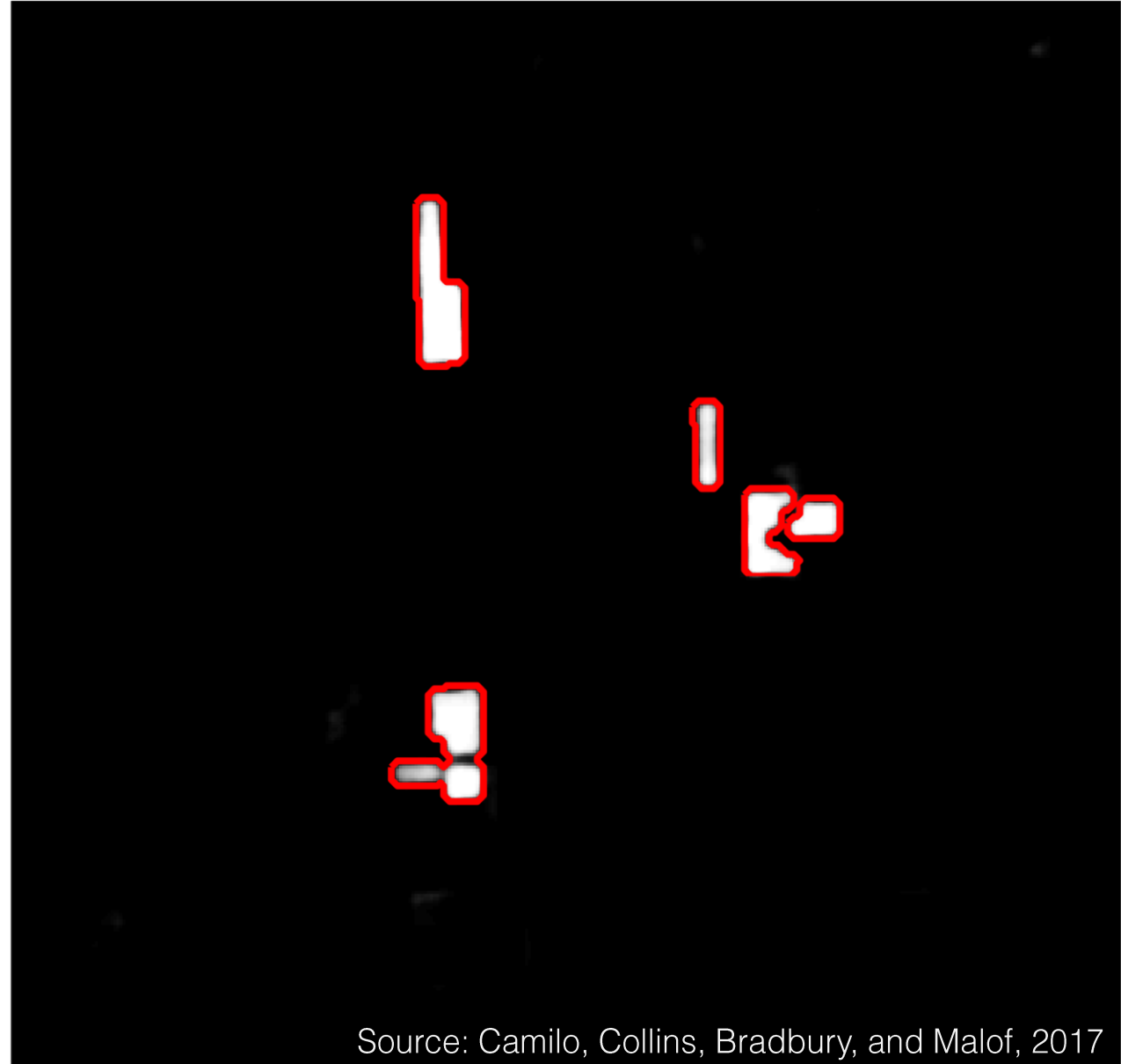
...to develop pathways to **sustainably** meet energy needs.







Original image with ground truth (red)



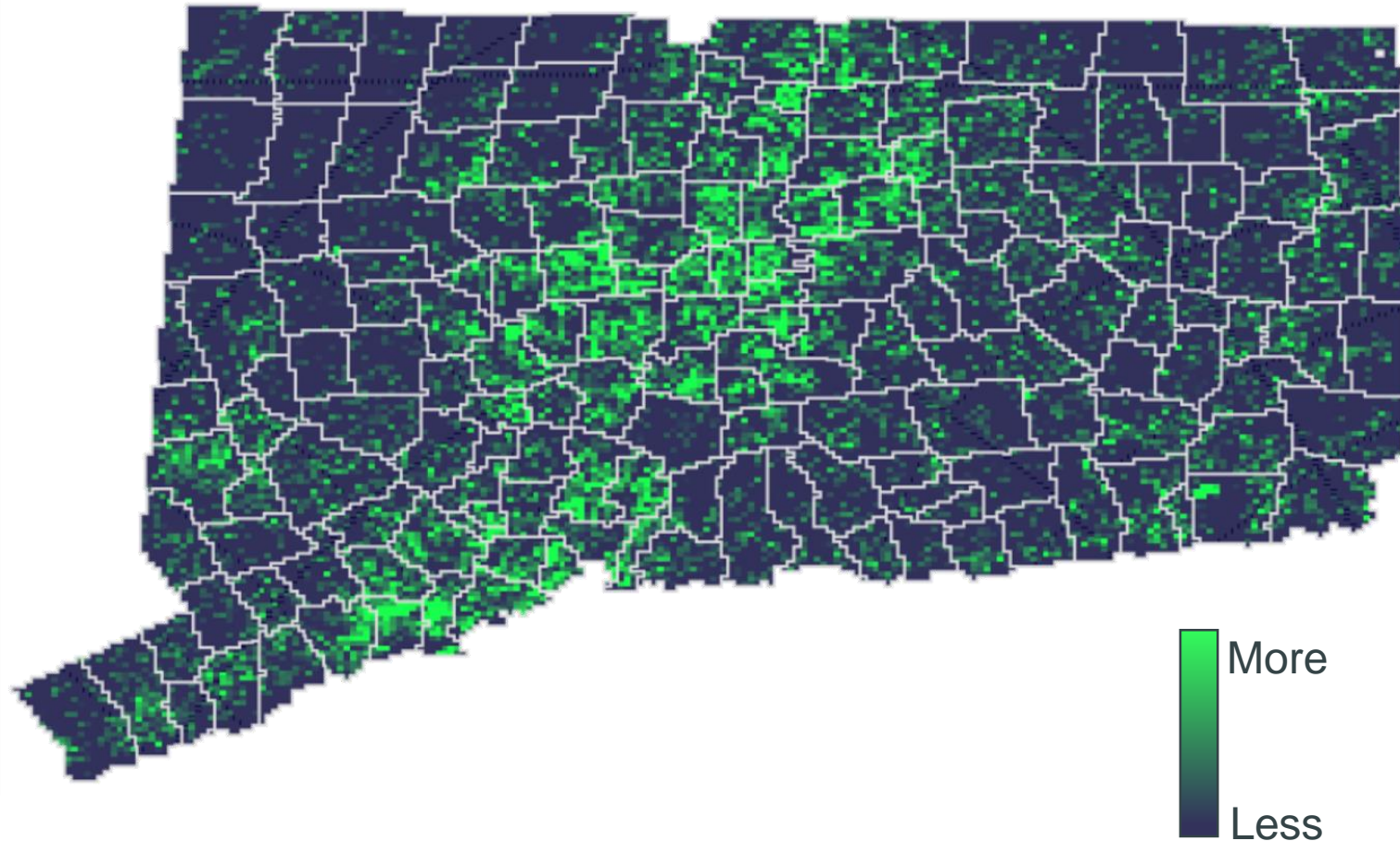
Detected solar arrays

Source: Camilo, Collins, Bradbury, and Malof, 2017

# 1 Distributed energy supply



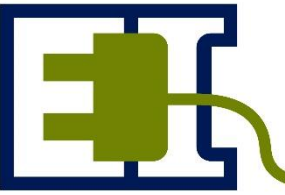
# Goal: estimate solar array **locations**, **power capacity**, and **energy generation**



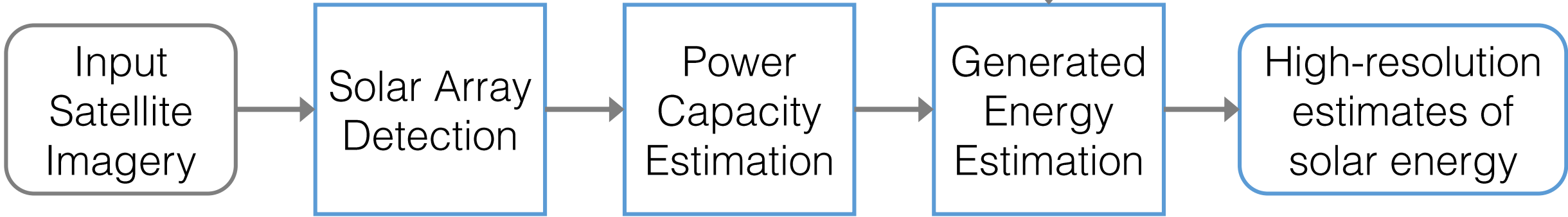
Public estimates of distributed solar are generally limited to state or national scales (U.S. EIA)



# Estimating Energy from Distributed Solar Using Satellite Imagery



Process



Example

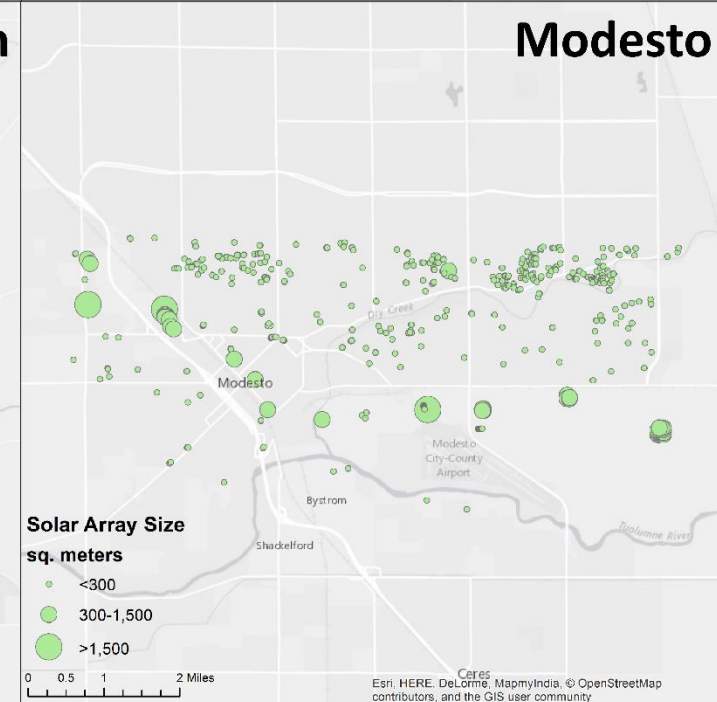
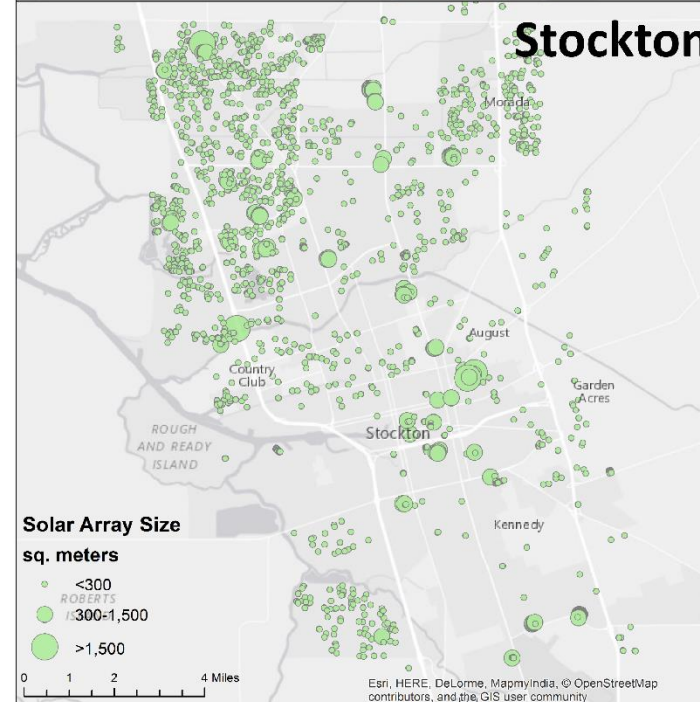
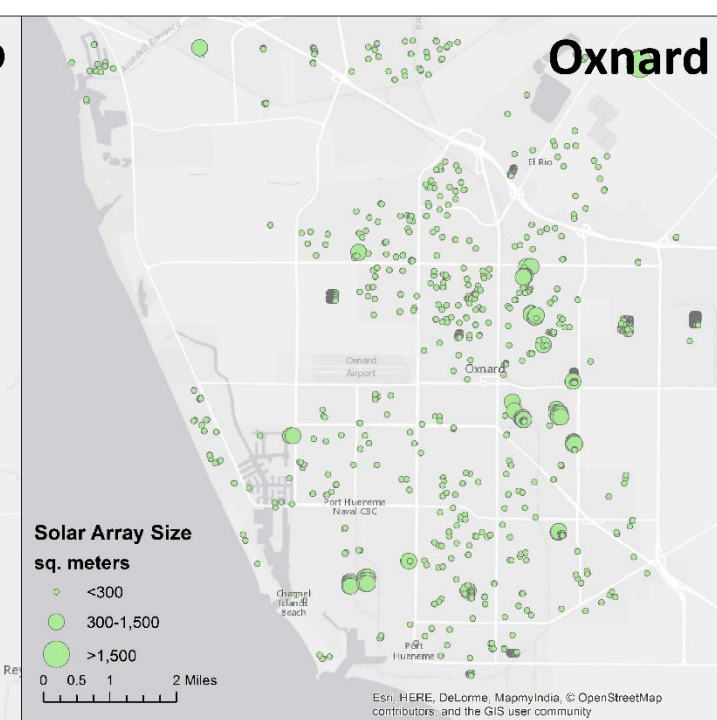
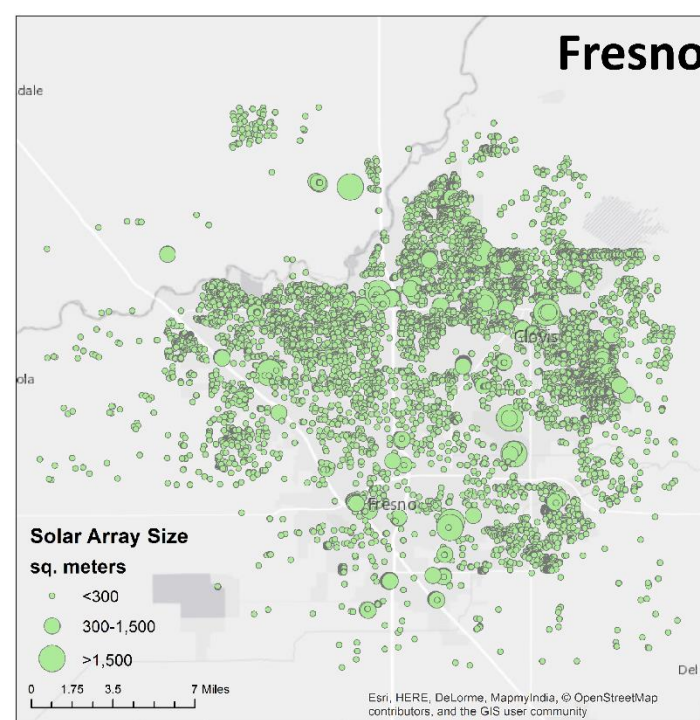


# Solar PV Data

Dataset created by summer 2015 Data+ team and published in Scientific Data

19,000+ manually annotated solar arrays from 4 CA cities

Can be used to train deep learning techniques



Bradbury, K., Saboo, R., Johnson, T., Malof, J., Zhang, W., Devarajan, A., Collins, L., Newell, R. (2016) "Distributed Solar Photovoltaic Array Location and Extent Dataset for Remote Sensing Object Identification." Scientific Data, 3. doi:10.1038/sdata.2016.106.

Available for download at:

<https://doi.org/10.6084/m9.figshare.c.3255643.v2>

# Convolutional Neural Networks

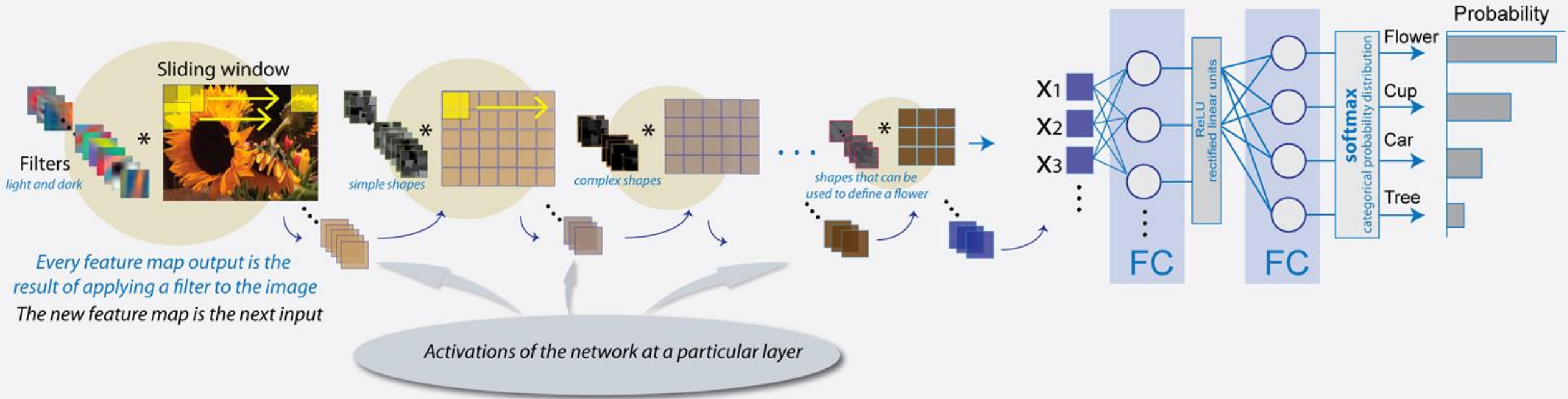
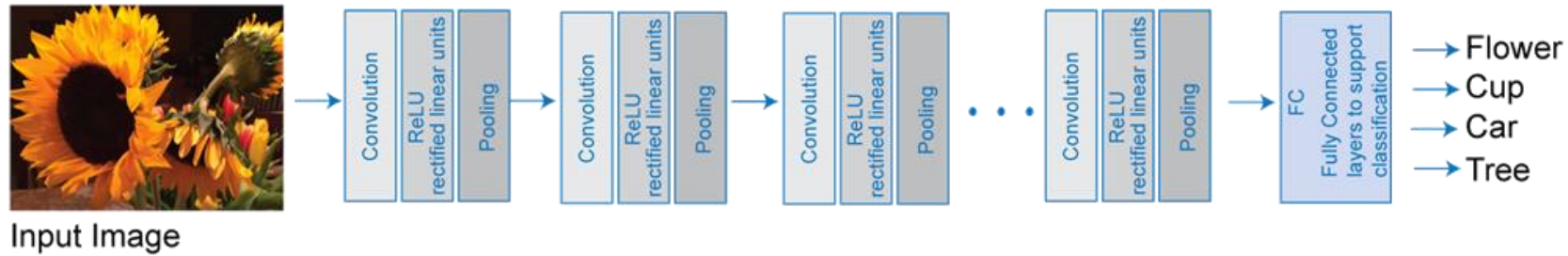
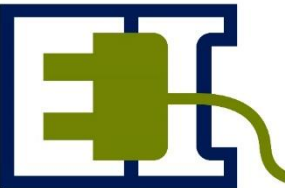


Image from the Mathworks





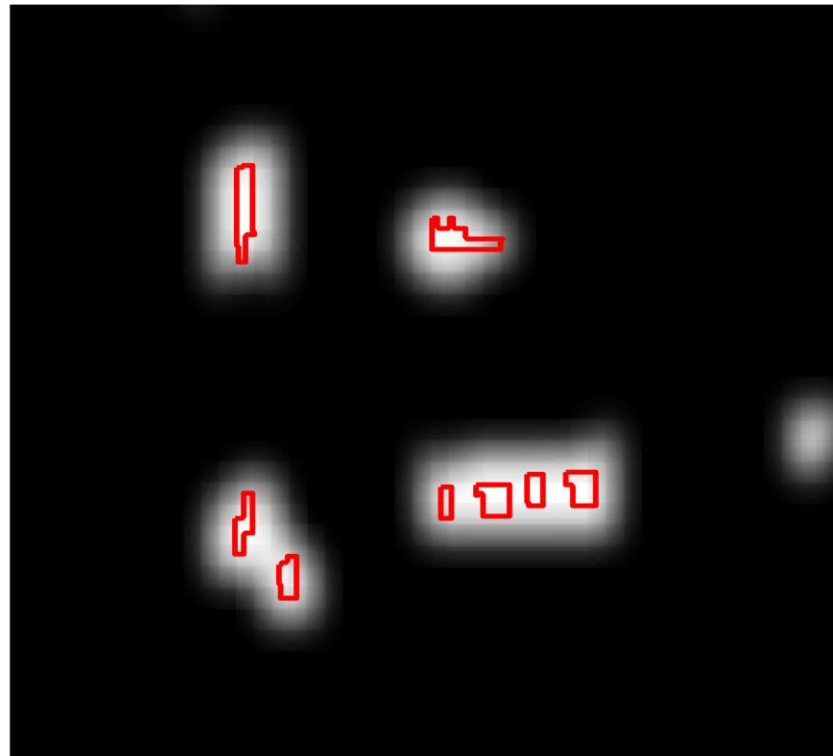
# Sample Results

Using convolutional neural networks (CNN)

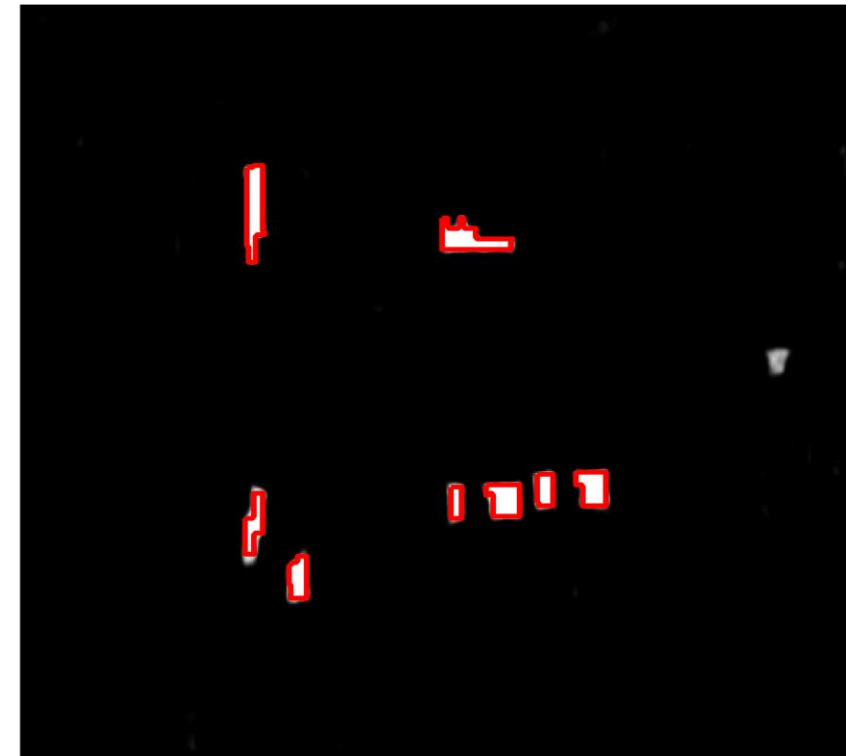
Original Image



VGG CNN



SegNet CNN

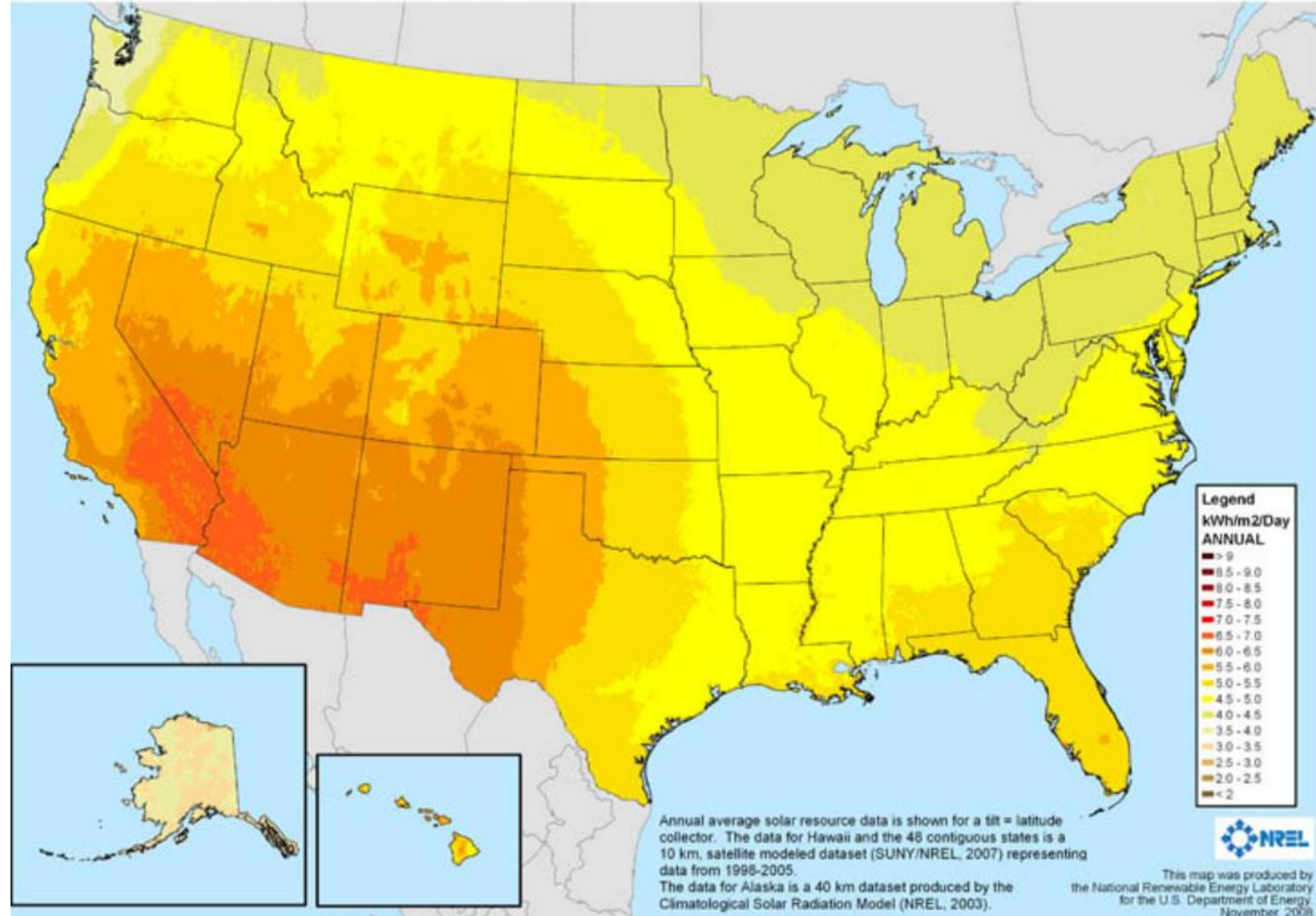
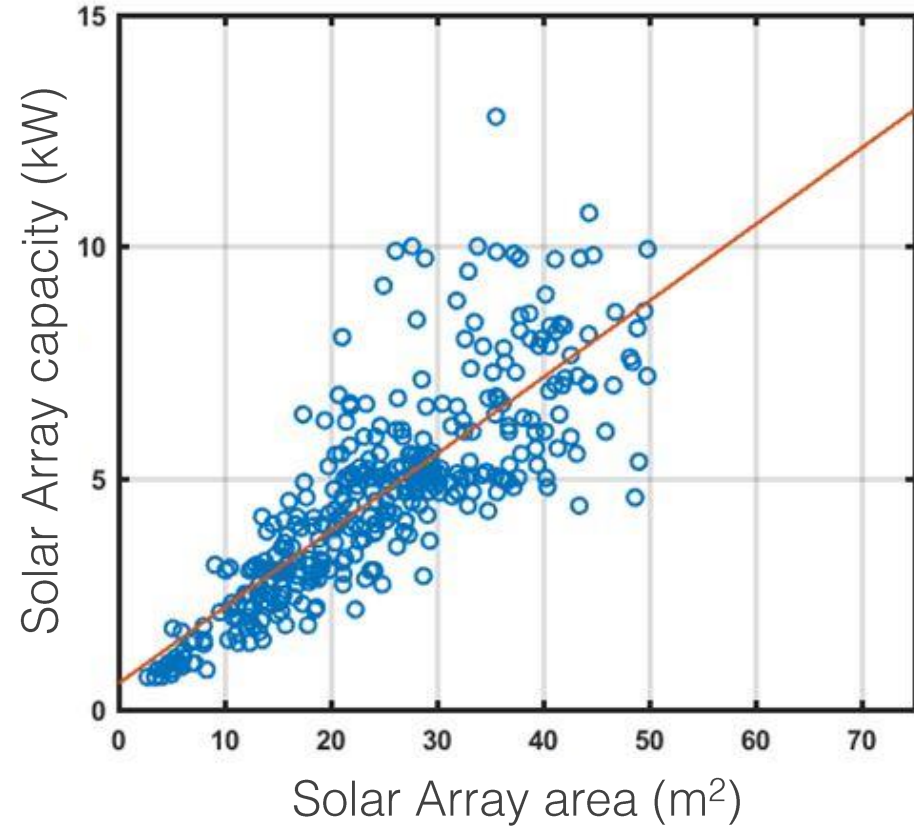


Camilo, J., L. Collins, K. Bradbury, J. Malof. "Application of a semantic segmentation convolutional neural network for accurate automatic detection and mapping of solar photovoltaic arrays in aerial imagery." Presented at the IEEE Applied Imagery Pattern Recognition Workshop in Washington, D.C., 2017.

# Capacity & Energy Estimation

Solar array area  $\rightarrow$  capacity (kW)

Solar array capacity  $\rightarrow$  energy generation (kWh)



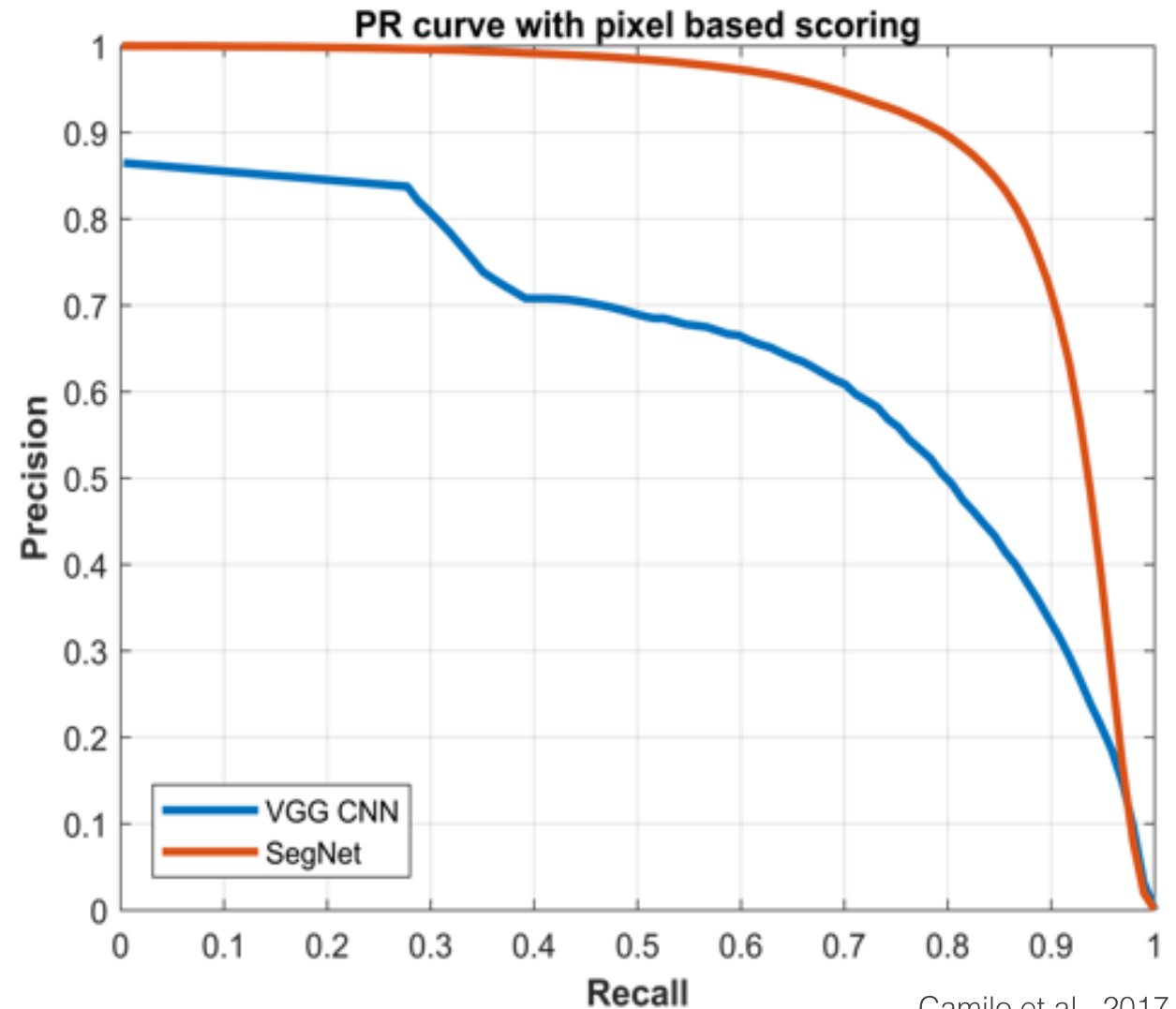
# Results

Pixel-wise results:

Object-wise results:

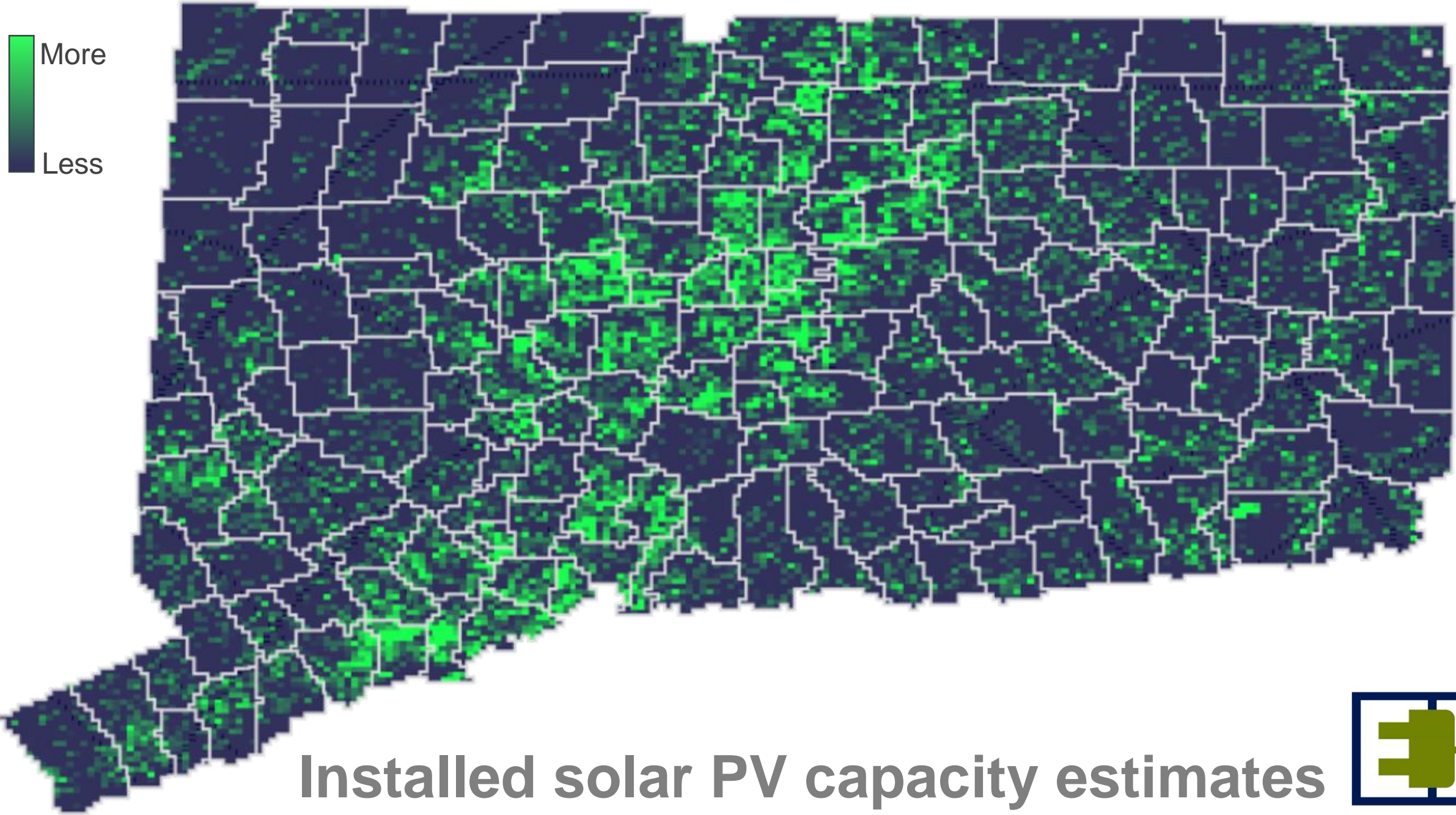
City	Precision	Recall	F <sub>1</sub> score
Fresno	0.77	0.78	0.77
Modesto	0.73	0.75	0.74
Stockton	0.73	0.70	0.71
<b>Overall</b>	<b>0.76</b>	<b>0.77</b>	<b>0.76</b>

Malof et al., 2019

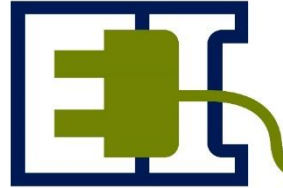


Camilo et al., 2017





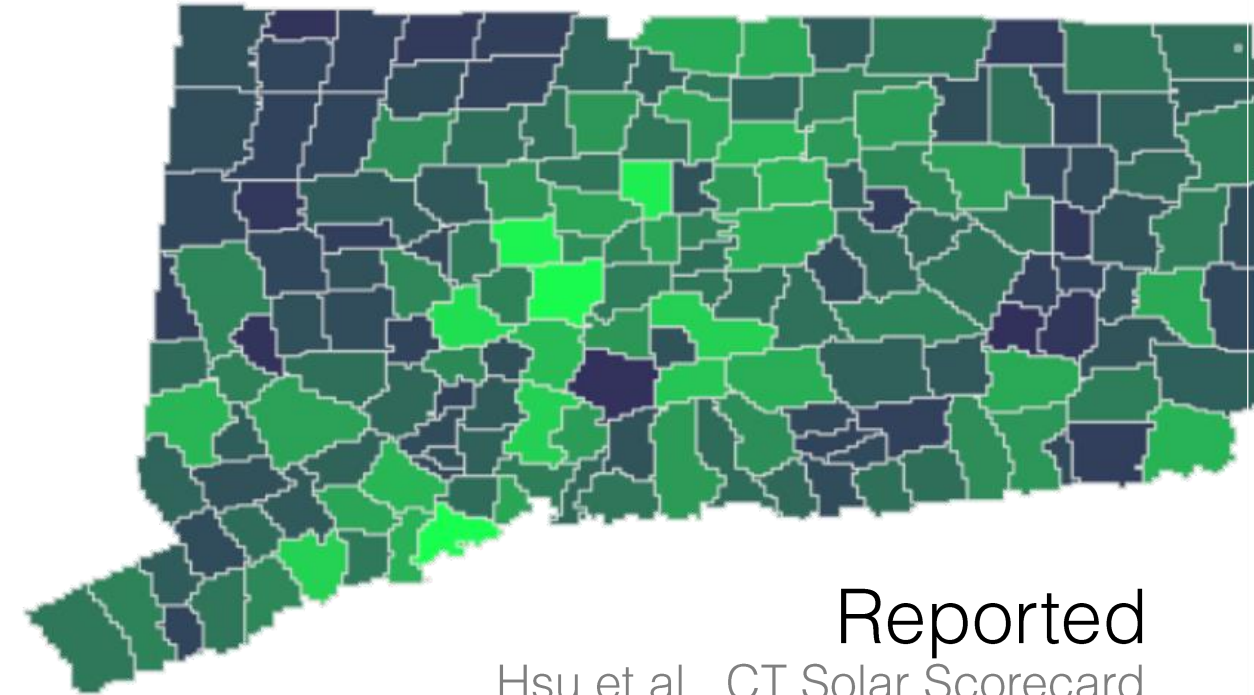
Installed solar PV capacity estimates



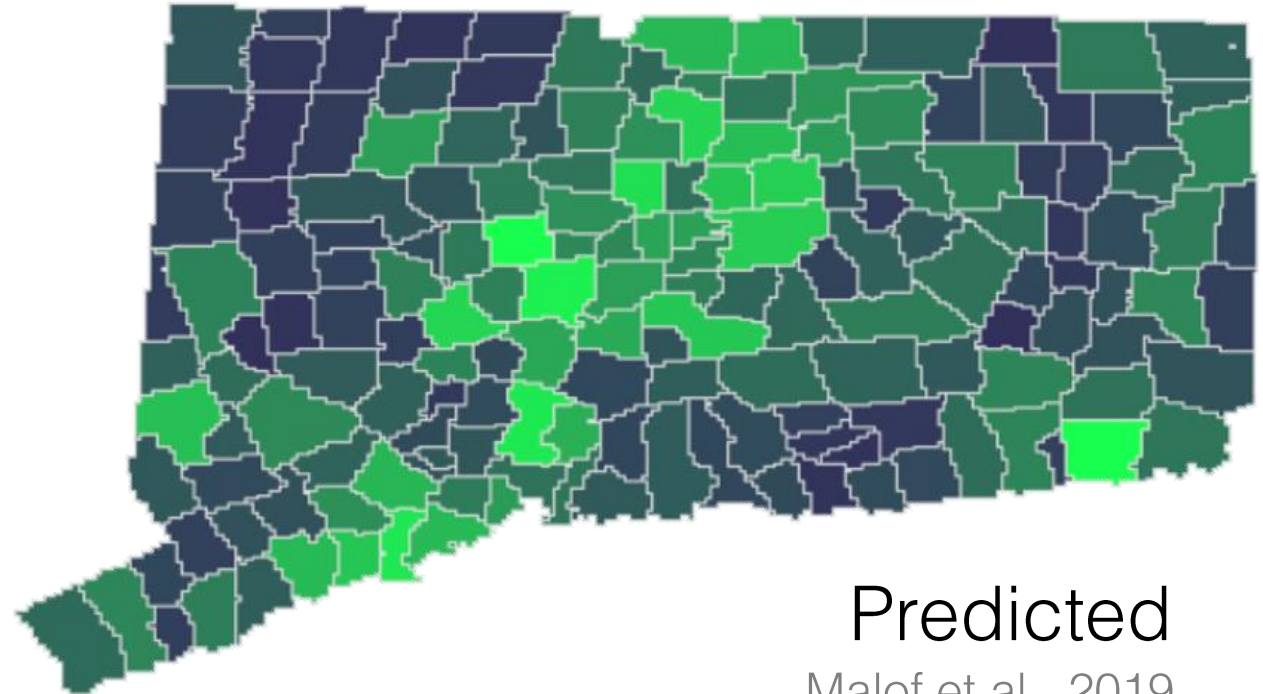
# Automated estimates of residential solar PV capacity for CT municipalities

Correlation coefficient of **0.89**

(Covers over 14,000 km<sup>2</sup>)



Reported  
Hsu et al., CT Solar Scorecard



Predicted  
Malof et al., 2019



## Substation



## Transmission



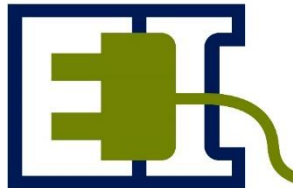
## Distribution

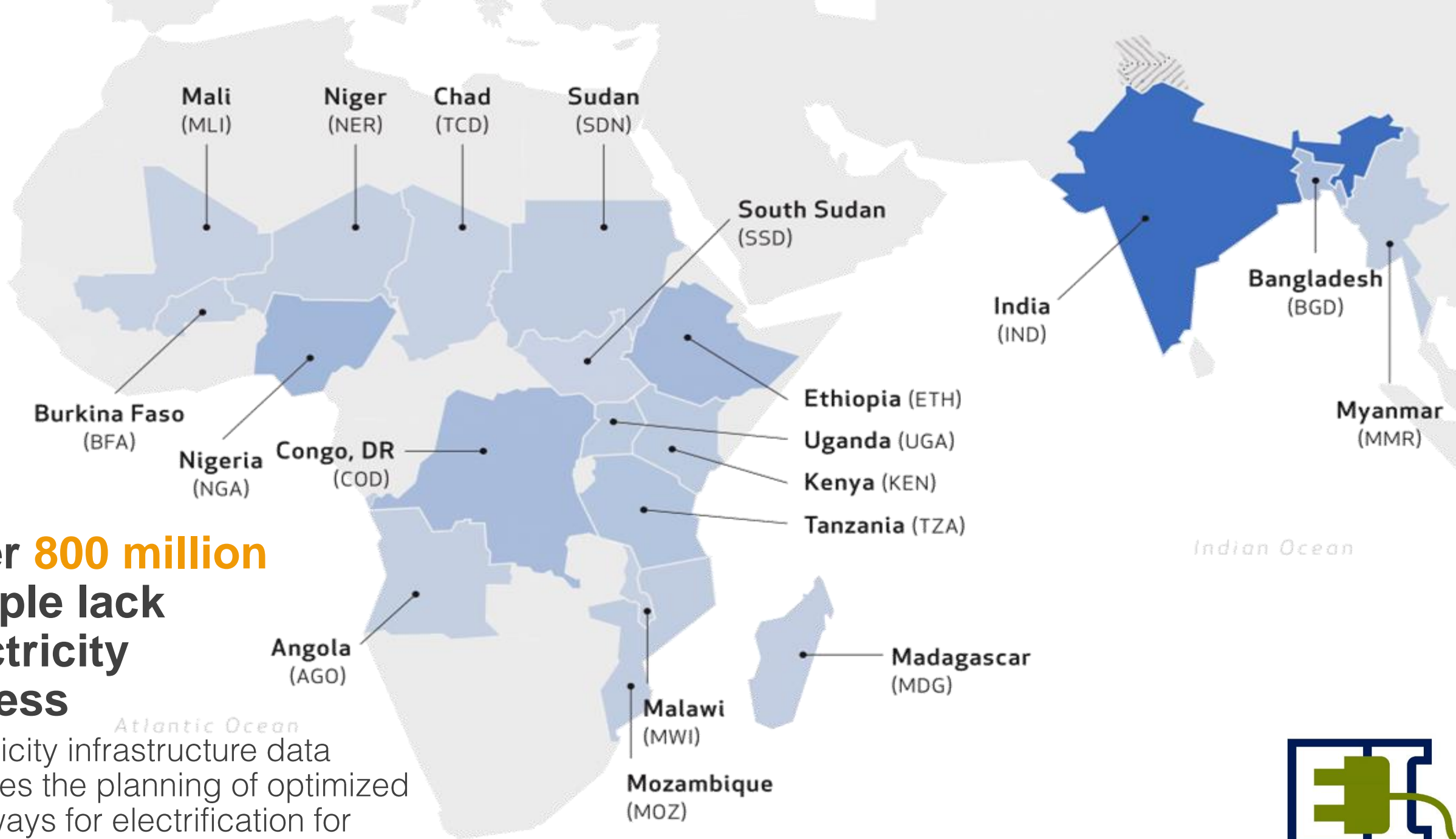


Sample images from Google Maps for demonstration purposes

# 2 Transmission & Distribution

Bohao Huang, Artem Streltsov, Kyle Bradbury, Jordan Malof, *Submitted 2019*

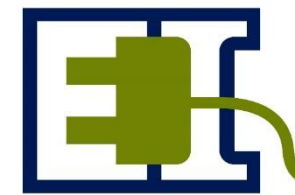


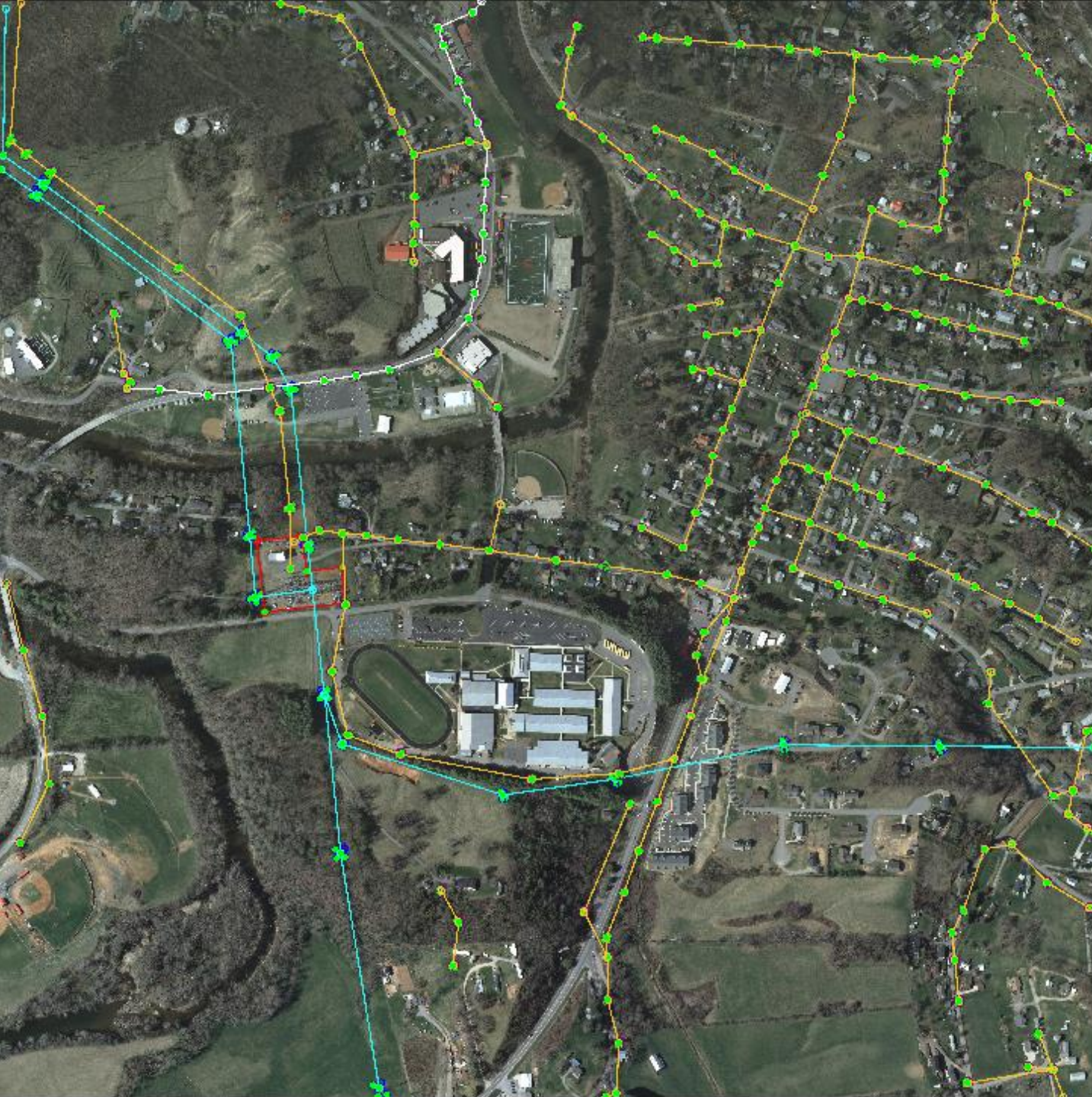


**Over 800 million people lack electricity access**

Electricity infrastructure data enables the planning of optimized pathways for electrification for rural communities

Image from Sustainable Energy for All, 2014





# Transmission Data

Annotated transmission, distribution, substation, and power plant imagery data

Data are available at:  
<http://bit.ly/2jy7WvL>





Ground truth

Visual

Topographic

Vis + Topo

# Results

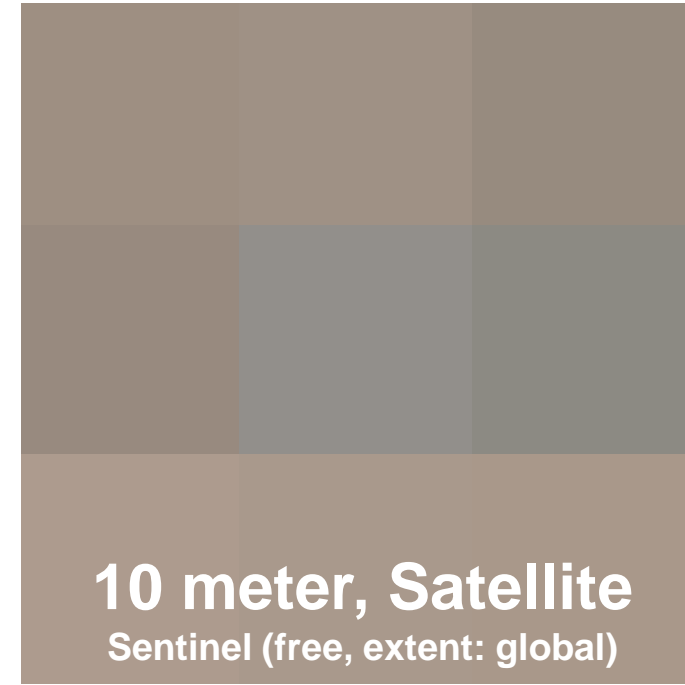
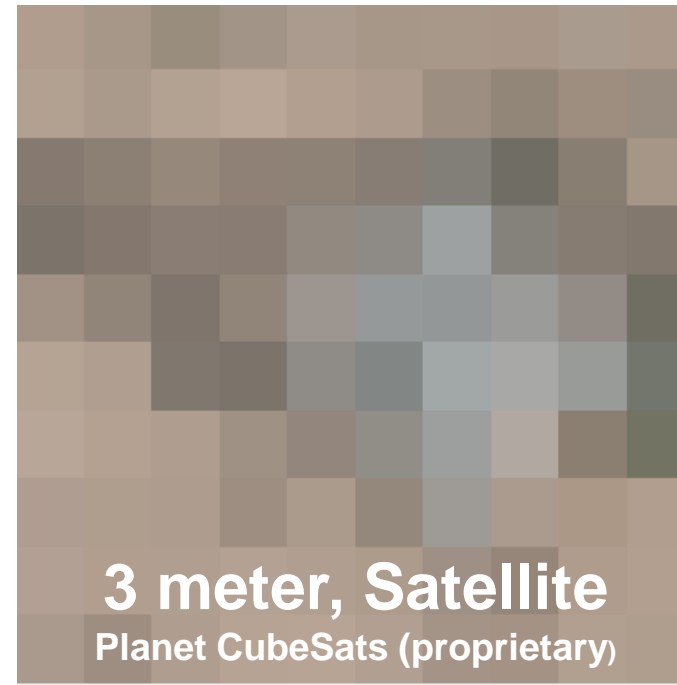


Huang, Bohao, Artem Streltsov, Kyle Bradbury, and Jordan Malof. "Automated Power Grid Mapping using Deep Learning and Overhead Imagery." *Submitted 2019*



# Challenge #1

Data  
resolution  
and  
availability





Tucson, AZ



Hartford, CT



Colwich, KS



Clyde, NC

**Challenge #2**  
**Geographic**  
**diversity**



# ENERGY

data analytics lab

Duke  ENERGY INITIATIVE

 SOCIAL SCIENCE RESEARCH INSTITUTE 

Duke  
PRATT SCHOOL of ENGINEERING



For questions, contact Kyle Bradbury ([kyle.Bradbury@duke.edu](mailto:kyle.Bradbury@duke.edu))

# References

- Bradbury, K., Saboo, R., Johnson, T., Malof, J., Zhang, W., Devarajan, A., Collins, L., Newell, R. (2016) "Distributed Solar Photovoltaic Array Location and Extent Dataset for Remote Sensing Object Identification." *Scientific Data*, 3. doi:10.1038/sdata.2016.106.
- Camilo, J., L. Collins, K. Bradbury, J. Malof. "Application of a semantic segmentation convolutional neural network for accurate automatic detection and mapping of solar photovoltaic arrays in aerial imagery." Presented at the IEEE Applied Imagery Pattern Recognition Workshop in Washington, D.C., 2017.
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- So, Brenda, Cory Nezin, Vishnu Kaimal, Sam Keene, Leslie Collins, Kyle Bradbury, and Jordan M. Malof. "Estimating the electricity generation capacity of solar photovoltaic arrays using only color aerial imagery." In *2017 IEEE International Geoscience and Remote Sensing Symposium (IGARSS)*, pp. 1603-1606. IEEE, 2017.
- Streltsov, Artem, Kyle Bradbury, and Jordan Malof. "Automated Building Energy Consumption Estimation from Aerial Imagery." In *IGARSS 2018-2018 IEEE International Geoscience and Remote Sensing Symposium*, pp. 1676-1679. IEEE, 2018.
- Malof, Jordan M., Boning Li, Bohao Huang, Kyle Bradbury, and Artem Streltsov. "Mapping solar array location, size, and capacity using deep learning and overhead imagery." arXiv preprint arXiv:1902.10895 (2019).
- Huang, Bohao, Artem Streltsov, Kyle Bradbury, and Jordan Malof. "Automated Power Grid Mapping using Deep Learning and Overhead Imagery." *Submitted 2019*