Domain Adaptable Deep Learning Models for Energy Infrastructure Detection using Synthetic Data Generation

Madeleine Jones¹, Caleb Kornfein¹, Alex Kumar¹, Aya Lahlou^{1,4}, Jaden Long¹, Madeline Rubin², Caroline Tang¹, Frankie Willard¹, Alena Zhang¹, Saksham Jain² Advisors: Kyle Bradbury^{2,3}, Jordan Malof^{2,3}, Simiao Ren²

¹Trinity College, Duke University, ²Pratt School of Engineering, Duke University, ³ Duke University Energy Initiative, ⁴ Duke Kunshan University





Energy & Environment

1 Energy Infrastructure Detection

Energy infrastructure mapping is vital for well-informed policy decisions in expanding energy access off of existing infrastructure. However, data on such infrastructure can be scarce. Our project aims to detect and map multiple energy infrastructures through publicly available satellite imagery globally



Transmission lines

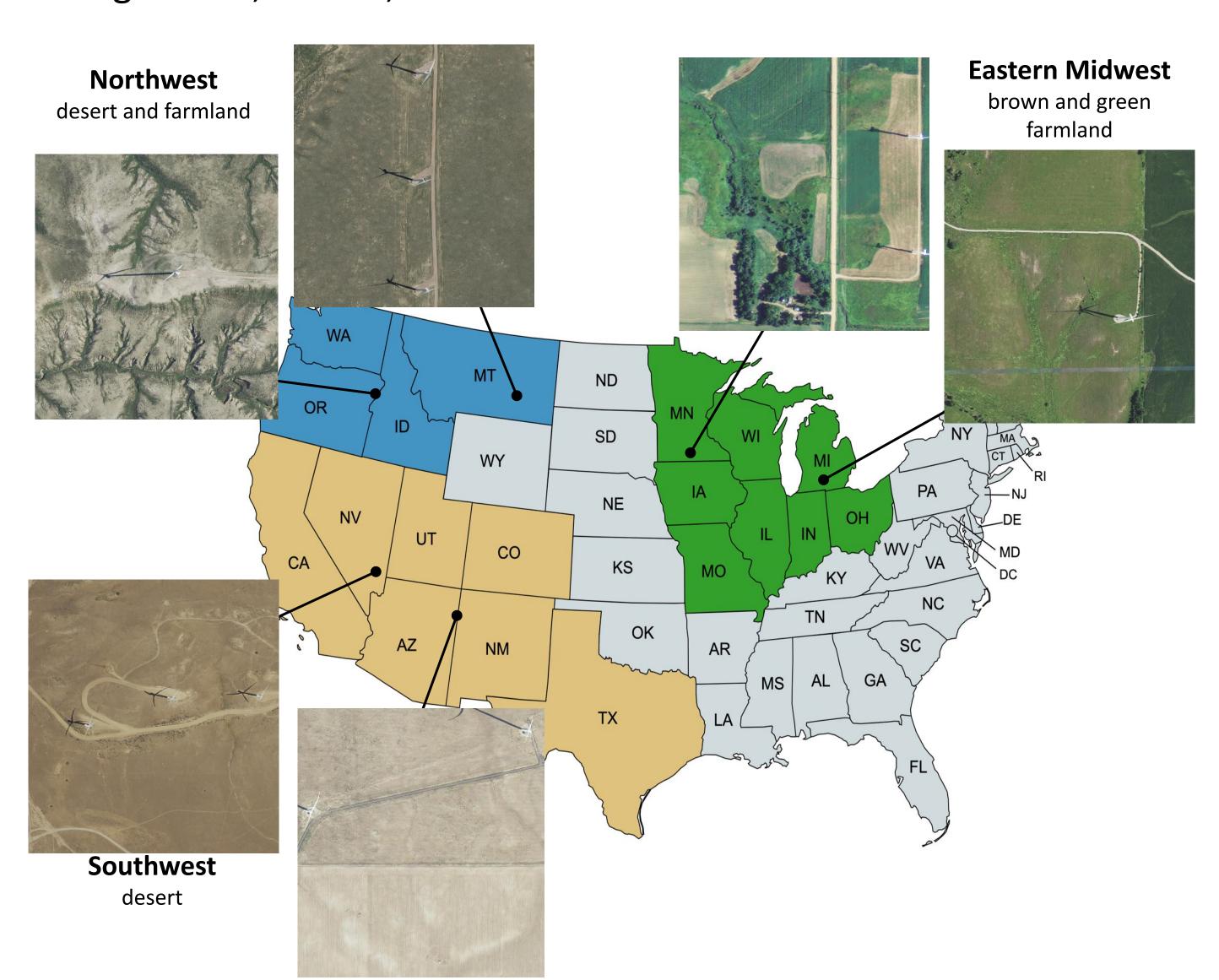


Solar Panels

Storage Tanks

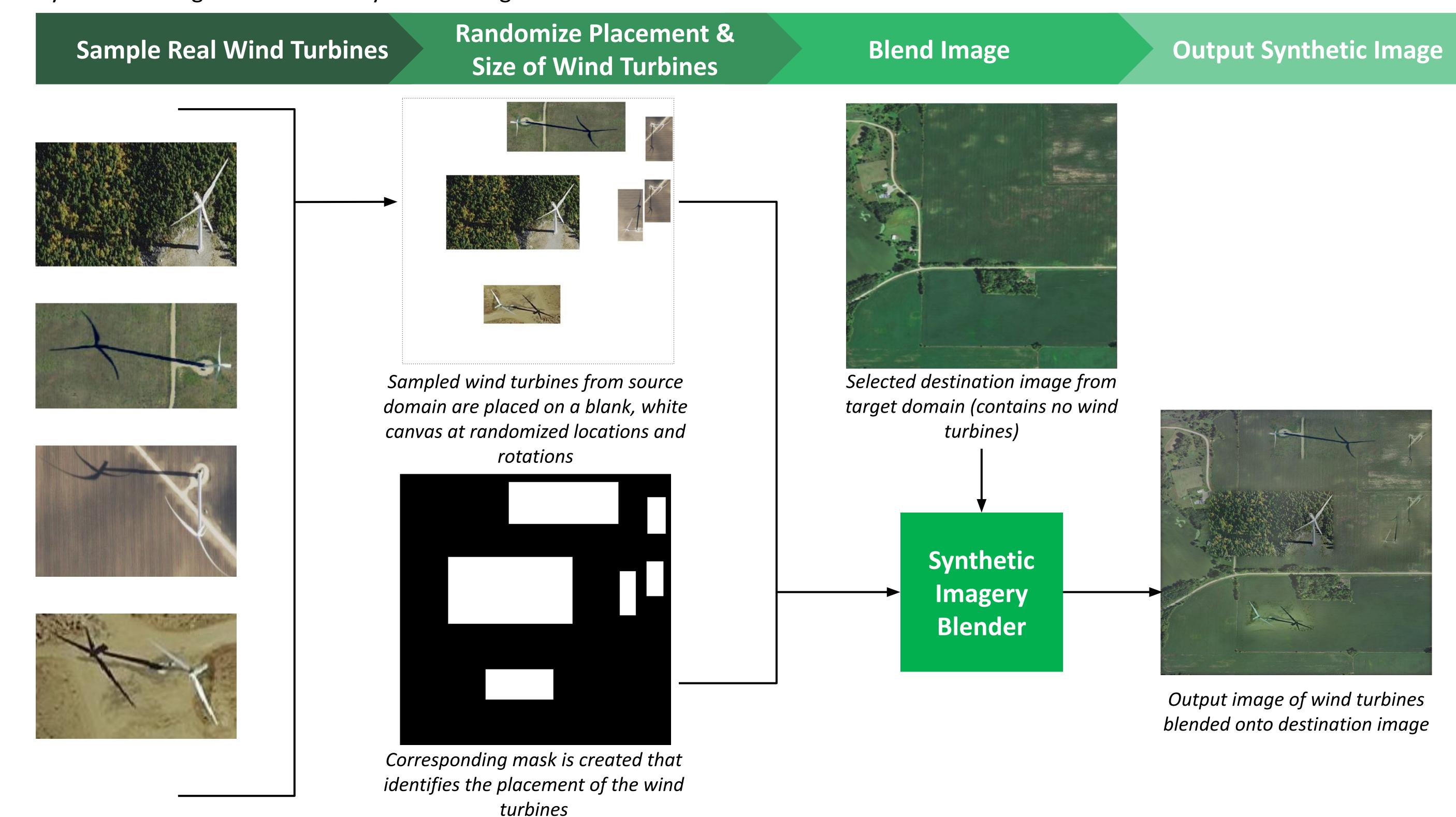
2 Domain Adaptation Problem

Computer vision techniques struggle when applied across satellite imagery from diverse geographies that vary in color, vegetation, terrain, etc..



B Proposed Solution: Synthetic Data Generation

Synthetic data generation with synthetic image blender GP-GAN.



Selected Results from Wind Turbines Experiment

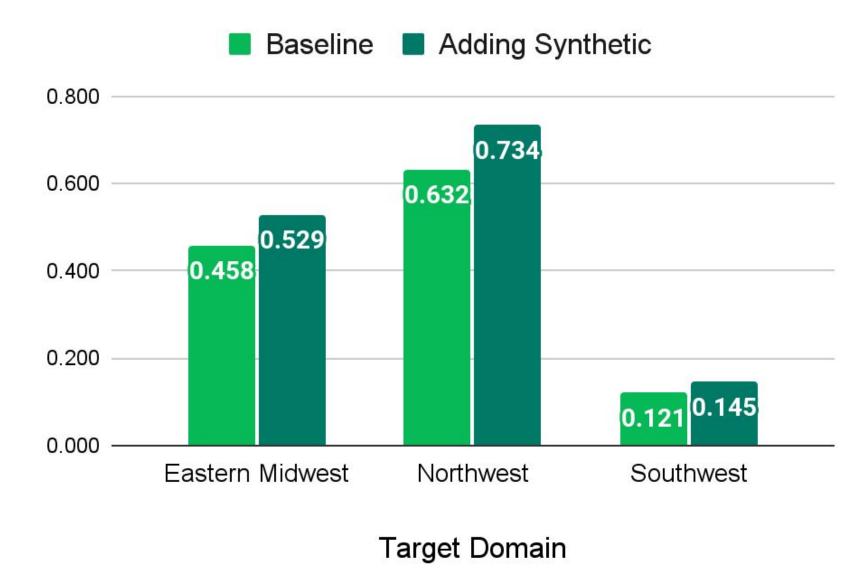
Synthetic data generation significantly improves energy infrastructure detection when it is cost-prohibitive to collect more real data.

Within-Domain Average Precision Baseline Adding Synthetic 1.000 0.750 0.822 0.887 0.895 0.936 0.566 0.599 0.250 Eastern Midwest Northwest Southwest

6.64% Performance Gain over Baseline

Target Domain

Cross-Domain Average Precision



24.7% Performance Gain over Baseline

For more information, visit our website



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