FPL Image Recognition for Pad Mounted Equipment

VDR4 Team 304



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Previous Work

Problem









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Objective

- Design a method for identifying pad-mounted transformers that have detected a faulty current using computer vision.
- Hardware and software solution





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Targets

Beacon (hardware)

Quick Installation

No Interference

Weather Resistant

Integrated with FCI

Inexpensive



Computer vision (software)

Video & Images

50 ft Detections

Varying Visibility/ Obstacles

80% Accuracy









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Solution - Hardware

- Beacon that moves up and down to indicate fault
 - LED as a backup when fault occurs
- Firmly attached to the top of transformer and tamper proof design
- Minimal power usage
- Other options
 - Lever on side, thermal strip, radio signal

Final Rating 1 0.2467 2 0.1017 5 0.4127 6 0.0949

Mount an external LED on the top of a stainless steel lever attached using rivets and powered by the transformer and infrared video input with YOLOv5 algorithm



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Solution - Software

- Image recognition model that can identify transformer, beacon, and state of beacon
- Model runs on drone footage from FPL Air





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Current Work

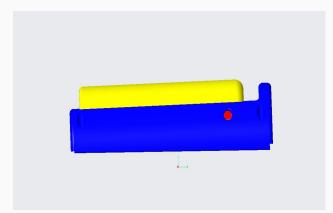
Current Work - Summary

- Designed a 3D model for printing
- Created a mock transformer for testing
- Developing of senior design website
- Tested drone camera capabilities
- Built a system for creating synthetic data



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3D Model



Device in its standard state

Device after it has received fault signal



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3D Print







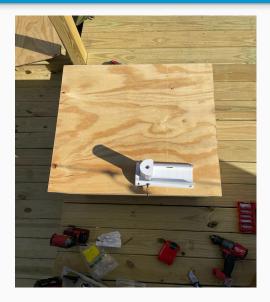


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Mock Transformer





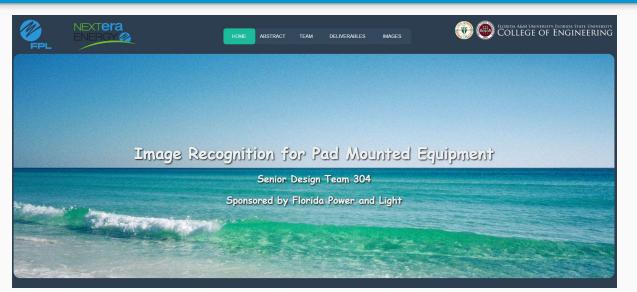




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Website

- All CSS/HTML
- Built from scratch
- Nav menu goes to Home, Abstract, Team, Deliverables





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Drone Testing and Use Case

- FPL Donated Drone
- Parrot Anafi Thermal
- Drone Licensing and Registration
- Future use for image collection
 - 4k and thermal imaging



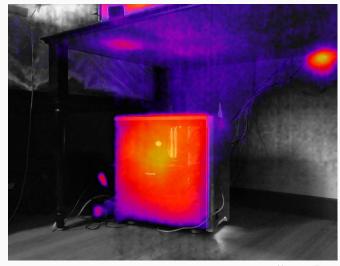


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Drone Pictures

- No flights, but we can still use the camera
- Thermal temperature adjustment
 - Adjust pictures to see certain temperatures
- Pictures or video
- Thermal or regular



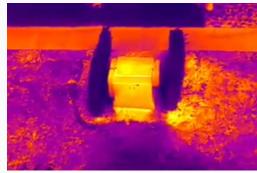


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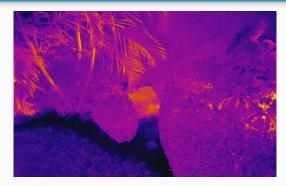
FPL Supplemental Images and Footage













Real and Synthetic Data

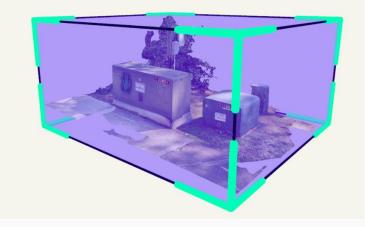
- Real:
 - Accurate to real life
 - Very inefficient to gather
 - Labeling done manually
 - Changes require gathering all new data
 - Limited to publicly accessible transformers

- Synthetic:
 - Similar to real life
 - Very efficient to create
 - Labeling done automatically
 - Changes require slightly modifying a script
 - More varied environments



Scan Transformers Using LIDAR

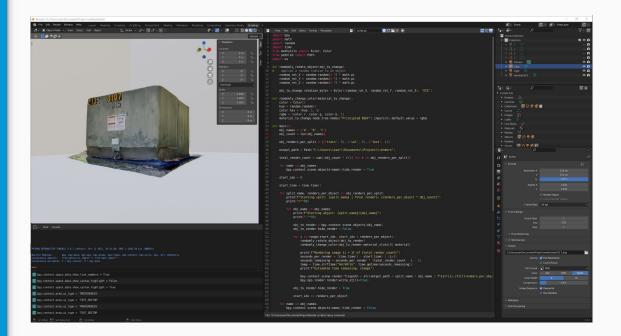
- Find publicly accessible FPL transformers
- Scan using Polycam app and LIDAR Scanner
- Clean 3D model for exporting
- Export 3D model for use in blender





Blender and Python

- What is Blender?
 - Blender is a free and open-source 3D rendering program.
- Python in Blender
 - Blender allows
 custom scripting
 using python and the
 blender api.



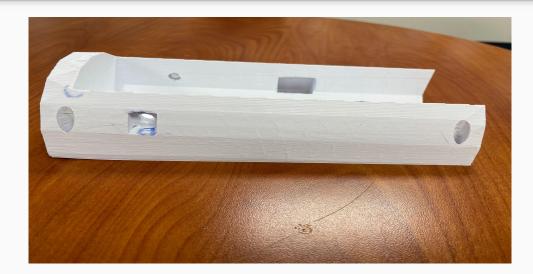




Future Work

Beacon Modifications

- Rivet holes
- Wire holes
- Servo housing
- LED housing
- Bolt Holes



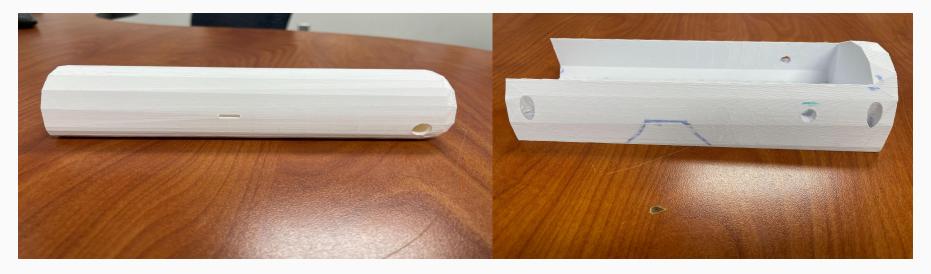


Beacon Modifications cont.





Beacon Modifications cont.





Order/Incoming Parts

- Wiring
- Low profile break-through multi-cord grip
- 10mm RGB LED
- 10mm plastic bevel LED holder
- Tapered proof screws
- Springs
- Bolts



Beacon Testing and Integration

- Merge incoming parts
- Verify lever rotation and LED lights
- Attach beacon to transformer



Synthetic Data - Improvements

- Gather more assets and 3D scans to make the data set more diverse.
- Look into creating synthetic thermal images for training.
- Implement more variability in rendered environments.
- Test synthetic data to make sure accuracy is improving.



YOLO Training and Testing

- Real image collection
- 200 Real Images/Video from FPL for testing
- Vary training dataset size
- Combinations of synthetic to real data:
 - Ex. all real, all synthetic
 - Ex. 10:1, 100:1, 1000:1 of synthetic:real
- Further optimization until 80% accuracy is reached



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Future Work - Summary

- Beacon modifications
- Data creation & collection
- Train & optimize YOLO model
- Finalize website
- Final demonstration & engineering day



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End

Any Questions?