

Generating Useful Data with Computer Vision Tools: 2 Use Cases (PDFs & Imagery)



Location matters

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Overview



Utah Department of
Health & Human
Services

- **Two Computer Vision (CV) Projects**
 - **DHHS Cooling Towers Project**
 - **UDOT Parcel Detection Project**
- **Motivation**
- **Process**
- **Results**



Motivation - DHHS Cooling Towers

- Legionella bacteria can cause a serious type of pneumonia called Legionnaires' disease
- Legionella can grow/spread in large building water systems
 - Water tanks, HVAC, large/complex plumbing systems, cooling towers
- Cooling towers are concerning because they can release aerosolized water into the atmosphere
 - If Legionella is present, the aerosolized water can spread the bacteria over miles*

*[CDC - Controlling Legionella in Cooling Towers](#)

Commons Sources of Infection

Outbreaks of Legionnaires' disease are often associated with large or complex water systems, like those found in hospitals, hotels, and cruise ships.

The most likely sources of infection include:



Water used for showering
(potable water)



Cooling towers (parts of large
air conditioning systems)



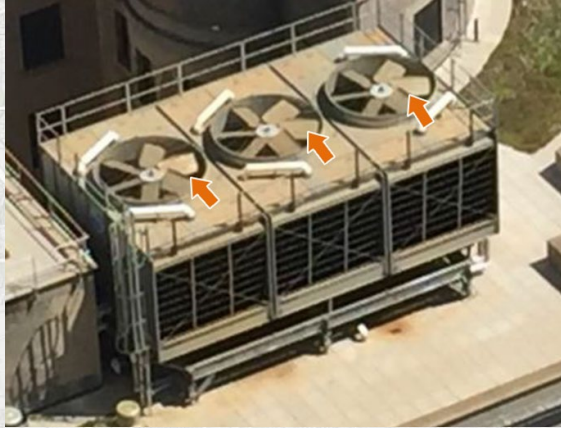
Decorative fountains



Hot tubs

[CDC - Legionnaires' Fact Sheet](#)

Motivation - DHHS Cooling Towers



- Cooling towers can cause outbreaks of Legionnaires' disease when they are not adequately maintained
- They are often investigated & located using aerial imagery during Legionnaires' outbreaks
- Cooling towers have distinctive features that make them identifiable
- Researchers and the CDC have used object-detection models to identify potential cooling towers in aerial imagery ([TowerScout](#))



[CDC - Photos of Cooling Towers](#)



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[CDC - Procedures for Identifying Cooling Towers](#)

Motivation - DHHS Cooling Towers

- Cooling towers can cause outbreaks of Legionnaires' disease when they are not adequately maintained

This problem is solvable with computer vision!

...located using Legionnaires' outbreaks distinctive features that

...and the CDC have used object-detection models to identify potential cooling towers in aerial imagery ([TowerScout](#))



[CDC - Photos of Cooling Towers](#)



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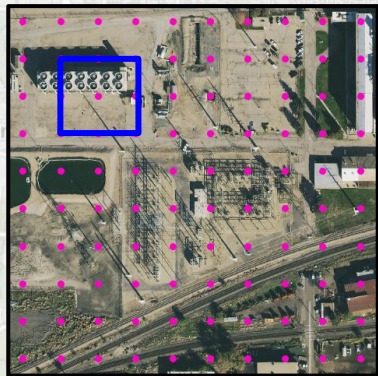
[CDC - Procedures for Identifying Cooling Towers](#)

DHHS Cooling Towers - Process Overview

- 1) Build Index & Footprint
- 2) Download Images
- 3) Detect & Locate Towers
- 4) Post-process results
- 5) Build Web Map



1



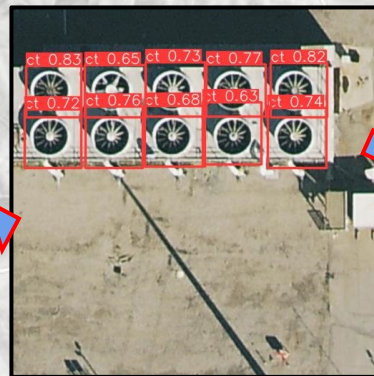
Build Index

2



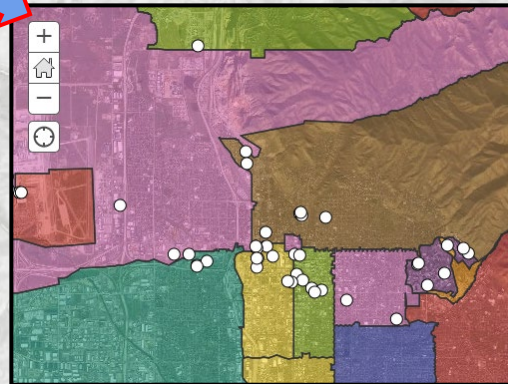
Download
Images

3



Detect &
Locate
Towers

4/5



Build Web
Map

Step 1 - Build Imagery Index & Footprint

- **Discover** Web Map Tile Services (WMTS) used for statewide imagery
- Must build imagery index at highest zoom level (20)
- Create point for upper left corner of each WMTS tile (row, col)
 - Calculate latitude/longitude based on row/col values
 - 275,000,000 tiles cover the state of Utah!
- Create processing footprint to select a subset of all image tiles
 - Buffer census places by 800m
 - Buffer large buildings (>5k sq ft) by 800m
- Select WMTS indices of tiles in processing footprint
 - Load tile index and footprint into Google BigQuery
 - Run Spatial SQL query to select tiles to process (~6%)
 - (data is waaaayyy too big for desktop GIS)



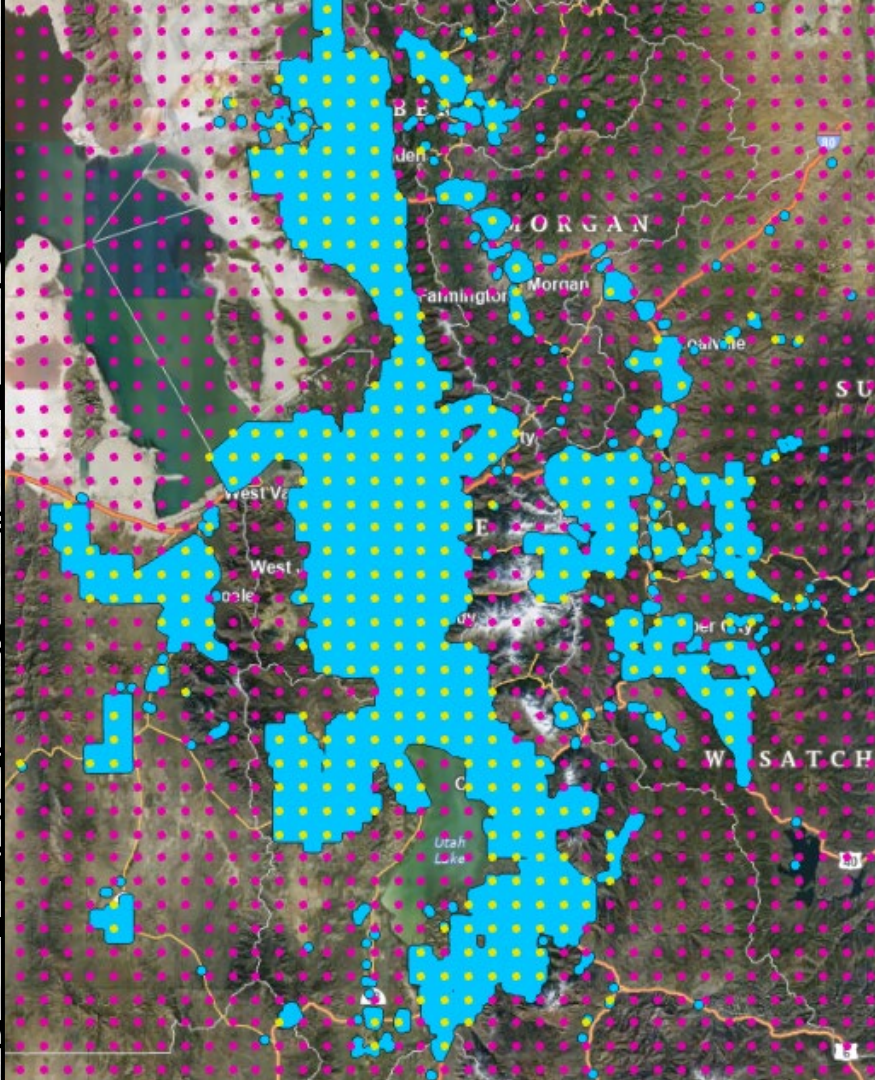
Build Index



Create Footprint

Step 1 - Build

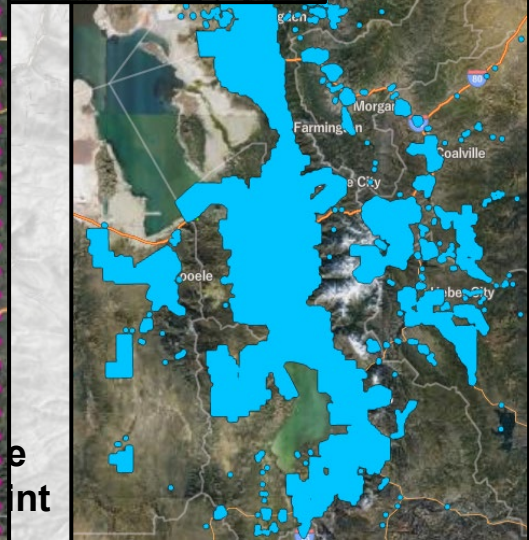
- Discover Web Map
- statewide imagery
- Must build image
- Create point for u
(row, col)
 - Calculate latitude
 - 275,000,000 tiles
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image tiles
 - Buffer census pla
 - Buffer large build
- Select WMTS indi
 - Load tile index a
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 - (data is waaaayy



Footprint



Build
Index



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int

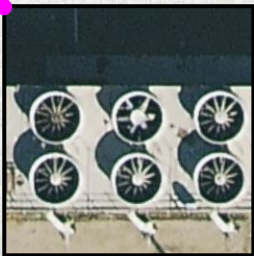
Step 2 - Download Images

- Iterate through tile indices within footprint
 - Download primary tile and 3 neighboring tiles with HTTPS GET requests:
 - https://discover.agrc.utah.gov/login/path/{quad-word}/tiles/15cm_hexagon_utah/20/{col}/{row}
- Build mosaic image

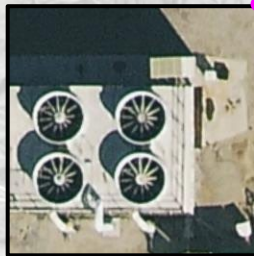


Each WMTS tile is 256x256 pixels

{col}, {row}



198263, 394029



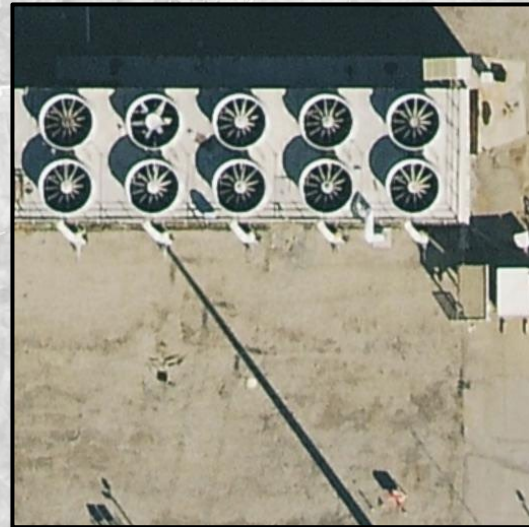
198264, 394029



198263, 394030



198264, 394030



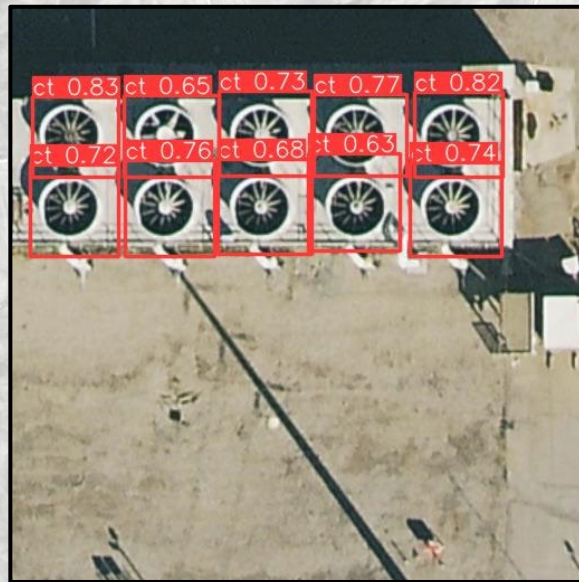
Model input performs best on 512x512 images

Step 3 - Detect & Locate Towers

- Run PyTorch model on each mosaic
 - Pre-trained "[TowerScout](#)" model, provided by CDC
 - YOLOv5 backbone
- Get results as a dataframe



Detect
Towers



*TowerScout created by Karen Wong, Jia Lu, Gunnar Mein, and Thaddeus Segura, licensed under [CC-BY-NC-SA-4.0](#)

Step 3 - Detect & Locate Towers

- Calculate X/Y coordinates of tower centroid
 - PyTorch model provides tower location bounding box in pixels from upper-left corner
 - Centroid is calculated from xmin, xmax, ymin, ymax values
 - Able to convert pixels into geographic space because we know:
 - 1) coordinates of the tile's upper-left corner
 - 2) pixel size at WMTS zoom level 20 (0.1492910708688 meters)

y

x



	xmin	ymin	xmax	ymax	confidence	x_centroid_px	y_centroid_px	x_centroid_3857	y_centroid_3857
9	265.932	130.687	347.821	217.394	0.633	306.877	174.041	-12460145.259	4978279.451
8	100.593	78.307	185.93	151.344	0.654	143.262	114.826	-12460169.686	4978288.291
7	184.386	135.914	268.444	220.131	0.676	226.415	178.023	-12460157.272	4978278.857

```
#: x, y = upper-left corner coordinates of tile in web mercator
```

```
#: calculate centroid x/y coords in web mercator
```

```
meters_per_pixel = 0.1492910708688
```

```
results_df["x_centroid_3857"] = x + results_df["x_centroid_px"] * meters_per_pixel
```

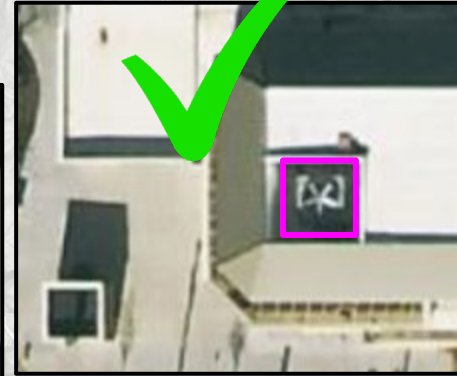
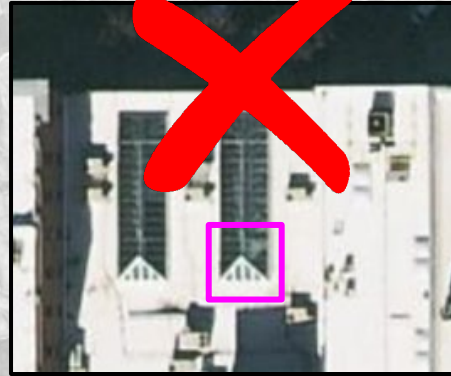
```
results_df["y_centroid_3857"] = y - results_df["y_centroid_px"] * meters_per_pixel
```

Calculate
X/Y

- Upload results to Cloud SQL Table

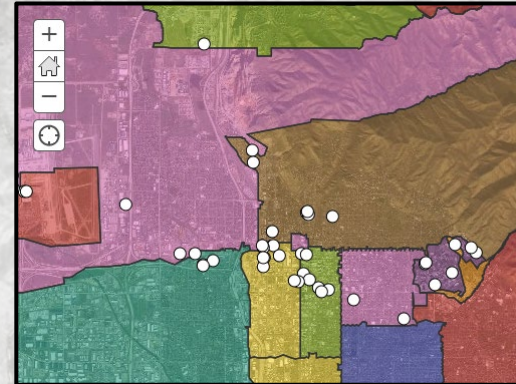
Step 4 - Post-process Results

- Manually validate detected cooling towers
- Enrich data with attributes from other statewide datasets
 - Nearby address
 - County
 - City
 - Zip Code
 - Small Health Statistical Area



Step 5 - Build Web Map

- Create hosted feature layer of cooling tower locations
- Build a web map for DHHS users
- Add other relevant, health-related layers and tools



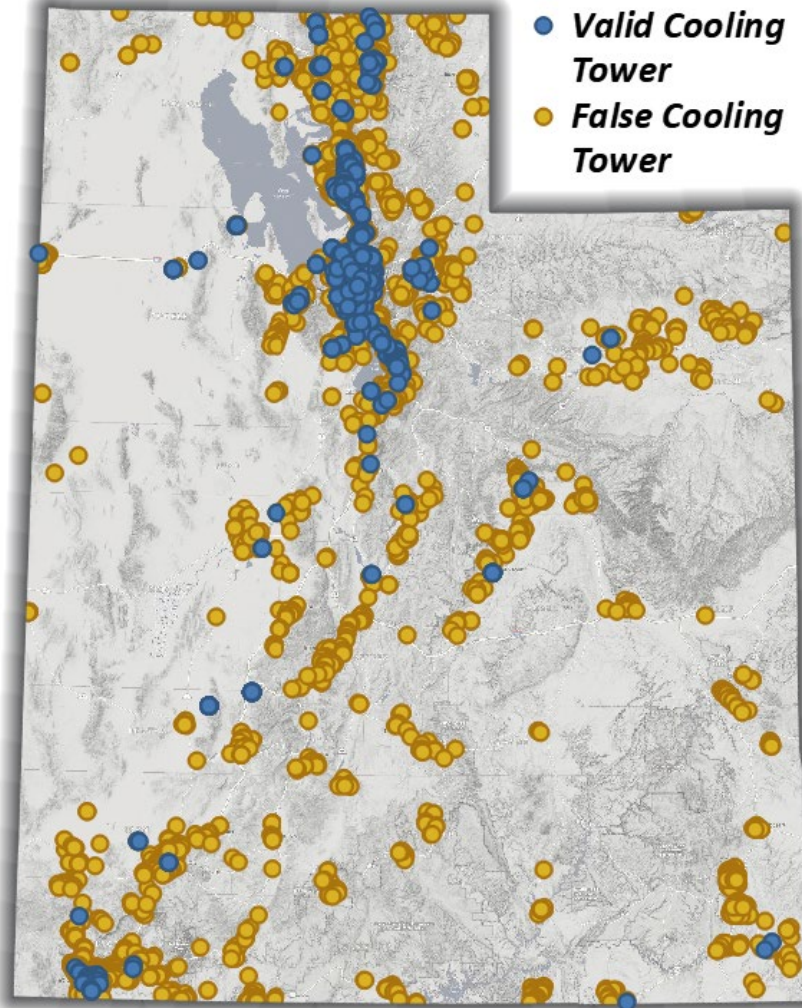
DHHS Project Results

- **Data**

- 6,324 potential cooling towers detected with confidence > 0.5
- 1,561 validated cooling towers (25%)

- **Processing**

- 225 days (5,400 hours) worth of wall-clock computing time completed
- Scaling up with cloud computing enabled completion in 9 calendar days!



Final words

- A lot of useful data and information can be locked away in imagery (or documents)
- Computer Vision tools can help unlock that data
- Cloud computing can reduce processing time by orders of magnitude
- DHHS cooling towers and UDOT parcel detection projects highlight these possibilities

