

Using Satellites and Advanced Analytics to Track Fires

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Who are we?

- 99 people, \$38M in VC investment
- Decades of experience in large-scale computing, remote sensing, and machine learning
- Goal: Build a digital twin of the Earth



Early wildfire detection and tracking

Smoke from the Ute Park fire

Detecting wildfires: The tried and true methods

• Traditionally, wildfires have been detected by people...



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 - Where they live



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 - Where they are travelling



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 - Where they live
 - Where they are travelling
 - By pilots flying overhead
 - And by dedicated fire suppression infrastructure and operations
 - Fire lookout towers
 - Reconnaissance flights



- Traditionally, wildfires have been detected by people...
 - Where they live
 - Where they are travelling
 - By pilots flying overhead
 - And by dedicated fire suppression infrastructure and operations
 - Fire lookout towers
 - Reconnaissance flights
- However, big gaps still exist...
 - Spatially: Remote areas still exist
 - People cannot be everywhere at once
 - Temporally: Day versus night
 - Visibility decreases during the night
 - People go to sleep



What can we do to help?

- We are building an "early-warning" wildfire detector using remote sensing imagery and machine learning
- Augment, not replace existing detection methods
 - There is no one detector to rule them all
- Concrete goals:
 - Detect at least three large wildfires per year, before they are reported
 - Produce as few false positives as possible

GOES-16 (GOES East)

GOES-17 (GOES West)



GOES-16 and -17: Keeping an eye on the Western Hemisphere

- Image are collected:
 - Every 5 minutes over CONUS/PACUS
 - Every 15 minutes over the "full disk"
- It takes us around 4 minutes to ingest these images into our platform after they have been captured
- Images are collected in the thermal infrared spectrum
 - Useful for detecting and tracking hot things (wildfires)
 - Pixel size: 2 km × 2 km
- Enables us to build a detector that is real time and always on
 - Comes at a cost: We are best at detecting fast burning wildfires



The first day of the Camp Fire, as seen by GOES-16

- The animation on the right shows:
 - Temperature (in red), as measured by the GOES-16 3.9 μm thermal infrared band
 - 911 calls related to the disaster, as released by the Butte County Sheriff's Office
- Fast burning wildfires begin as a subpixel area in the image
 - Results in one pixel that is slightly hotter than its surroundings
 - How can we leverage this to build a wildfire detector?



Wildfires as anomalies

- For a given location, wildfires are rare and cause this area to be much hotter than normal
 - In this way, wildfires are anomalous
- Hotter than normal:
 - In an absolute sense
 - In a physical sense
 - In relation to the past
- Strategy: Model what normal looks like, and compare this model with the current situation to detect wildfires





ΔT_{3.9 - 10.3 μm}

T_{3.9 μm}













Conclusions, and where we're headed

- We've built a wildfire detector, and early results seem promising
- Next step is to operationalize our detector, begin making detections in real time
 - Start at the state level
 - Scale up to CONUS/PACUS, eventually the Western Hemisphere
- Simultaneously, we want to start trialling our detector with wildfire response organisations
 - \circ ~ Aim is to build an operationally useful detector, feedback is critical

Questions?

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