

CSC 411

Wildfire Prediction

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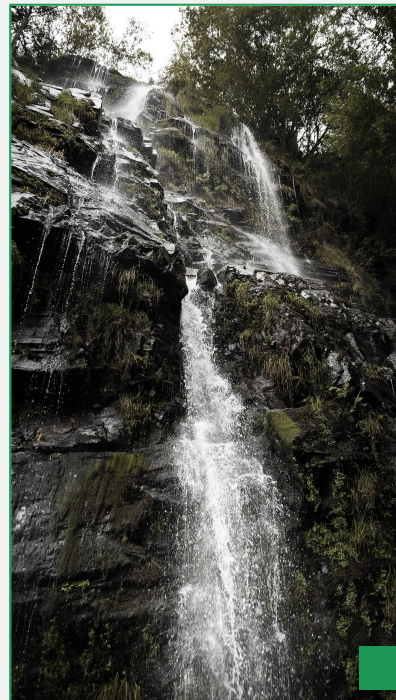
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01.

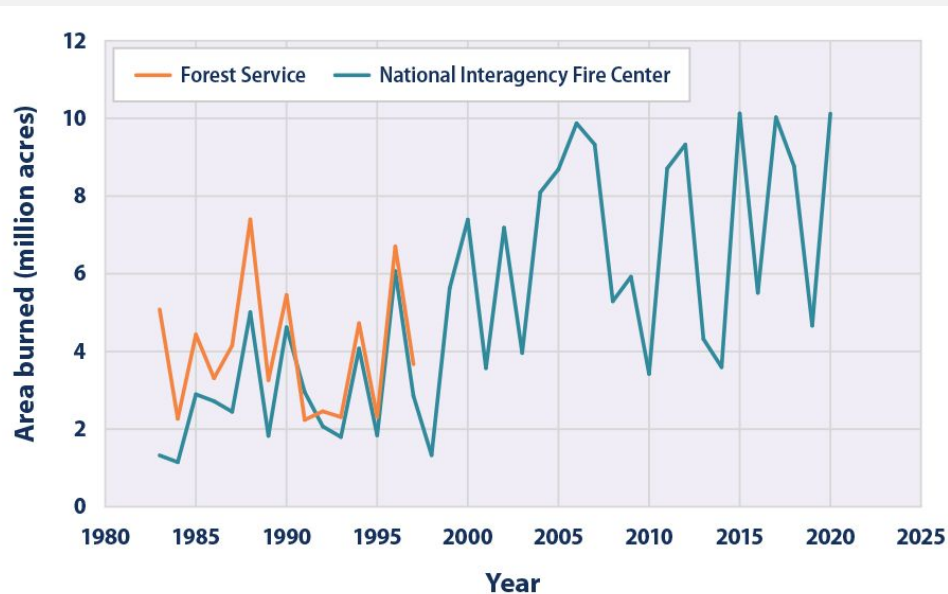
Introduction





Why Wildfires

- Negative effect
 - Public health
 - Natural environment
 - Federal and State budgets





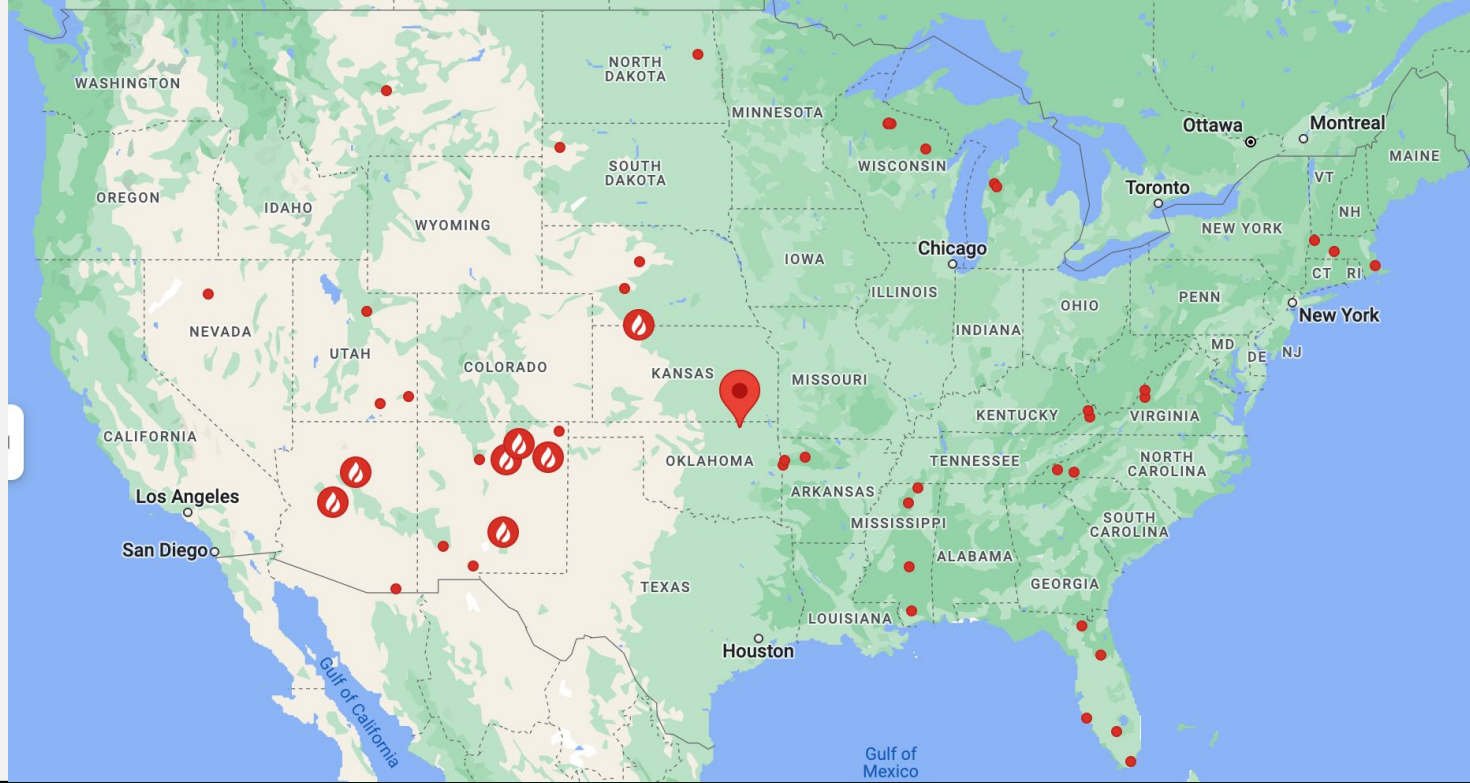
17,300

Fires since Jan 13th 2022

831,842

Acres burned since Jan 13th 2022





Current Wildfires on April 28nd, 2022



02.

Data





Data Source

Application for Extracting and Exploring Analysis Ready Samples
(AppEEARS)



National Aeronautics and
Space Administration



Land Processes
Distributed Active Archive Center



National Snow and Ice Data Center
Distributed Active Archive Center



Socioeconomic Data and
Applications Center



Oak Ridge National Laboratory
Distributed Active Archive Center





Requesting Data



1. Input coordinates
2. Select the satellite
3. Select the variables you need
4. Input the date range
5. Submit

Uploaded coordinates (ID, Category, Lat, Long): 3

36.7136, -96.2458

36.7808, -95.8378

37, -97

Selected layers

| | | |
|--|--------------|---|
|  dayl | 1000m, Daily | — |
|  prcp | 1000m, Daily | — |
|  srad | 1000m, Daily | — |
|  swe | 1000m, Daily | — |
|  vp | 1000m, Daily | — |

Enter a name to identify your sample

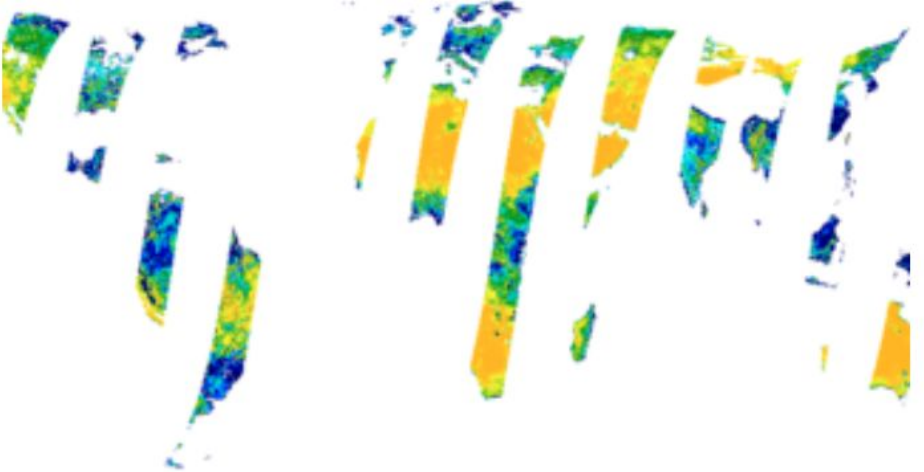
Sample_name



Data Capturing

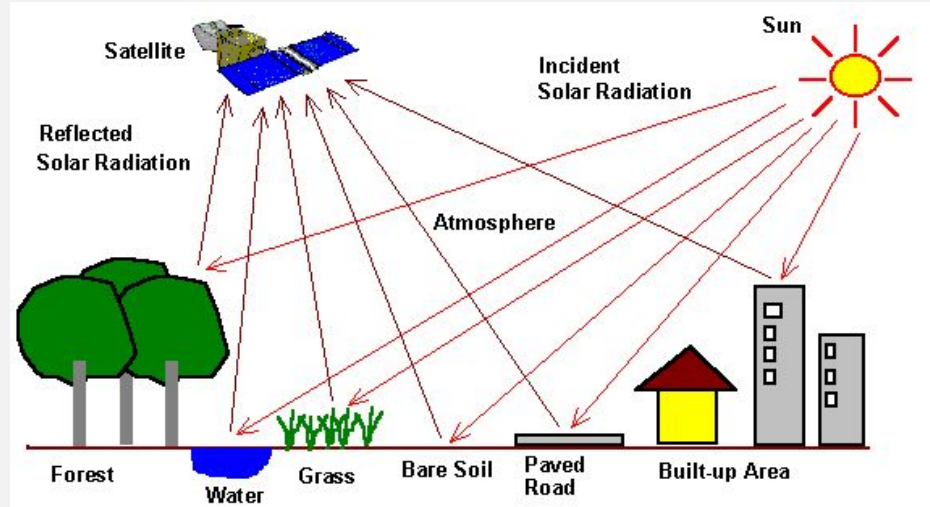
- Two different Satellites
 - MODIS
 - Daymet

- Wildfire Points
 - Federal
 - State
 - Local



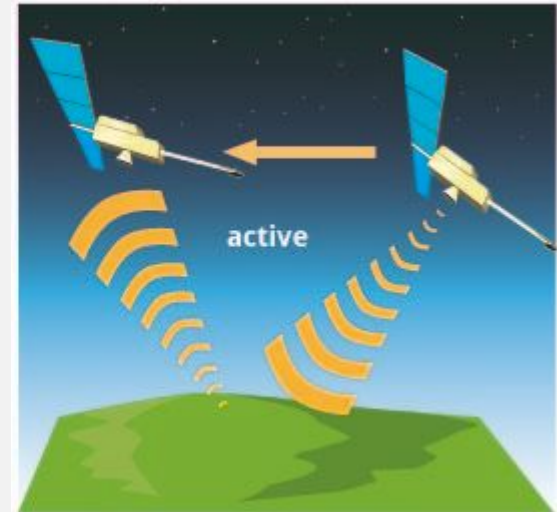
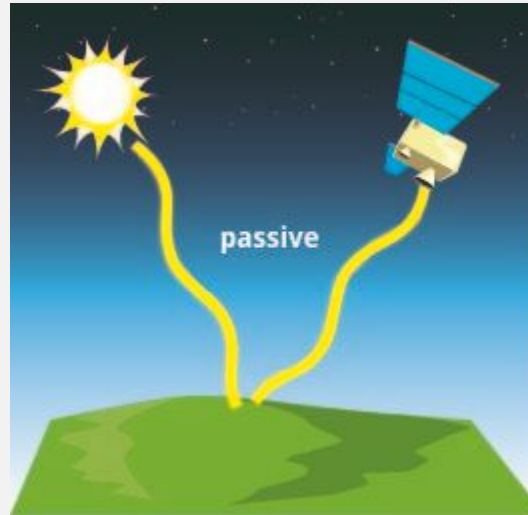
Remote Sensing Advantages

- Remote Areas
- Fast upload speeds
- Can cover a lot of land



Remote Sensing Disadvantages

- Spatial resolution
- Temporal resolution
- Spectral resolution





03.

Variables

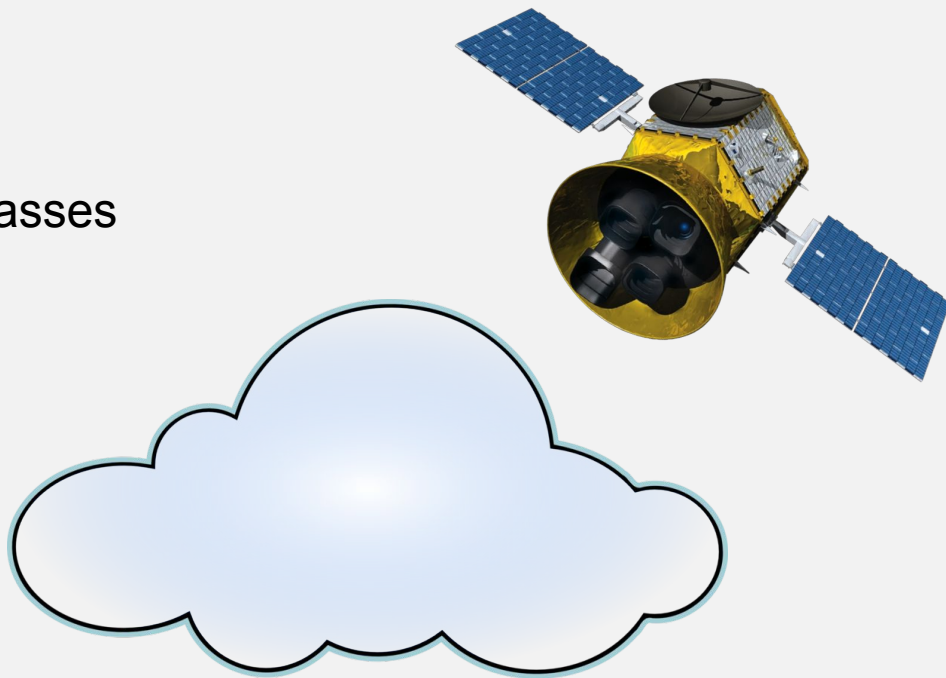




Feature Selection



- Longitude and Latitude
- Quality of measurement
- High, medium, and low classes
- Visible clouds
- SMAP





Daylight

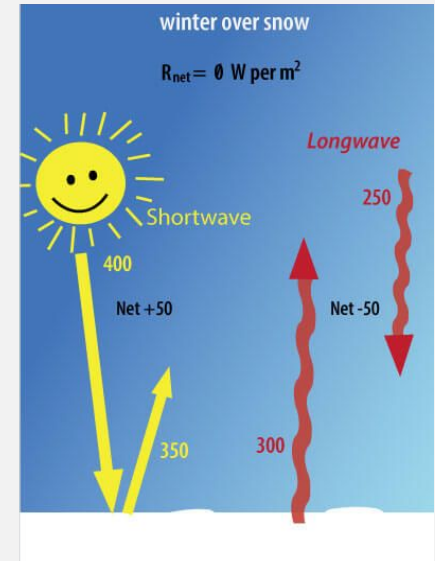
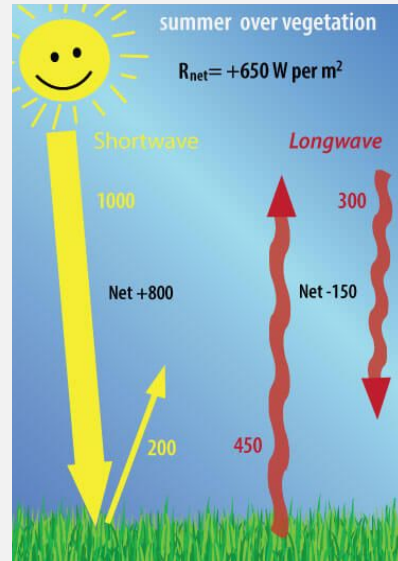


- Units: seconds per day of daylight
- Daylight can infer the season
- Considered very significant in the model



Short Wave Radiation

- Units: W/m^2
- Considered very significant in the model
- The intensity of the sunlight reflected off the earth

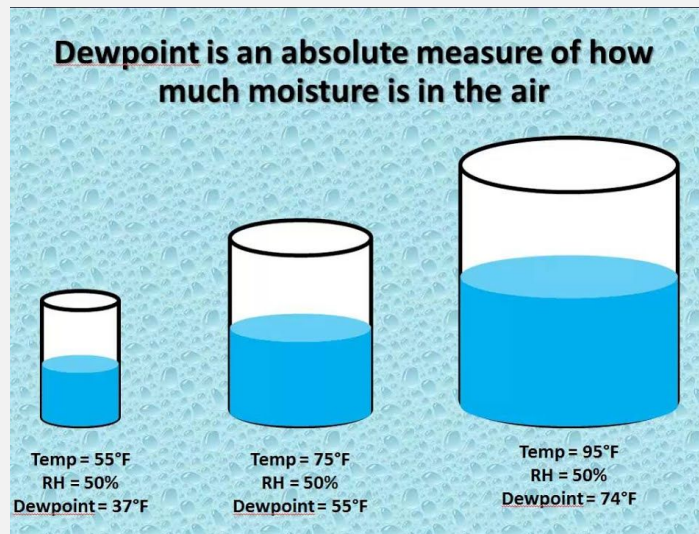




Vapor Pressure



- Units:Pa
- Measures humidity
- Considered very significant in the model
- Low pressure indicates active weather
 - High winds
 - Rain

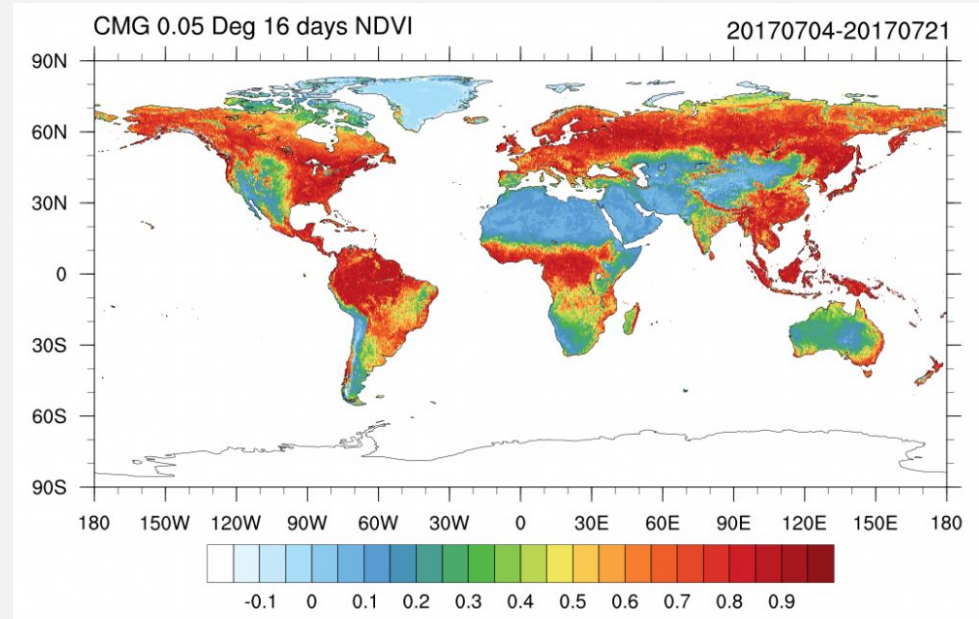




Vegative Indexes

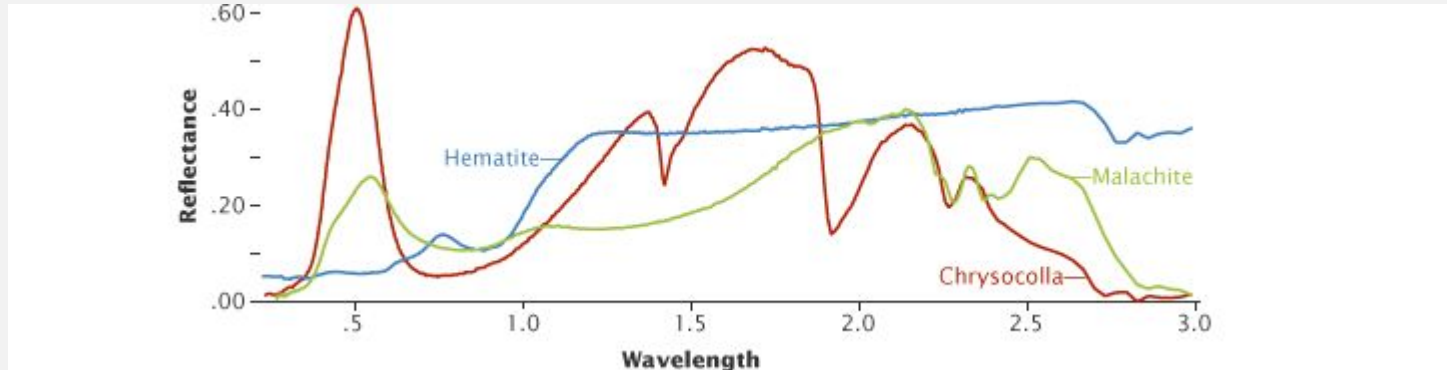


- Enhanced vegetation index (EVI)
- EVI was considered very significant
- Normalized vegetation index (NDVI)
- NDVI was considered insignificant



Near Infrared Reflectance Spectroscopy

- Unit:nanometers
- Different organisms absorb different wavelengths of infrared waves
- Helps remotely classify different organisms
- Considered very significant in the model





Classification Methods

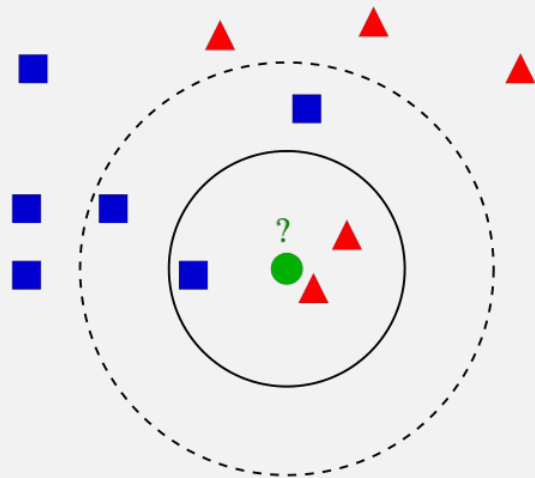
04.





K-Nearest Neighbors

- Classifies based on its similarity to neighbors
- K is amount of neighbors to be considered for each new point
- Euclidean distance is used to find the closest K Neighbors
- Hyper parameter optimization found 5 as the optimal K value
- `train(Formula, data,"KNN",CV,tunelength)`



Truth

Prediction

| | Non-Fire | Fire |
|---------|----------|------|
| NonFire | 50 | 4 |
| Fire | 0 | 43 |



Classification Tree

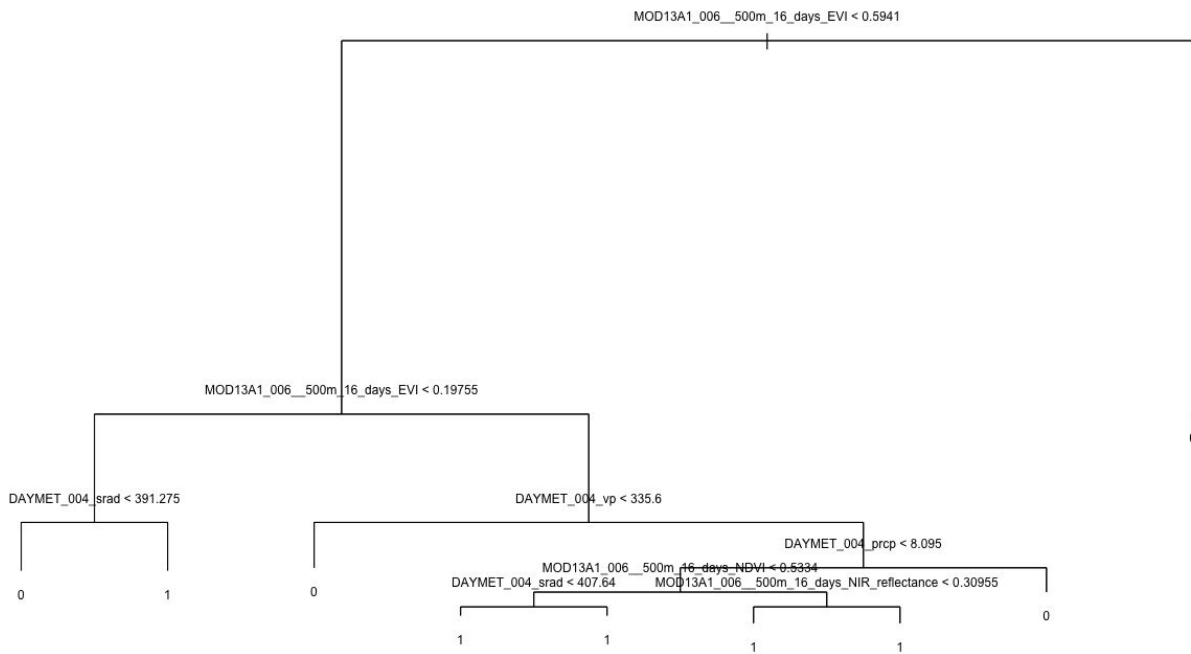
- A mapping of binary decisions that leads to a classification
- Best visually understandable classification method
- Splitting each decision into subpopulation based off binary decisions
- `tree(Variable Formula,dataframe)`

| | Truth | |
|------------|----------|------|
| | Non-Fire | Fire |
| Prediction | | |
| Non-Fire | 40 | 6 |
| Fire | 1 | 50 |





Tree Visualization





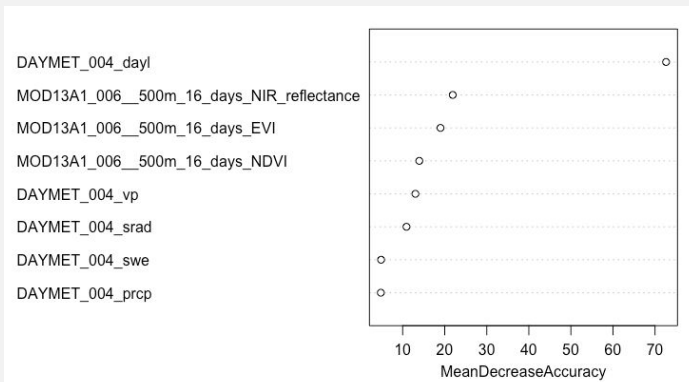
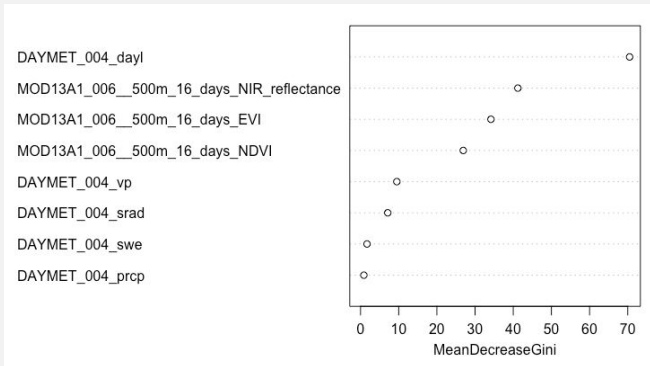
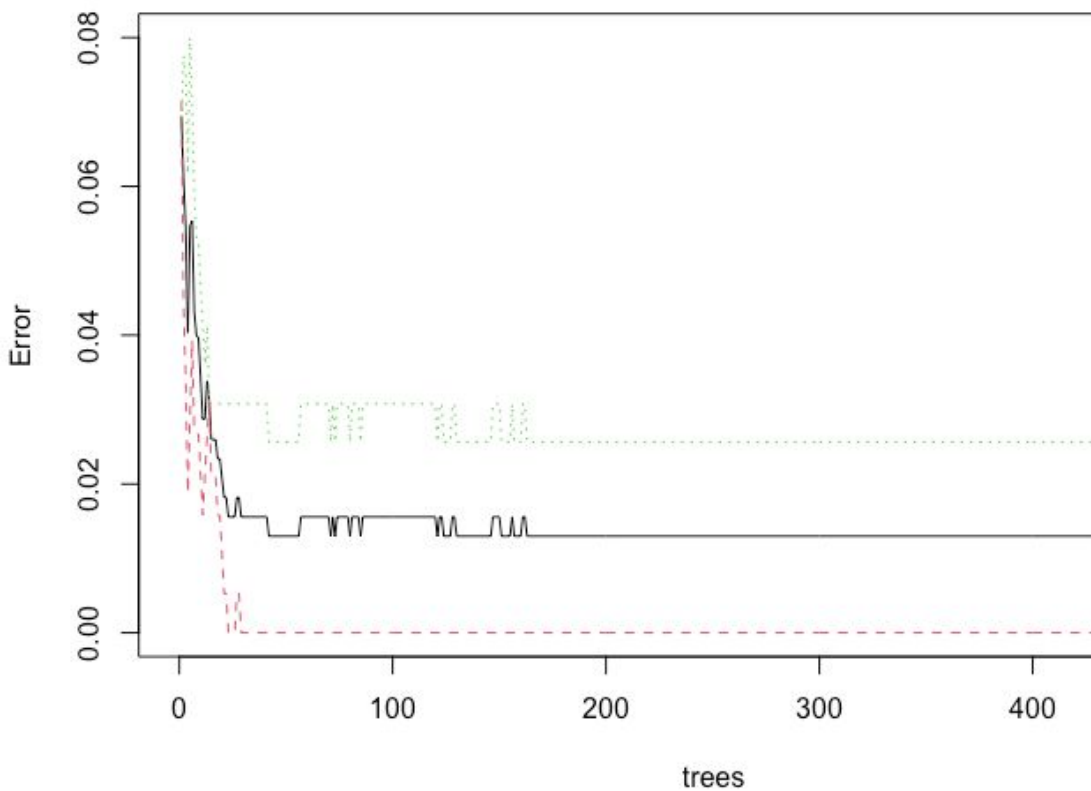
Random Forest

- Hyperparameter tuning found number of variable to be considered at each step to be 3
- `train(Formula, data, subset, "RF", tuneGrid, trControl)`

| Prediction | Truth | |
|------------|----------|------|
| | Non-Fire | Fire |
| Non-Fire | 50 | 1 |
| Fire | 0 | 46 |



Random Forest Visualization





Naive Bayes

- Probabilistic classification
- Applies Bayes theorem with a strong assumption of independence between the variables
- `naiveBayes(Formula, data, trControl)`

LIKELIHOOD

The probability of "B" being True, given "A" is True

PRIOR

The probability "A" being True. This is the knowledge.

$$P(A|B) = \frac{P(B|A).P(A)}{P(B)}$$

POSTERIOR

The probability of "A" being True, given "B" is True

MARGINALIZATION

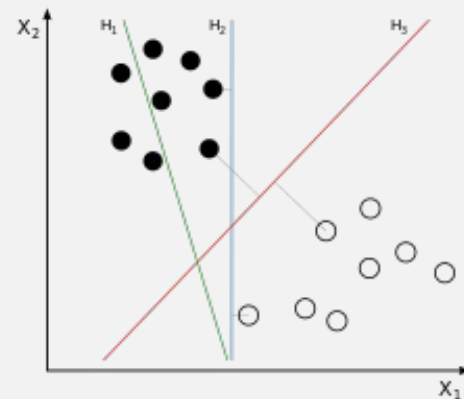
The probability "B" being True.

| | | Truth | |
|------------|----------|----------|------|
| | | Non-Fire | Fire |
| Prediction | Non-Fire | 31 | 35 |
| | Fire | 0 | 66 |



Support Vector Machines

- Supervised learning method
- Non-probabilistic binary linear classifier
- `tune(svm, Formula, data, kernel, ranges, trControl)`
- `svm(Formula data, kernel, cost, scale)`



Truth

Prediction

| | Non-Fire | Fire |
|----------|----------|------|
| Non-Fire | 44 | 2 |
| Fire | 4 | 46 |





Results






| Classification Method | Accuracy | Runtime in Seconds |
|-----------------------|----------|--------------------|
| KNN | 95.8% | 3.1 |
| Classification Tree | 92.7% | .7 |
| Random Forest | 98% | 40 |
| Naive Bayes | 73.5% | .4 |
| SVM | 93.8% | 1.09 |

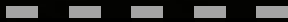




Demo



Raging wildfire. Photo: SciJinks, NOAA





03.

Conclusion





Summary

- Classifies areas that are susceptible to wildfires
- Compared different classification models
- Performed with NASA satellite data





Challenges



- Getting useable satellite data
- Time between data requests
- File organization





Future Work



- Results could be found in real time with a large server
- Coordinate with government models
- Add live weather data
- Factor humans into the model





Acknowledgements



The University of Miami

Dr Aguiar





Github



<https://github.com/JamesMcSweeney/WildFires>

