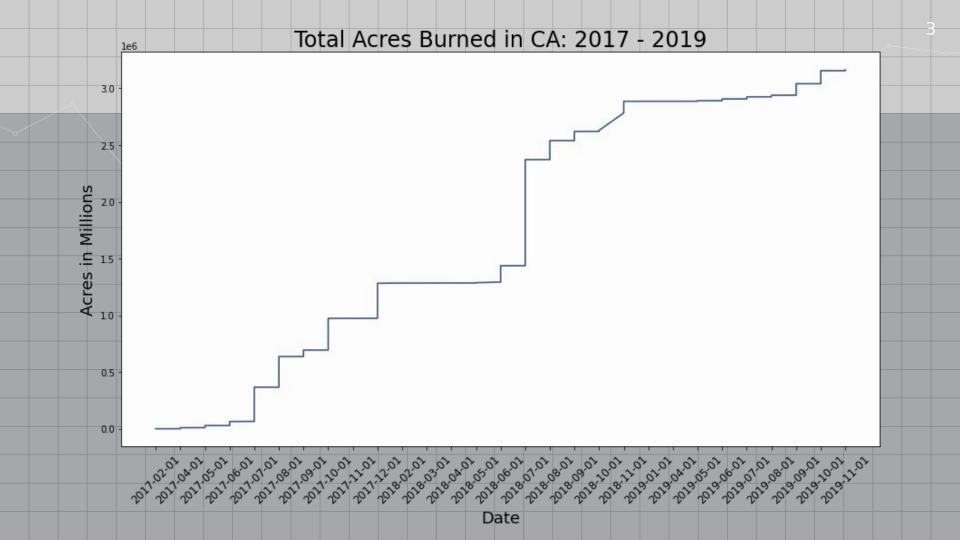
# California Wildfires

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## Problem Statement

We've set out to predict the potential severity of future fires based on past fire data, weather patterns and atmospheric pressure within California.



STEP 1 Gathering Data

STEP 2 Analysing Data

STEP 3 Creating a Model

# Gathering Data

### DATA GATHERING: 3 TYPES









Atmospheric Carbon Data



NOAA Weather Data (12 unique CSVs)



### GENERAL DATA CLEANING

- Drop high percentage of nulls
- Drop categories that were irrelevant(ex. all fires 100% contained)
- Average weather station data by county
- Rounded fires to counties and month.

#### MERGE 15 DATAFRAMES: Weather Stations

STATION	NAME	LAT	LONG
USW0002 3190	SANTA BARBARA MUNICIPA L AIRPORT, CA US	34.4258	-119.8425
USW0002 3232	SACRAME NTO EXECUTIV E AIRPORT, CA US	38.5069	-121.495





#### **COUNTY NAME**

Santa Barbara

Sacramento

Image from https://geopy.readthedocs.io/en/stable/

# Analysing Data

### FEATURE ENGINEERING

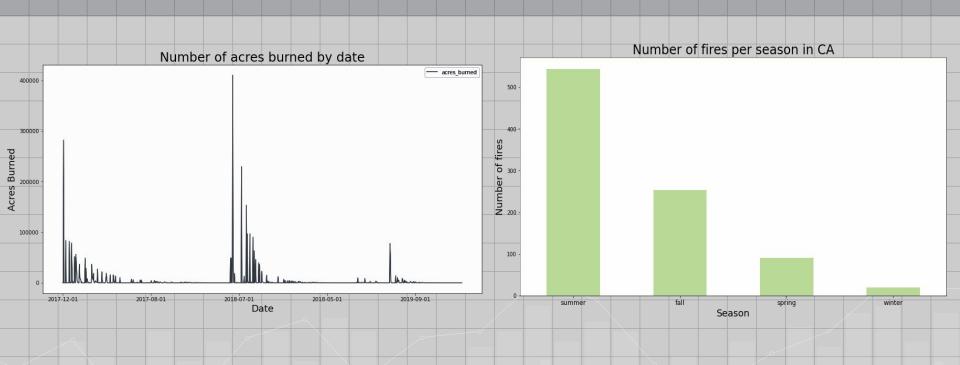
#### **COLUMNS CREATED:**

- Month
- Season
- Duration (of fire)
- Bins (of acres burned)

#### FEATURE ENGINEERING: Bins

- Bin 1: 50 acres or below
- Bin 2: 100 acres or below, and greater than 50 acres
- Bin 3: 250 acres or below, and greater than 100 acres
- Bin 4: 500 acres or below, and greater than 250 acres
- Bin 5: 1000 acres or below, and greater than 500 acres
- Bin 6: Greater than 1000 acres

## FEATURE ENGINEERING: Seasons



## FEATURE ENGINEERING: Duration

Duration = Fire Extinguished - Fire StartedDuration ratio = Acres Burned / Duration



## CORRELATION FINDINGS: Weather Data

Avoid Multicollinearity:

Drop weather data that is over 80% correlated to one another

# Modeling



#### TARGET & BASELINE ACCURACY

- Multiclass Classification Problem
- Our target is to classify fires into bins representing severity
  - An engineered feature based on acres burned
  - Fires are binned smallest to largest, 1 to 6
- Baseline Accuracy: 39%
  - Bin 1: 50 acres or below

#### MODELS

#### **Logistic Regression**

- A simple model to predict categorical outcomes
- Similar to a linear model in structure

#### KNN

- Supervised Machine Learning
- Assumes if points are near each other they are related

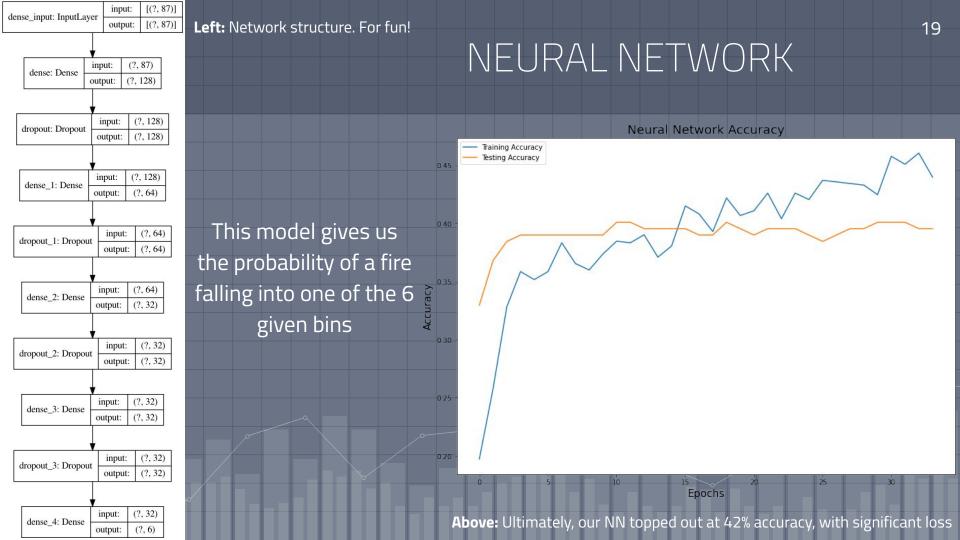
#### MODELS

#### **Random Forest**

- An industry standard model for classification
- Improves decoupling of our decision trees
- Every tree in the forest gets a "vote"

#### SVC

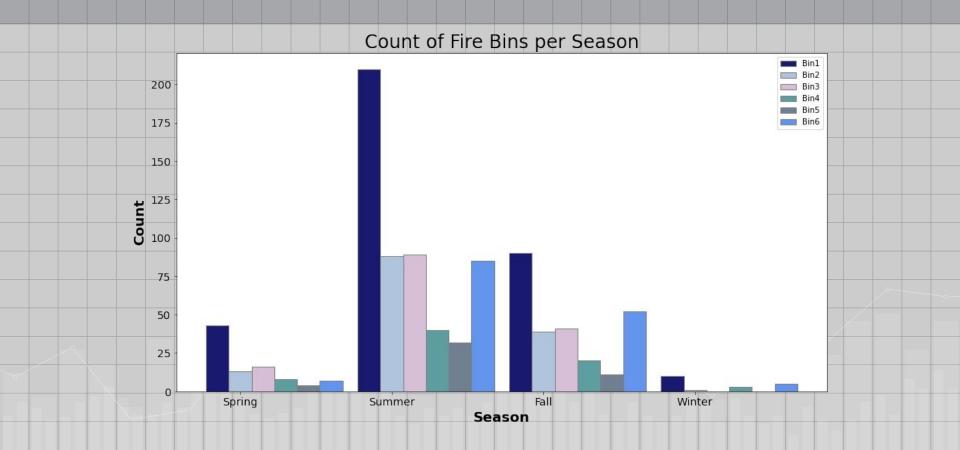
- Kernel Support Vector Classifier
- Our model displays moderate accuracy, but lacks potentially crucial interpretability



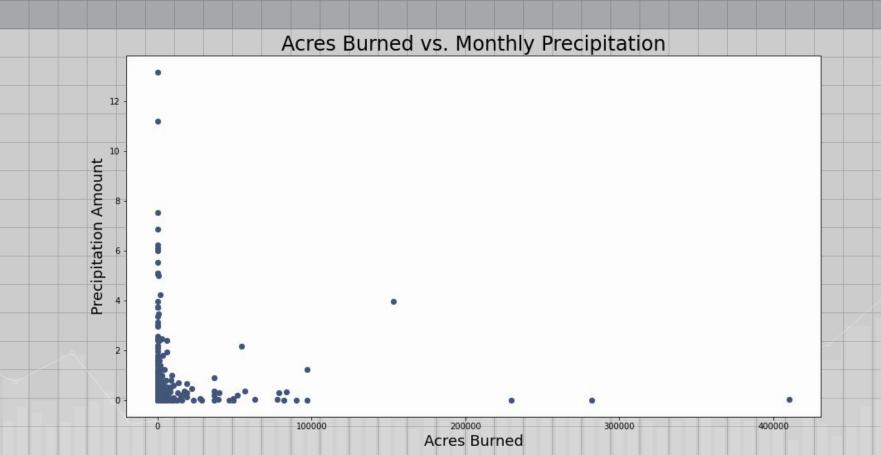
## Results

	Training Accuracy	Testing Accuracy
Logistic Regression	55.58%	42.85%
Support Vector	48.97%	51.65%
K Nearest Neighbor	43.72%	38.46%
Random Forest	99.86%	100%

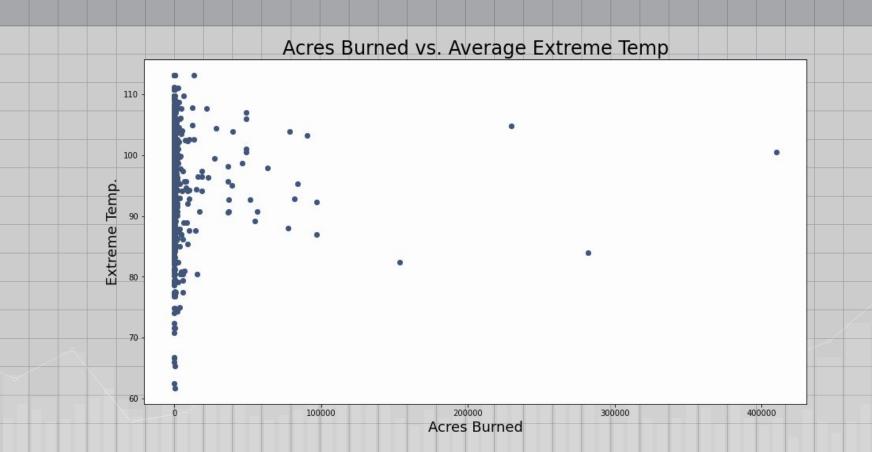
## RESULTS: Seasonality and Severity of Fires



## RESULTS: Acres Burned to Average Monthly Prec.



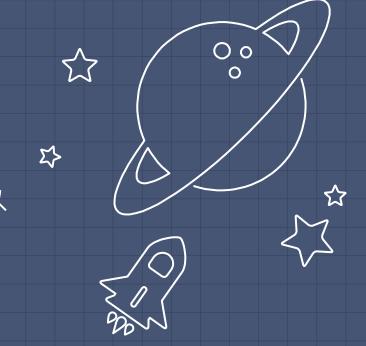
## RESULTS: Acres Burned to Average Extreme Temp



## RESULTS: Acres Burned to Average Elevation



Recommendations & Conclusions



## RECOMMENDATIONS: Hire extra crews for

- Counties in higher elevations
- If there has been low rain amounts for the season
- If there has been an extreme monthly temp. above 90 deg.
- Summer/Fall season (specifically July you can expect a lot of small fires to occur)

## RECOMMENDATIONS: Helping at risk areas

- Controlled burns in at risk areas
- Clear fuel for wildfires
- Reallocation of resources and firefighting units based on environmental and weather conditions

### FURTHER EXPLORATION:

- Look at wind speeds, look at daily numbers instead of monthly
- Compare rural vs. urban areas
- Look at a longer time period
- Research climate change in more depth

Questions?

#### SOURCES

- Kaggle California Wildfire Incidents Dataset
- NOAA Meteorological Datasets
- NOAA/Mauna Loa Observatory Atmospheric Carbon Dataset
- California Department of Forestry and Fire Protection