



FIRE RISK

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Team Burnout





FIRE IS A CLEAR AND PRESENT DANGER

2020 was the most active fire season in the Western United States's recorded history.

Worst Fire Season in CA, Worst in a decade AZ, Most destructive OR, Largest Fires on Record in WA and CO

2020 Fire Damage Report



Land Up in Flame

10.2 million acres of land burned.



Carbon Emissions

Oregon and California broke records for carbon emissions



Lives Lost

46 (32 in California, 11 in Oregon, 1 in Washington, 2 in Colorado)

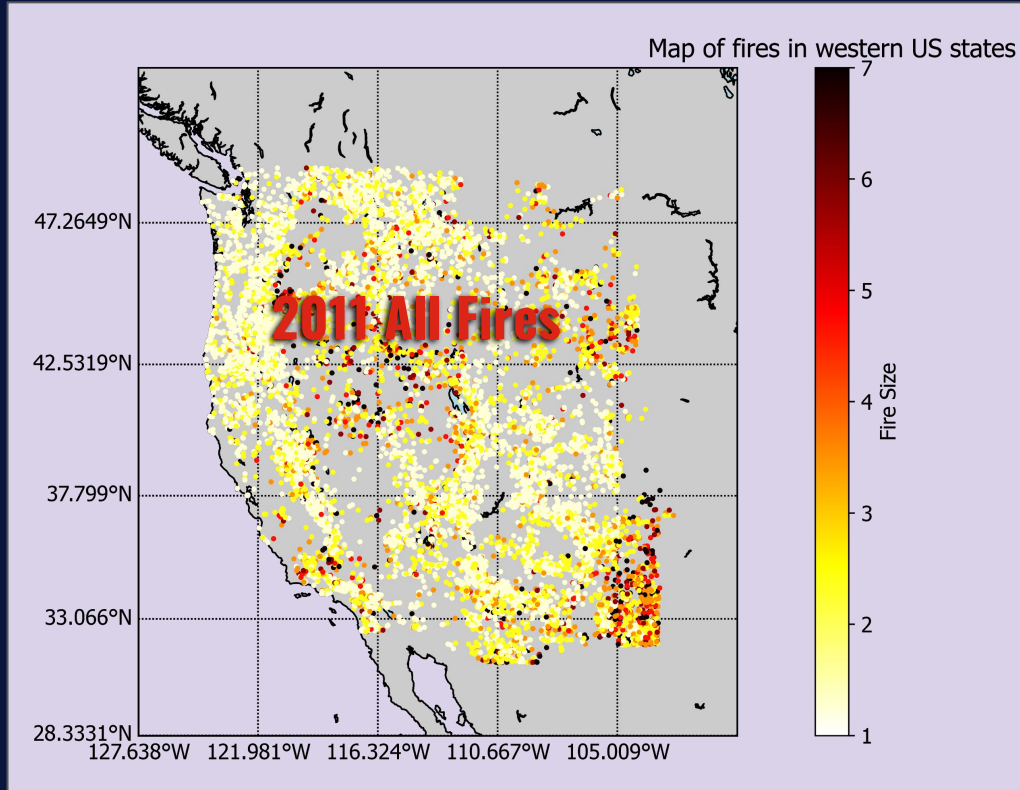


Properties Lost

10,488 structures in CA
\$19.884 billion Cost



COVERAGE AREA



-  Arizona
-  California
-  Colorado
-  Idaho
-  Oregon
-  New Mexico
-  Nevada
-  Montana
-  Utah
-  Washington
-  Wyoming

OUR DATASETS



WEATHER

120 years' worth of
precipitation,
temperatures, and drought
indices



FIRES

24 years of fires' history,
incl. cause, location, final
size (in acres), and dates

CLEANING CHALLENGES



DATA CLEANING

Cleaning our data:

- Standard stuff like lowercase columns etc.
- Very careful about looking at **NaNs**
- New Challenge: Date Time - **Julian to Gregorian**
- Limited States to the **West**
- Limited years to **1987 - 2015**
- Removed Outliers - Fires that burned **over 5,000 days**



COMBINING DATA SETS

Combining our data:

- Data Size was a limitation
- Set up **AWS S3** bucket
- Set up **AWS CLI** for the team
- Merge file difficulties - **memory**
- Unable to get all data that we wanted
- SQL to pkl to csv fun



CHALLENGES

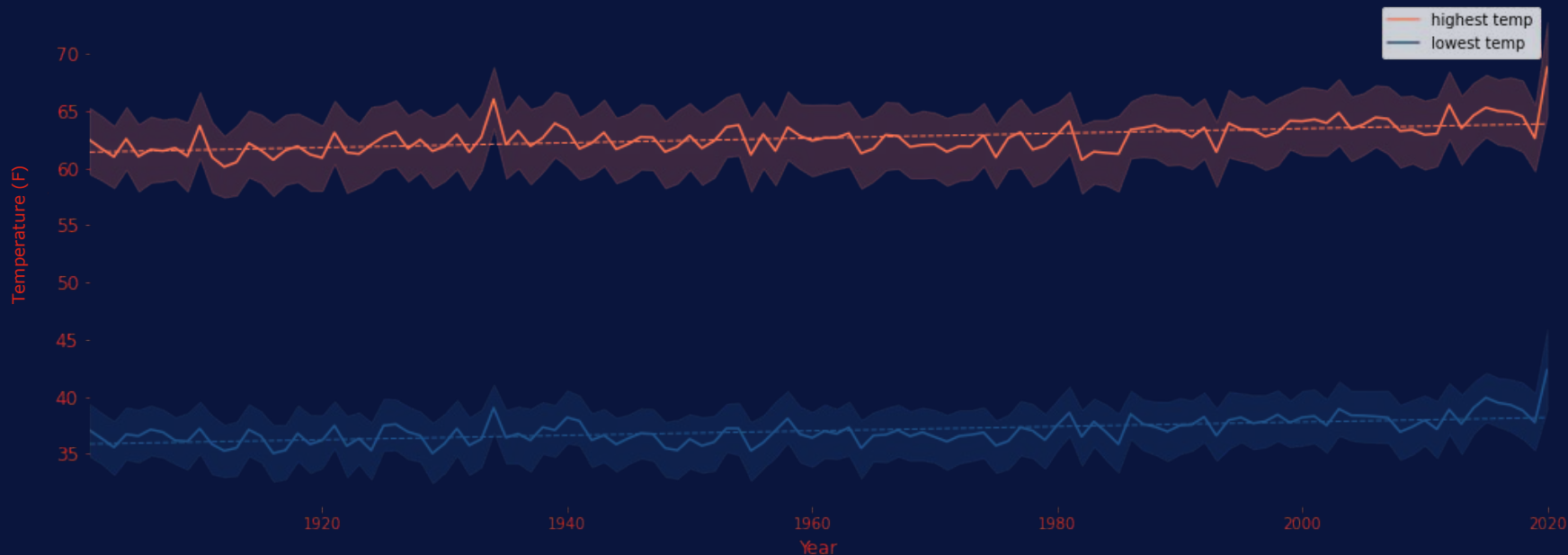
Difficulties with data:

- We realized that adding **foliage** and **wind** data would have been extremely helpful.
- Briefly considered trying to do something with **lightning**. - That consideration was quickly squashed.
- Limited data sets - Fire data only had **1992 - 2015**
- **Resource** limitations (processing power!)



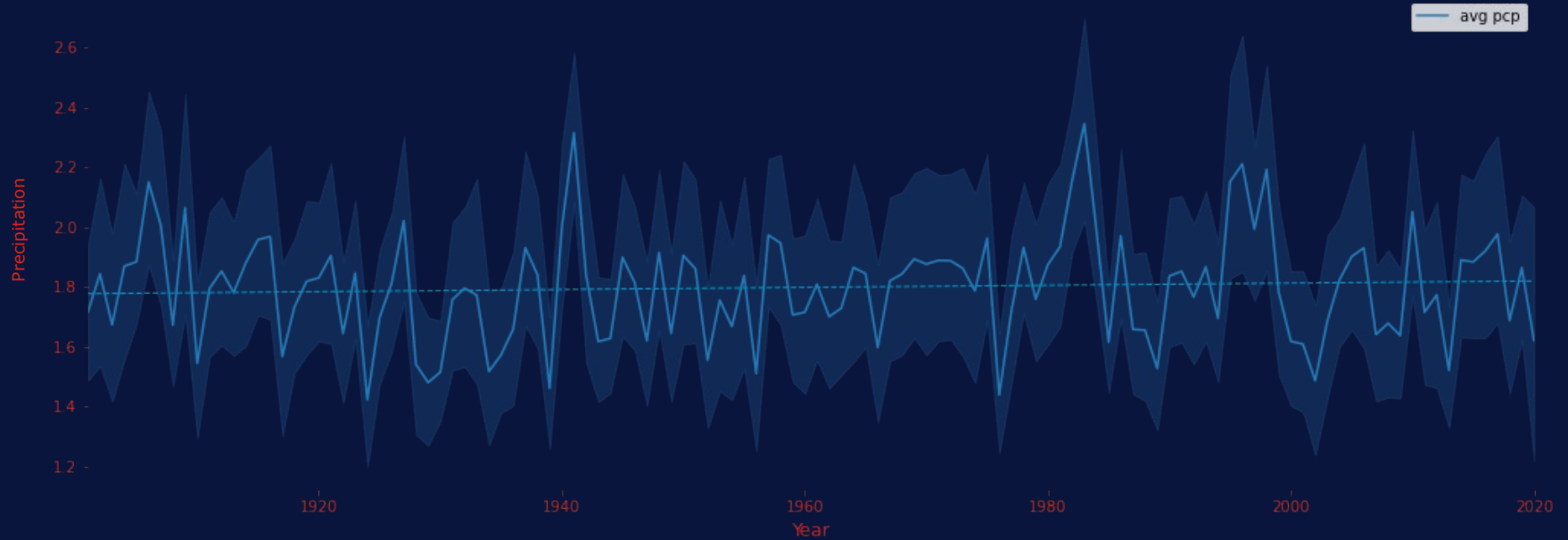
WEATHER TRENDS ARE SLOW BUT POWERFUL

Lowest and Highest Temperature by Year



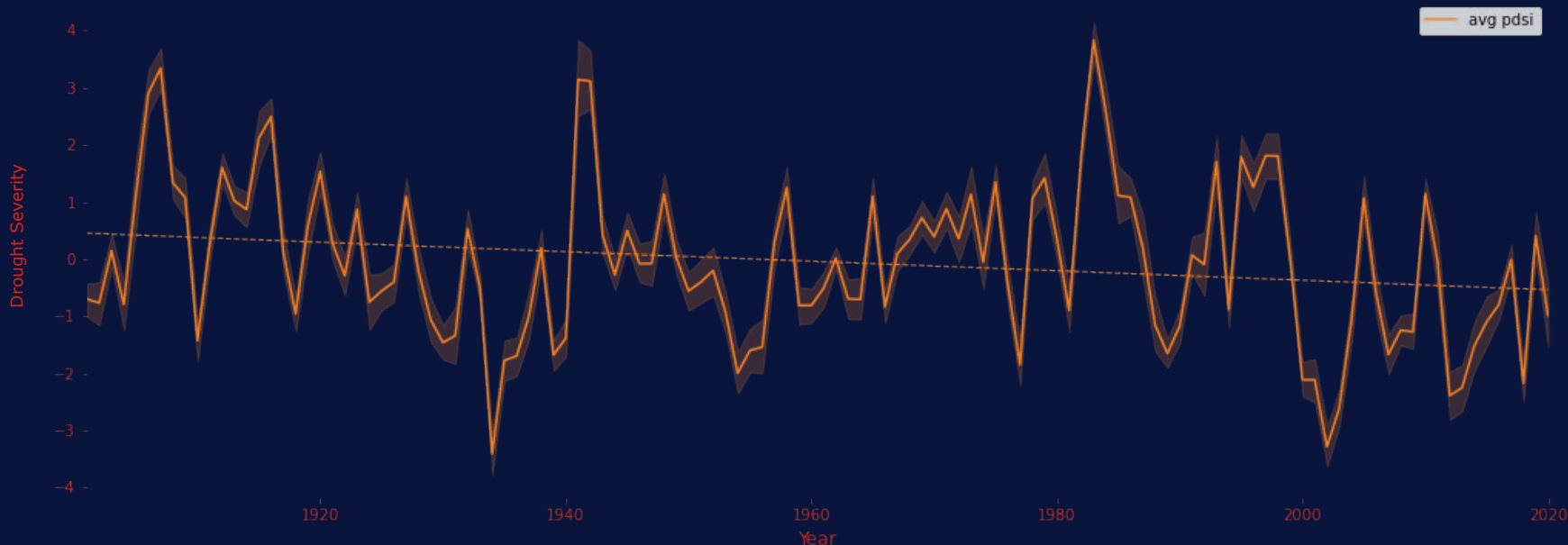
WEATHER TRENDS ARE SLOW BUT POWERFUL

Precipitation Index by Year



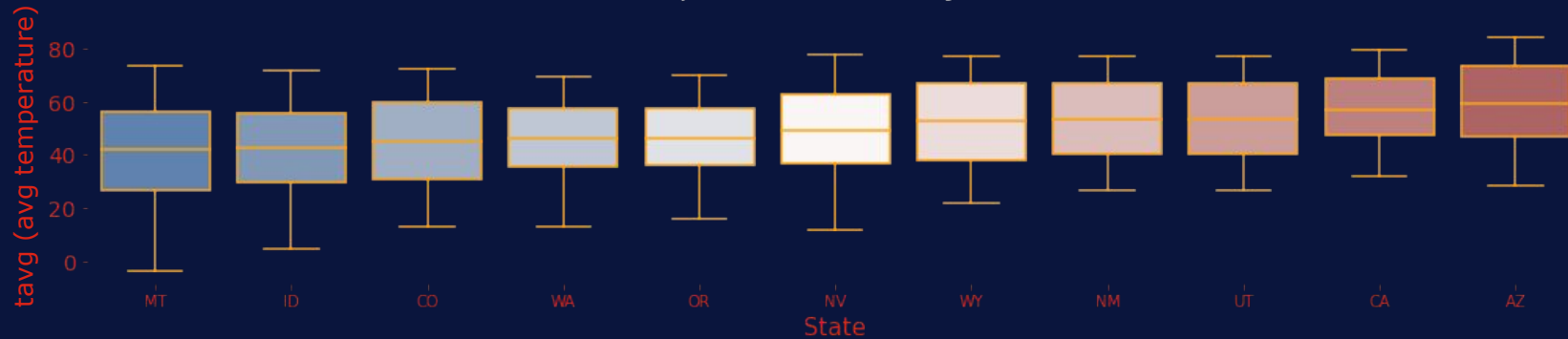
WEATHER TRENDS ARE SLOW BUT POWERFUL

Average Palmer Hydrological Index Values by Year (-10 = dry, 10 = wet)

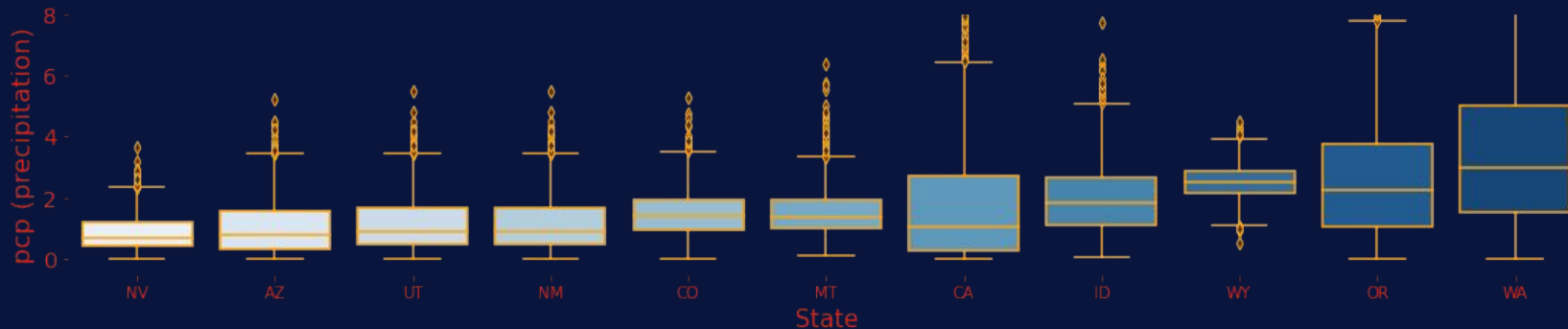


NOT ALL STATES ARE CREATED EQUAL

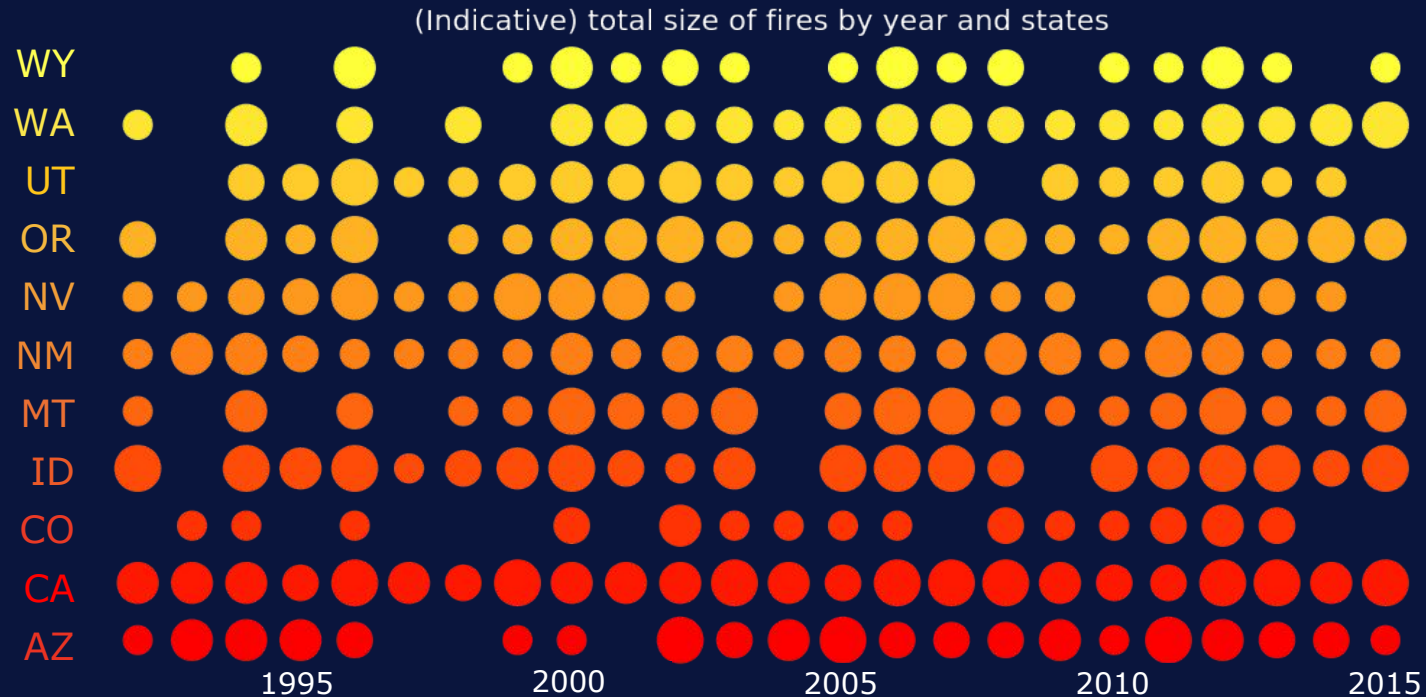
Temperature Index by State



Precipitation Index by State

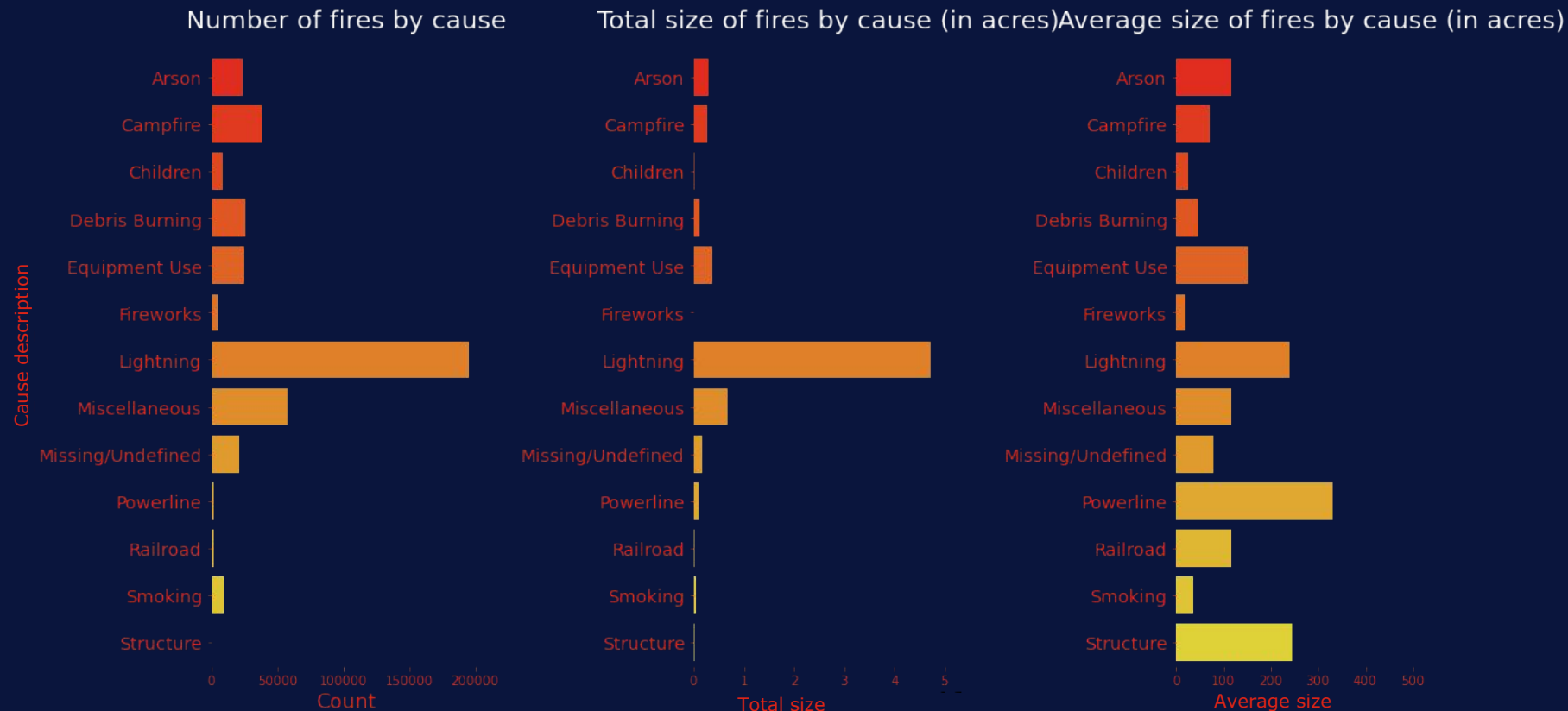


... AND NOT ALL STATES FEEL THE BURN EQUALLY



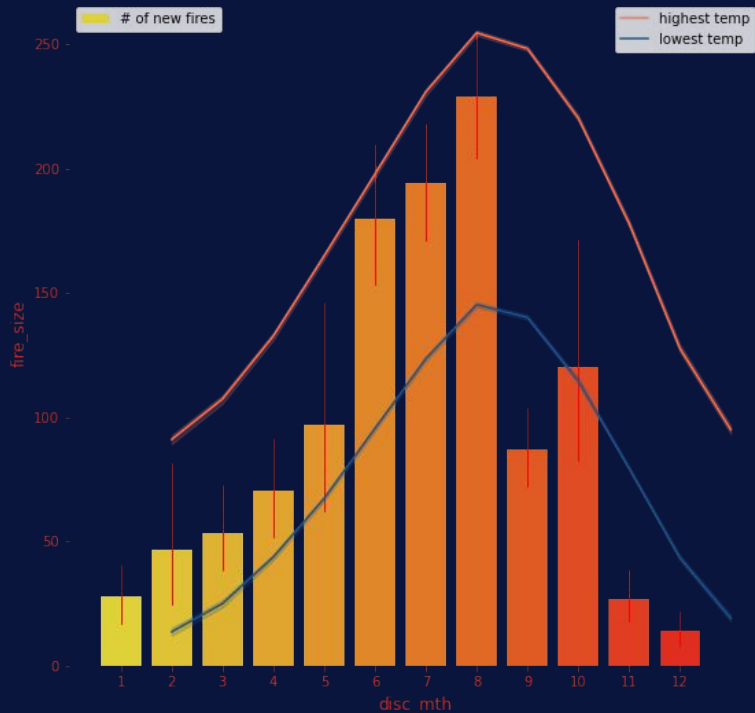
● - total fires > 500,000 acres | ● > 200,000 | ● > 100,000 | ● > 50,000 | ● > 25,000

WE DIDN'T START THE FIRE... SO WHO DID?

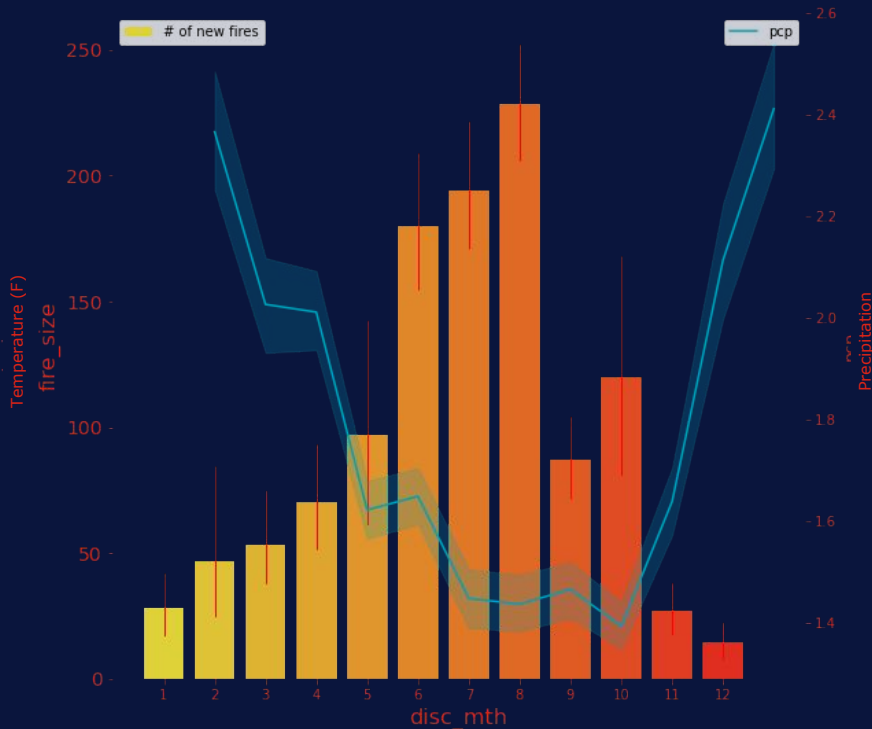


HIGH TEMPS + LOW PRECIPITATION = FIRE

Avg temperature ranges vs avg final size of fires by month of discovery



Avg precipitation index vs avg final size of fires by month of discovery



MODELING

REGRESSION

- Not the best fit model type for the scope of this project.
- Ridge Regression came out at around 15% R^2

THE MODEL IMPROVEMENT PROCESS

- Classification
- Bootstrapping
- Clustering
- Trailing Averages
- Class Weighting

CLASSIFICATION



RANDOM FOREST

- Purpose: feature importance

SVM

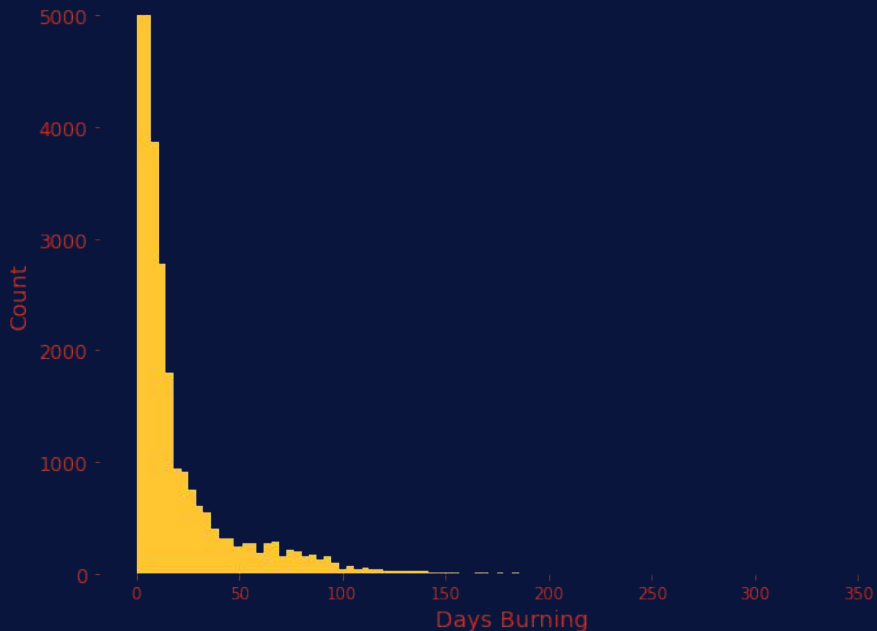
- Support Vector Machine with bootstrapped data

NEURAL NETWORK

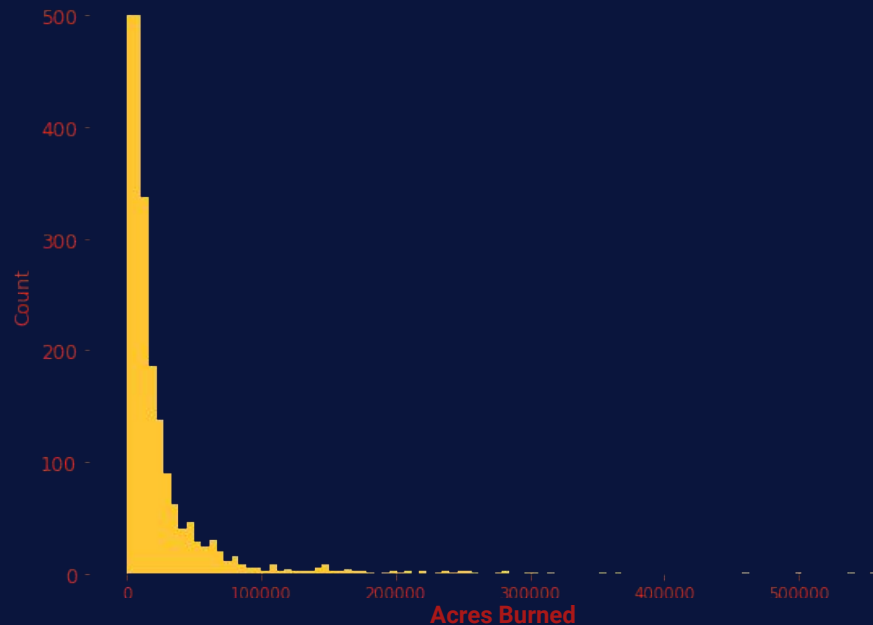
- The Neural Network had the highest Recall Rate across the most important classes.
- This is the best predictive model.

OUR CLASSES WERE MASSIVELY IMBALANCED

Histogram of Fires Duration (days)



Histogram of Fires Size (acres)

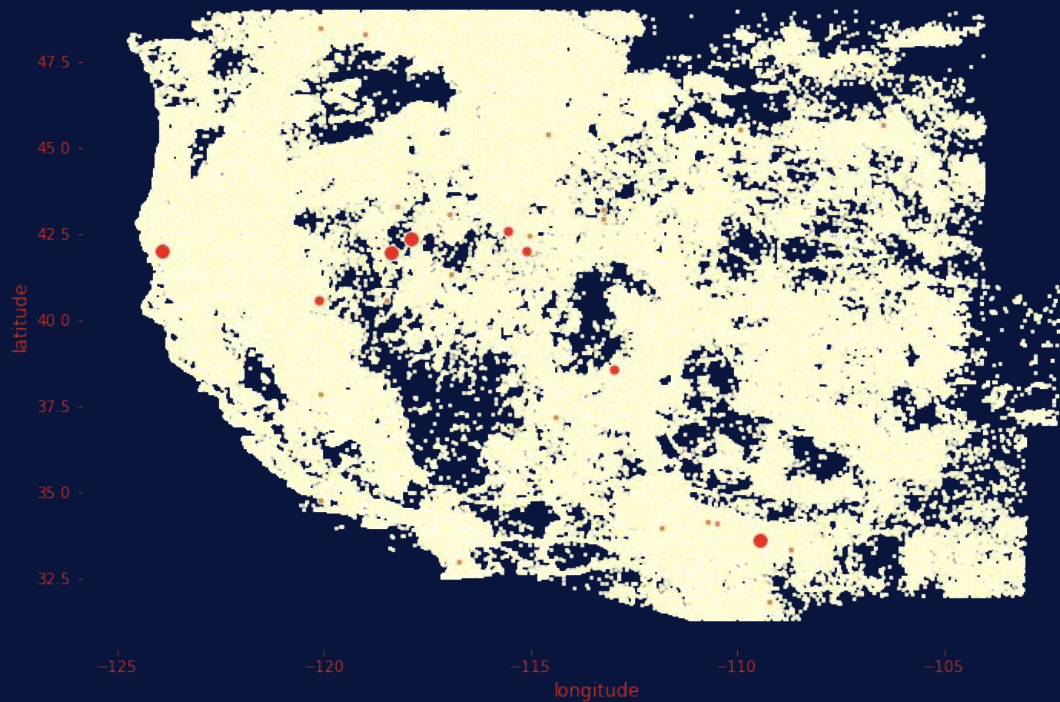


TARGET LABELS:

"A" - fires smaller than 0.25 acres, "B" - 9.9 acres, "C" - 99.9 acres, "D" - 299 acres ,
"E" - 999 acres , "F" - 4999 acres, and "G" - 5000+ acres.

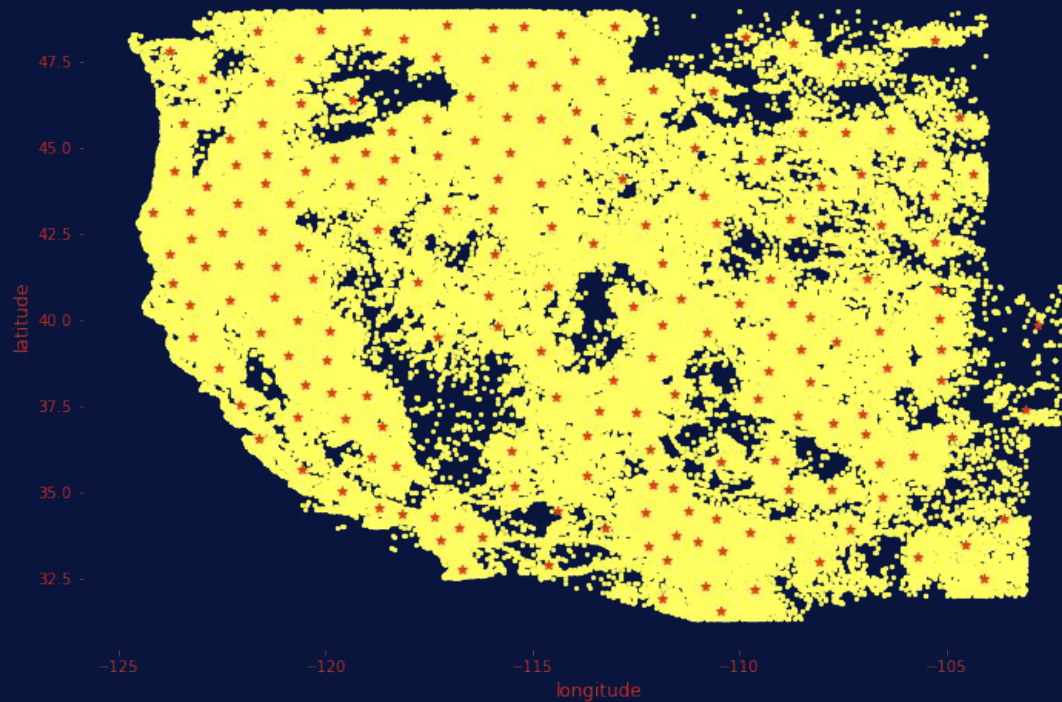
WHY WE CLUSTERED LATITUDE & LONGITUDE DATA

Where are the fires?



WHERE DID THE CLUSTER CENTROIDS LAND?

Location of cluster centroids



THE CHOSEN MODELS

NEURAL NETWORK CLASSIFIER

Purpose: **predictive power**



Optimized for Recall score, not Accuracy. Our highest Accuracy model was >70%. The chosen model came through at closer to 50% Accuracy, but a much higher recall rate.

Wildfire Class by Size and Recall:

Modeling

- A: 59%
- B: 1%
- C: 20%
- D: 50%
- E: 58%
- F: 59%
- G: 86%

Baseline

- A: 62%
- B: 29%
- C: 6%
- D: 2%
- E: 1%
- F: 2%
- G: .07%

RANDOM FOREST CLASSIFIER

Purpose: **feature importance**

Most important feature
(location excluded):

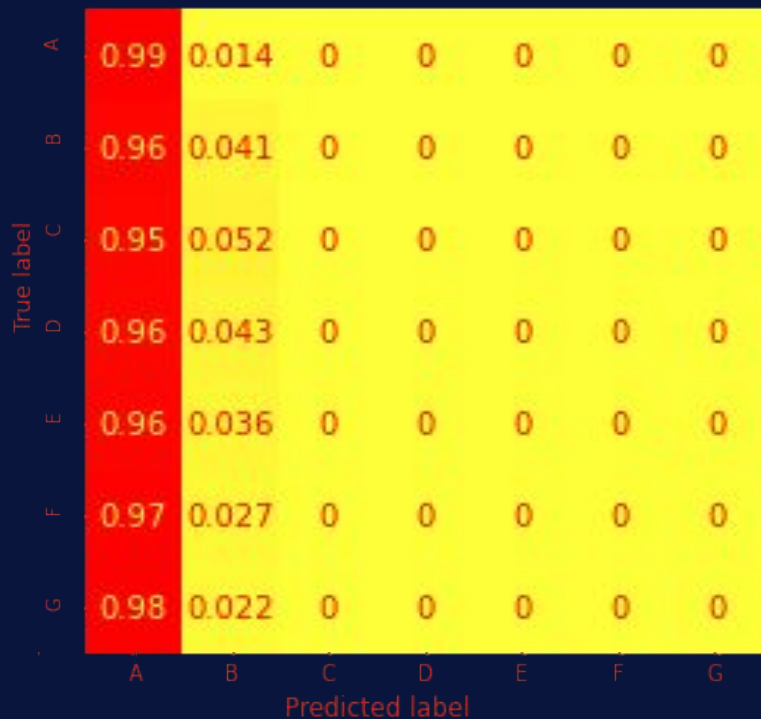
Temperature average past 3 months

Second most important
feature: **Temperature average past 6 months**

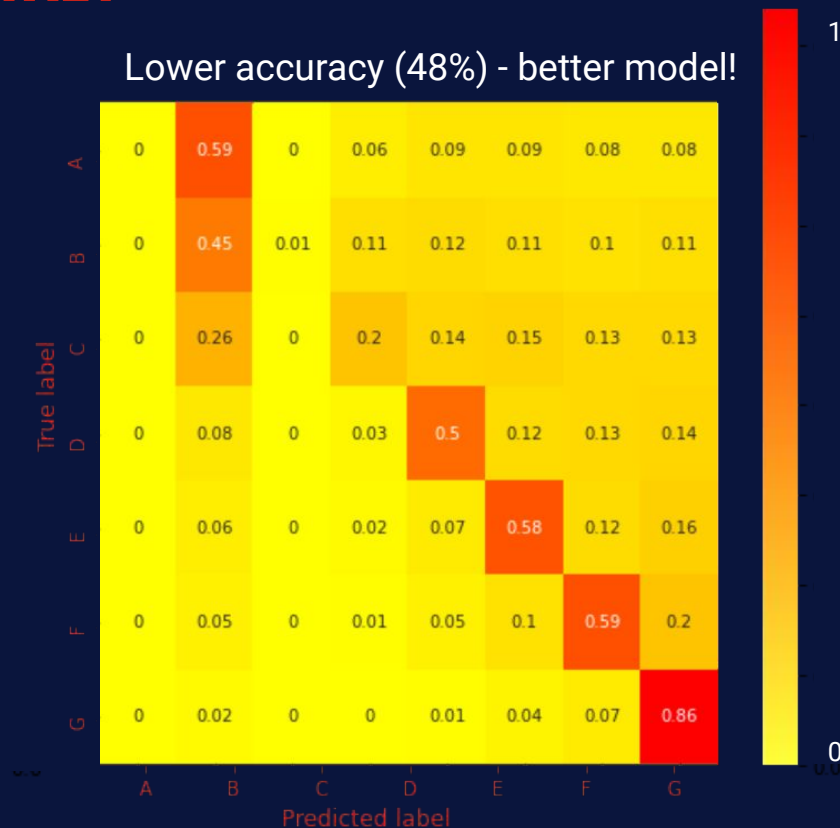
Third most important
feature: **Monthly Precipitation**

WHAT DOES “GOOD” LOOK LIKE?

Higher accuracy (62%) - useless model



Lower accuracy (48%) - better model!



SUMMARY & NEXT STEPS

What we learned

Location is important!



Temp & Precipitation have direct relationships



Bootstrapping helps with unbalanced classes



Focusing on recall helps us predict disastrous fires.



What we would still like to explore

Looking into Local Resources



Adding Altitude and Vegetation Data (Earth Engine)



Add Wind Data (NOAA)



With Climate Change and Warmer Temps



More Risk

Dry and hot conditions
will bring more fire risk.



Need for Security

We will need predictive
models to protect
communities.



More Resources

Being prepared ahead
and being able to
respond will be critical.

QUESTIONS?



THANKS!

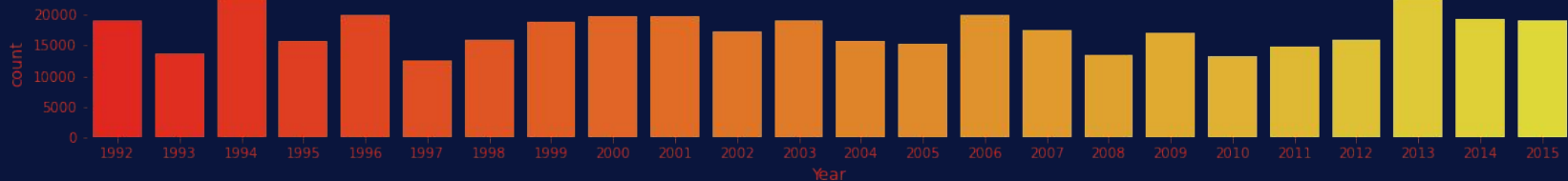


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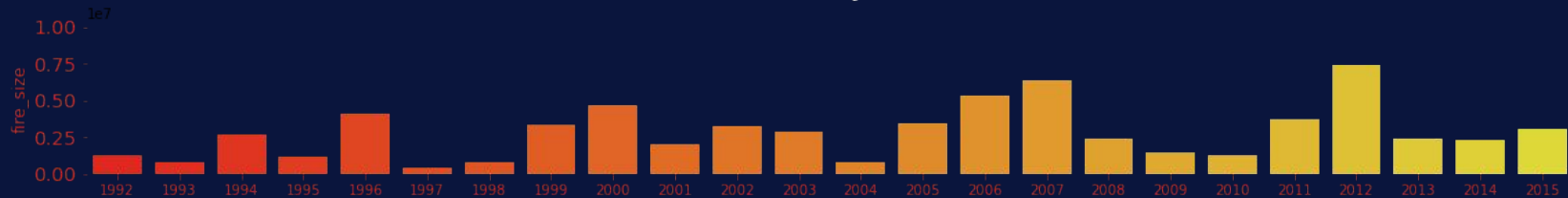
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DO FIRES GET LARGER AS CLIMATE GETS WARMER?

Number of Fires by Year



Total Size of Fires by Year (in acres)



Average Size of Fires by Year (in acres)

