

# PREDICTIVE CLIMATE

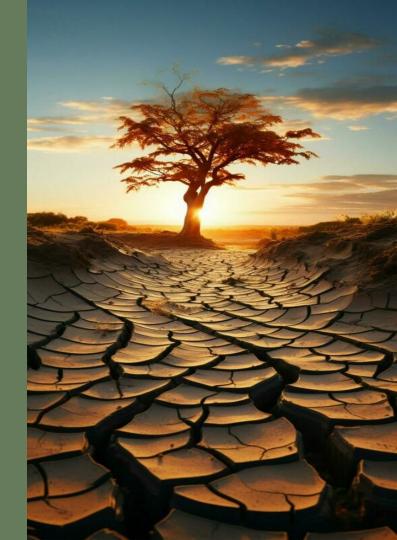
Data Estimating Future Weather Patterns

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DS 4002 – 11/20/23

## PROJECT OVERVIEW

- 1. Project Details
- 2. Data Acquisition + Explanation
- 3. Analysis Plan + Justification
- 4. Tricky Analysis Decision
- 5. Bias + Uncertainty Validation
- 6. Results + Conclusion
- 7. Next steps



# PROJECT DETAILS

#### **MOTIVATION:**

Since 1901, average precipitation and temperature has increased by 0.20 inches and 0.17°F per decade, in the US. (EPA 2023)

Resulting in soil erosion, rising sea levels, and extinctions. Could scientists of the past have predicted this? Today, can we?

#### **RESEARCH QUESTION:**

Can multilinear regression models accurately predict average temperature, based on historical US climate data?

#### **NULL HYPOTHESIS:**

Precipitation, PDSI, minimum temperature, and maximum temperature are NOT accurate predictors of average temperature.

#### **MODELING APPROACH**

Use multilinear regression model on US climate data to predict average temperature and find p-values from model's residuals.

#### **GOAL**

To investigate how accurately a multilinear regression model can predict average temperature based on past US climate data and either reject or fail to reject the null hypothesis.

# Data Acquisition & Explanation

## **Final Dataset**

#### **Data Dictionary**

Column	Description
Date	Date of data entry, providing year and month
Avg Temp	Average temperature in °F
Min Temp	Minimum temperature in °F
Max Temp	Maximum temperature in °F
Precipitation	Average amount of precipitation in inches over the duration of one month
PDSI	Average Palmer Drought Severity Index (PDSI) value over the duration of one month

## Acquisition

- Data gathered from National Centers for Environmental Information.
- 2. Downloaded 5 separate files, each containing one a weather parameter.
- 3. We merged those 5 files into one using Python.

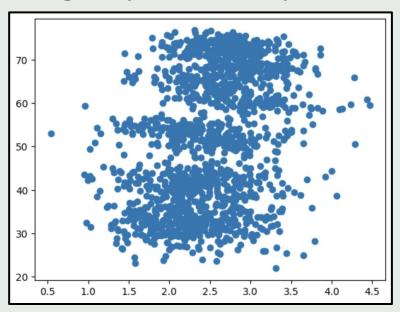
## Cleaning

- 1. Checked the dataset for NA values.
- Removed duplicate variables resulting from merging datasets.
- 3. Created dataset shown to the left.

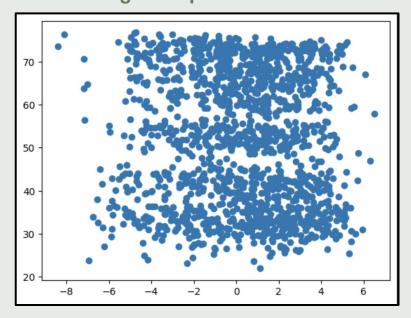
## **Analysis Plan + Justification**

## **EXPLORATORY DATA ANALYSIS**

#### **Average Temperature vs Precipitation**



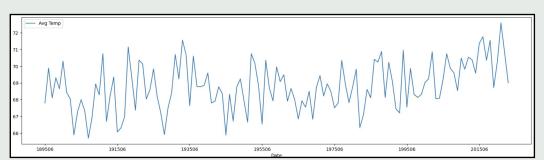
#### **Average Temperature vs PDSI**

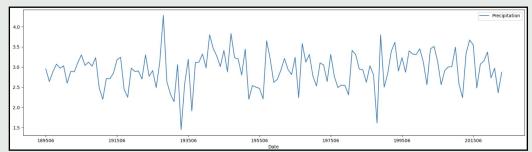


## **EXPLORATORY DATA ANALYSIS (continued)**

Average Temperature in June over time





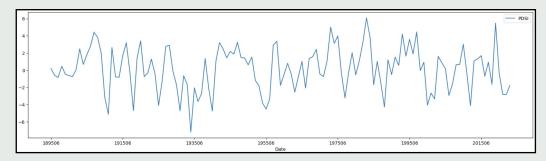


Average Precipitation in June over time



Average PDSI in June over time





# **Analysis Plan**

- 1. After doing our EDA, we noticed patterns in the variables over time.
- 2. We used multilinear regression to create a predictive model of average temperature.
- 3. We found p-values for each of the independent variables.

# TRICKY ANALYSIS DECISION

- Accounting for seasonality in average temperature, minimum temperature, and maximum temperature?
  - Planned to use simple linear regression.
  - Can't use simple linear regression model with highly variable data due to seasonal changes.
- What kind of model to use: ARIMA or Linear Regression?
  - Using the statsmodel package, we performed seasonal decomposition.
  - Removed seasonality from data, subtracting its values from our values.
  - Then allowed for multilinear regression.



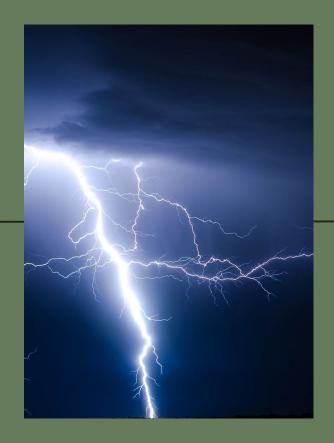
# BIAS & UNCERTAINTY VALIDATION

## **Spatial Considerations:**

a. This data is from all over the United States.
 However, weather and climate conditions vary a lot across the country.

### **Multilinear Regression:**

b. The data was modified to remove the seasonality in order to use multilinear regression. This may have skewed the results.



## **RESULTS & CONCLUSIONS**

## **Precipitation and PDSI**

- The p-value for precipitation was 0.611 and 0.872 for PDSI, meaning that we cannot reject our null hypothesis.
- These variables most likely do not play a role in determining average temperature.

## Minimum and Maximum Temperature

- The p-values for minimum and maximum temperatures were 0, meaning that we can clearly reject our null hypothesis.
- This makes sense, as minimum and maximum temperature play a role in calculating the average temperature.



## **NEXT STEPS**

### Flip our analysis

a. Instead of trying to see if certain variables affect average temperature, we would like to see how other variables can be predicted.

#### Use different models:

b. Ignore seasonality and use ARIMA.

#### **Predict future values**

 Using our model, we would like to see if we could use it to predict future average temperatures, as it was very difficult using the statsmodels package.



## REFERENCES

#### **SOURCE #1:**

[1] NOAA National Centers for Environmental Information, "Climate at a Glance: Time Series," [Online].

Available: <a href="https://www.ncei.noaa.gov/access/monitoring/climate-at-a-glance/national/time-series">https://www.ncei.noaa.gov/access/monitoring/climate-at-a-glance/national/time-series</a>.

[Accessed: March 25, 2024]

GITHUB REPOSITORY LINK: <a href="https://github.com/keivonc/ds4002project2">https://github.com/keivonc/ds4002project2</a>