# **EVOLUTION OF EARTH'S CLIMATE ZONES VISUALIZING SHIFTS ACROSS TIME AND SPACE** Eurogroup Team 003 | CSE 6242 | Fall 2020

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Methodology

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# Summary

#### **Problem and Importance:**

Earth's climate is changing, but no universally accepted model for unsupervised clustering of exists. This is important to help the public and policy makers plan for a future with a changing climate, a global concern.

#### Algorithm

#### Innovation

> Novel use of Fast Fourier Transform for dynamic time warping

> Based on clustering analysis instead of human heuristics.

### Visualization

#### Innovation

> Animated: others of this type are static images

#### Goal:

I. Develop an interactive, dynamic visualization that clusters areas of the Earth into climate biomes using only weather data, avoiding the use of human heuristics

2. Examine if those clustered biomes exhibit a tendency to shift toward the poles.

Data

#### **Source:**

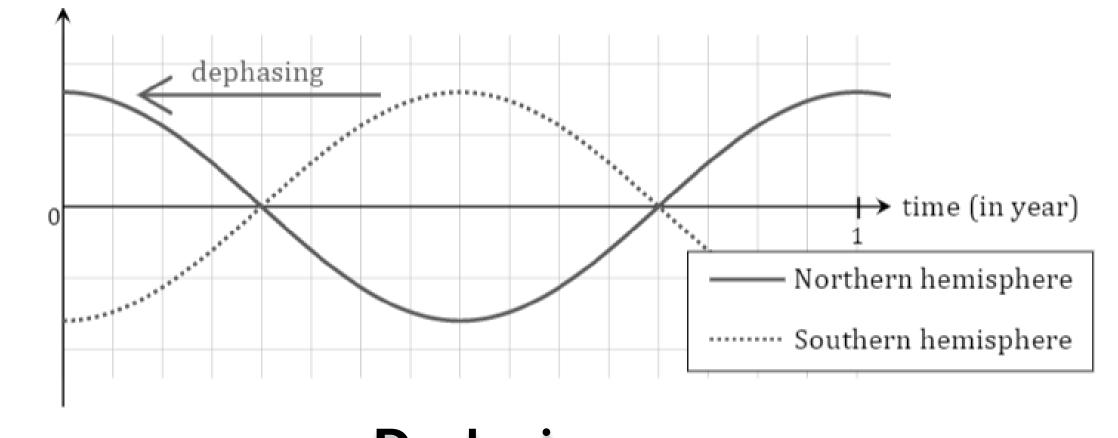
Integrated Surface Dataset detailing hourly weather 1901 observations from station to present. the Obtained and collated from yearly csv files via the National Centers for Environmental Information data website using a Python scraping script.

#### **Inspection & Preprocessing:**

Hourly data condensed to daily averages and spreads

#### Feature Extraction

We developed (generalized tensors representations of basis vectors and components) represent each weather component for to station-year tuples, using Dynamic Time Warping Transforms Fourier dephase Fast and to hemispheric seasonality.



Dephasing

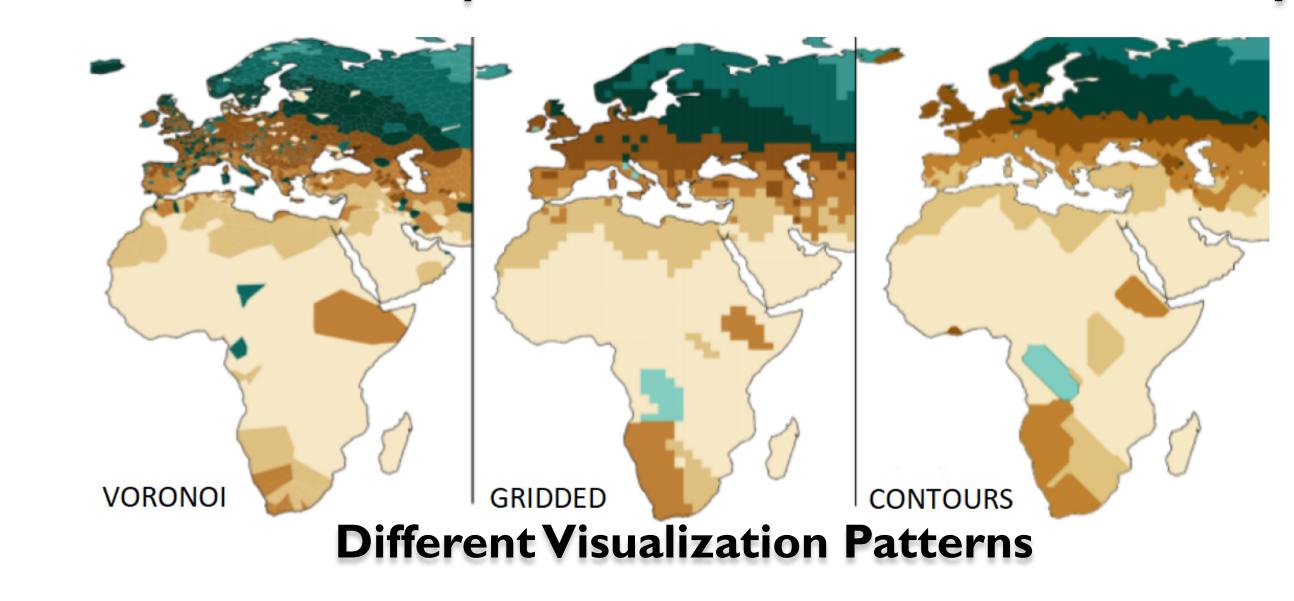
### Clustering

dimensionality reduction (none, > 3-levels of

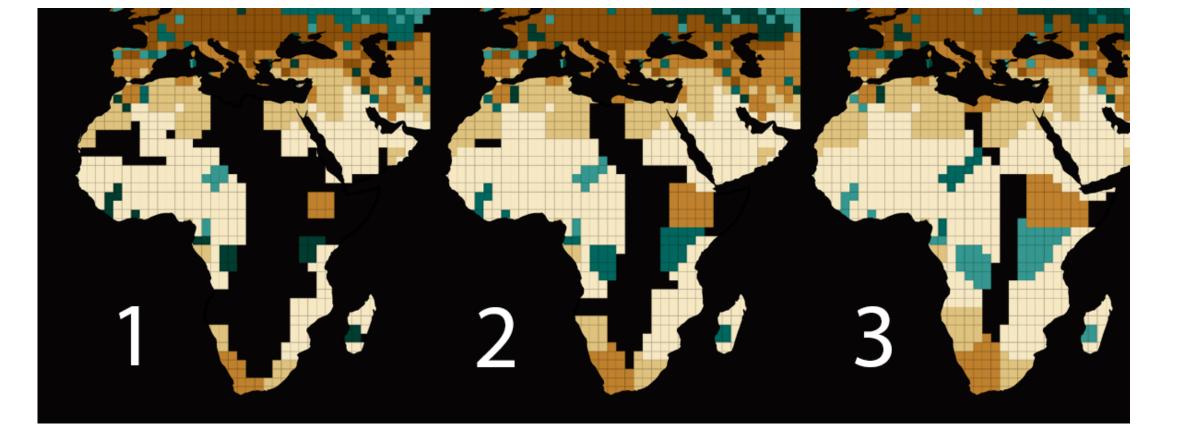
> Interactive: existing implementations do not have interactivity in their visuals

### **D3 Contours**

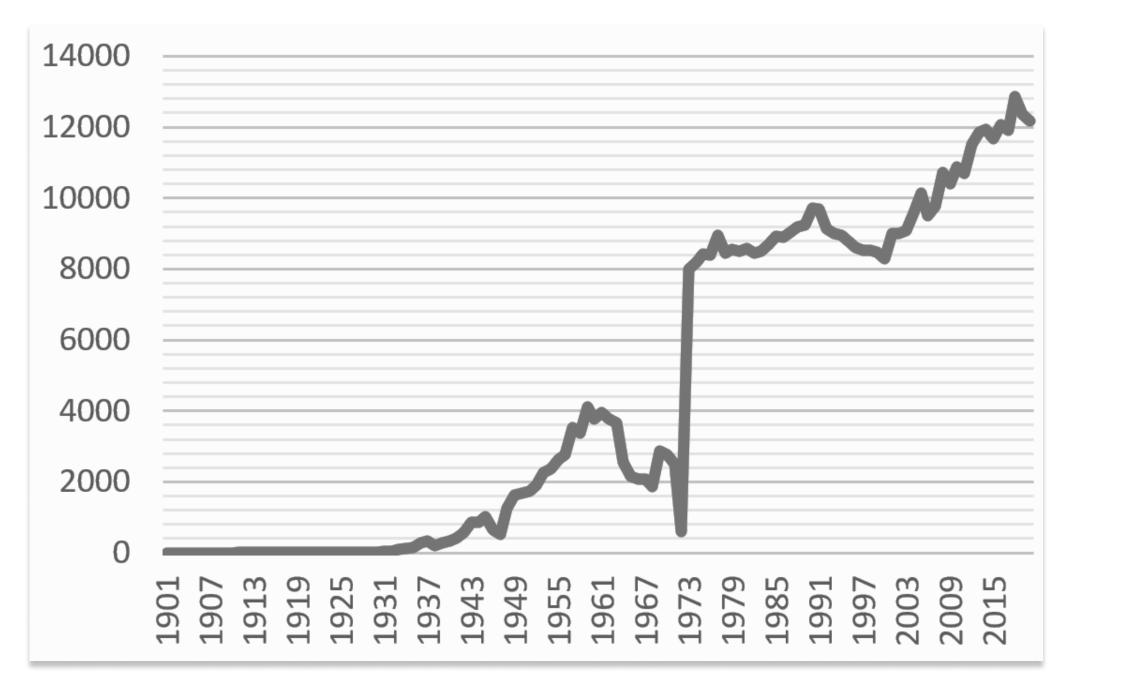
After trying other approaches, we settled on contours for max performance and readability.



**Color Assignment, Filling, Smoothing** 



reducing file size from 1.2 TB to ~34 GB. We removed years below 1974 which lacked a sufficent number of Although 23 stations (see figure). features were available, we chose to retain 5 after analyzing for covariances.



Number of available stations by year

#### **Inference:**

Many stations attribute missing had data or no information for certain periods (hours to days). In these cases, we decided to linearly interpolate where possible.

#### ICA, PCA)

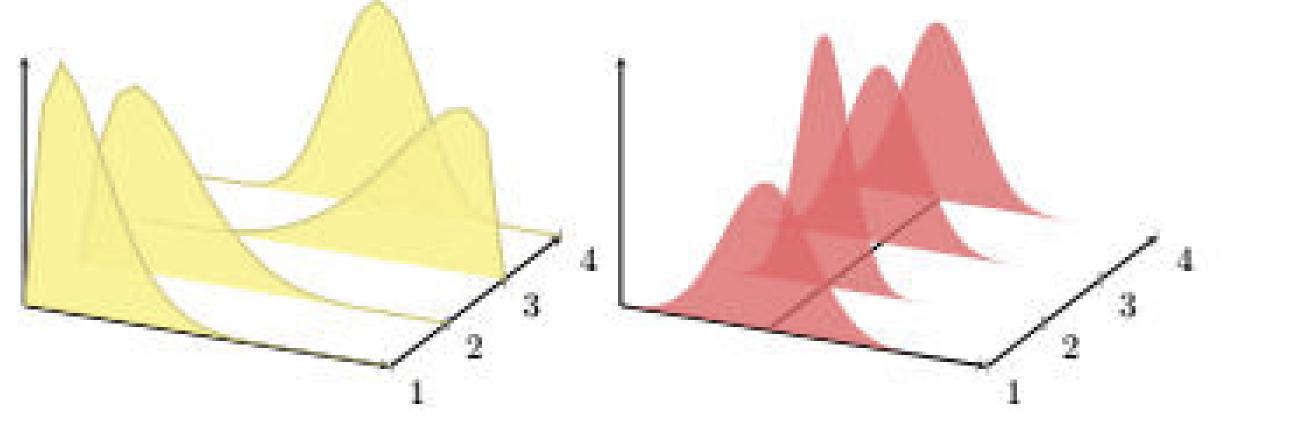
algorithm 2-levels clustering based on k-means and probablistic (deterministic Expectation Maximization [EM])

>> k-means emphasizes latitudinal variation >> EM depicts known biomes more reliably

#### **Scoring Metrics**

To select the most promising models out of the 900+ we made, we developed 2 scoring metrics and combined them into a single score per model:

Confidence Measure \* Accuracy Measure = Score



Aligning and scaling the metrics across clusters, assuming Gaussian distribution.

Filling Example

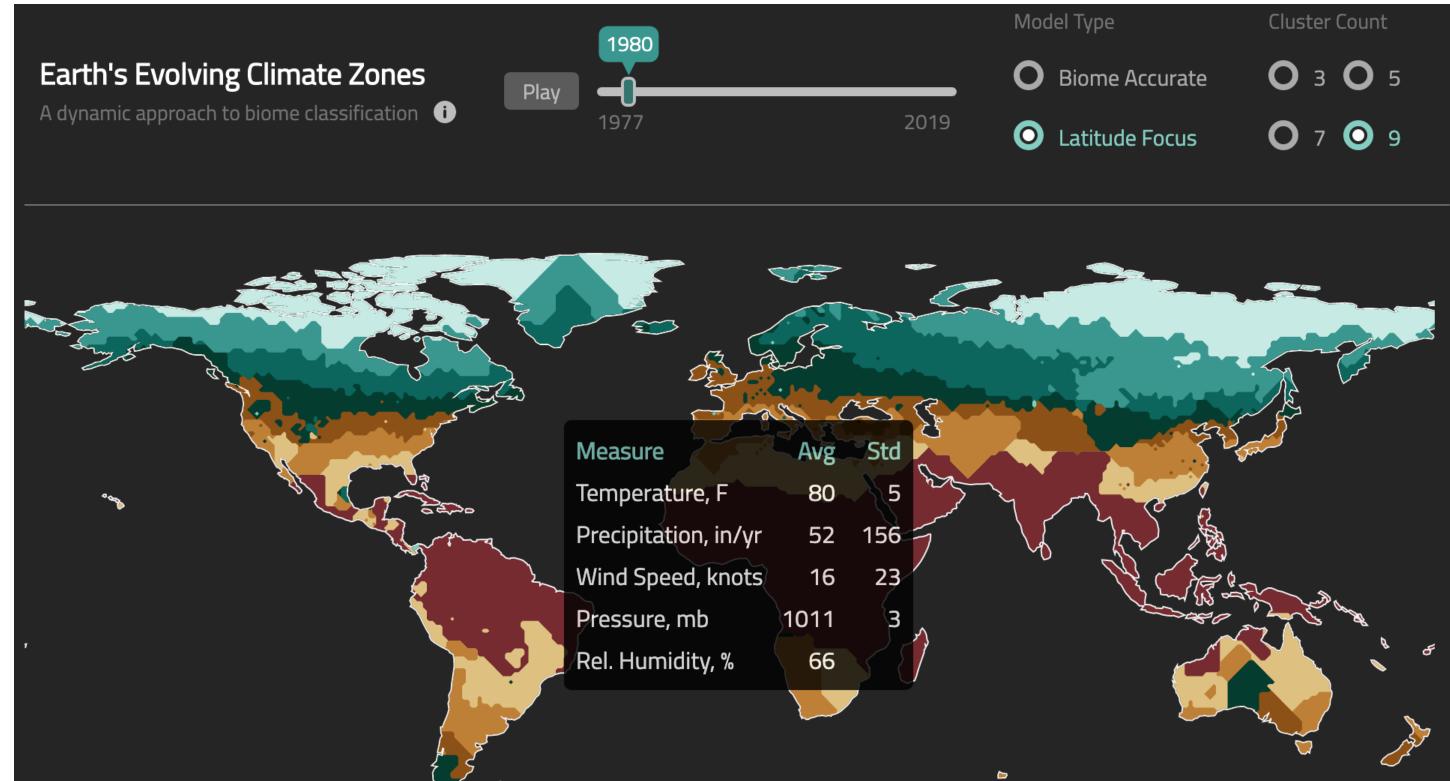
> Used clusters' average latitudes to consistently space labels across our chosen color palette.

> CUSUM-like change detection algorithm to reduce spacial and temporal "flickering"

> Fill empty spaces by "growing" the clusters iteratively.

#### Interactivity

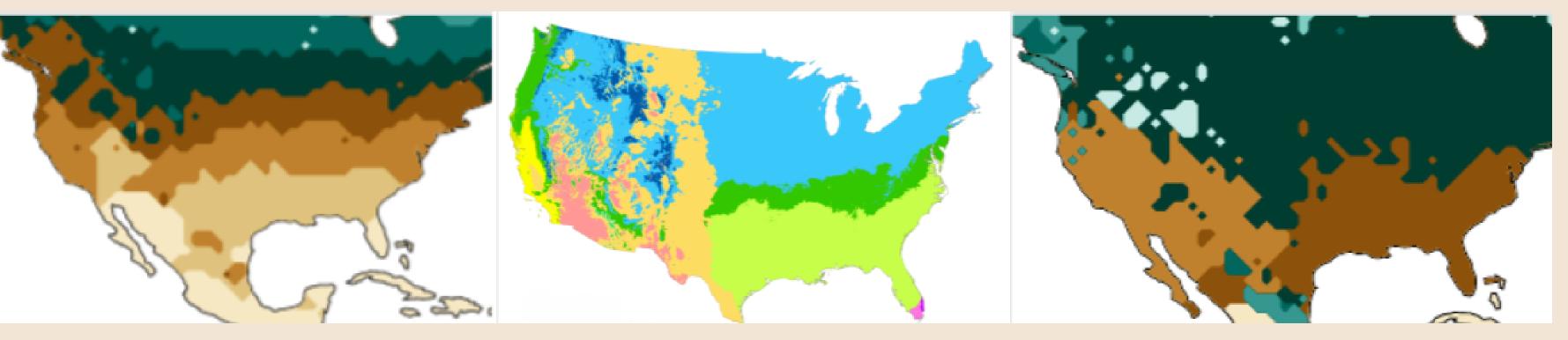
Model-Selection, Map Animation, & Mouse-Events



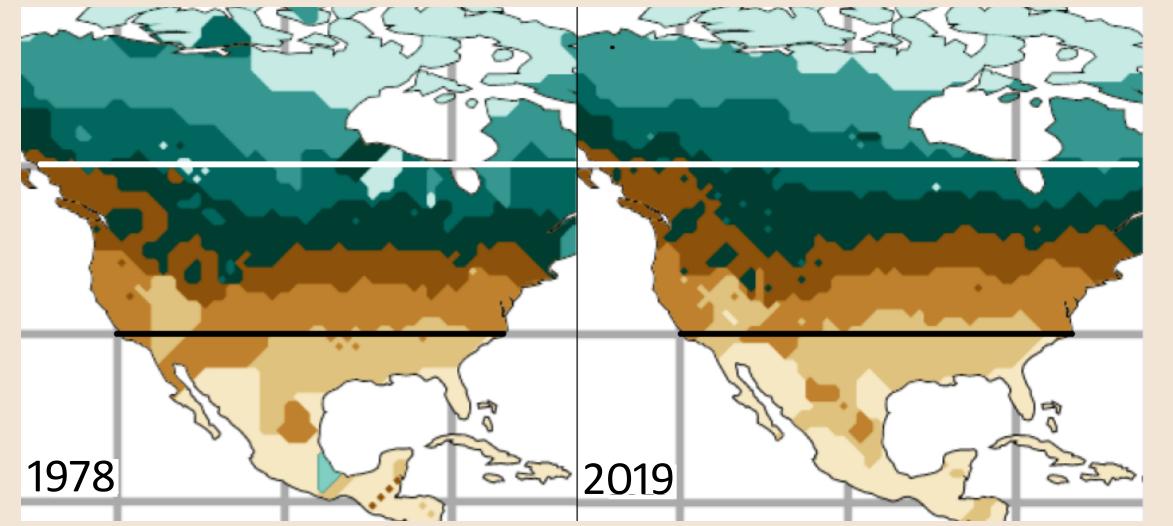
### **Comparing to Reality**

> With a score of 90.31 out of 100, the 9-cluster, k-means, low resolution, 3-parameter model without dimensionality reduction performed the best.

> Below is a comparison of how that model (left) and the similar Expectation-Maximization model (right) compare to the Trewartha climate classification model.



# **Results and Conclusions**



### <u>A Shifting Climate</u>

Evidence of pole-ward shift is seen in our model outputs. Above, the graphic shows these shifts for the United States over the period from 1978 to 2019. Using the black and white lines as guides the shift is easily noticable.

From simple weather station data, indicating precipitation, temperature, dew point, wind speed, and pressure, we have created dynamic models that accurately classify climate biomes and that illustrate how those biomes are shifting towards the poles.

We believe that the ideas and methodology presented here can help inform the public and policy makers by making it easier and quicker to visualize the changing climate.