



Arctic Sea Ice: Exploring global warming

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Outline

- Motivation
- Background
- Approaches
- Preliminary Analysis





Motivation

- UCI Data Science Climate Hackathon
 - Investigate: California's drought, Arctic sea ice, or West Coast wildfires.
- Why Arctic sea ice?
 - Interested in exploring if climate change is predominantly caused by human influences.





Background

The data are provided in NetCDF-4 file

Variable	Description
latitude	Latitude in degrees
longitude	Longitude in degrees (0 to 360)
seaice_conc	Sea ice concentration in percent with values from 0 to 100, inclusive, Land is indicated by -1.
seaice_source	Describes the source of the data for each month of data beginning with January 1850
time	Time of the data observation in days since 1850- 01-01 00:00:00





Contributing data sources

- From 1850 to 1978, analog sources were used:
 - Charts, yearbooks, ice extent grids, whaling ship logs, and etc.
- From 1979 to present:
 - Sea ice concentration from satellite passive microwave data.





Merging data sources

- Ranking
 - Each possible sources for a concentration value was given a rank (higher numbers outrank lower).
 - Satellite data is the exception to the ranking rule.
- Temporal consistency
 - Monthly data points represent ice at the 15th or 16th of each month.
 - Mid month value was selected instead of monthly average.





Approaches

One of the most visible of signs of warming is the retreat of Arctic sea ice.

Develop visualizations to help analyze the spatial and temporal similarity:

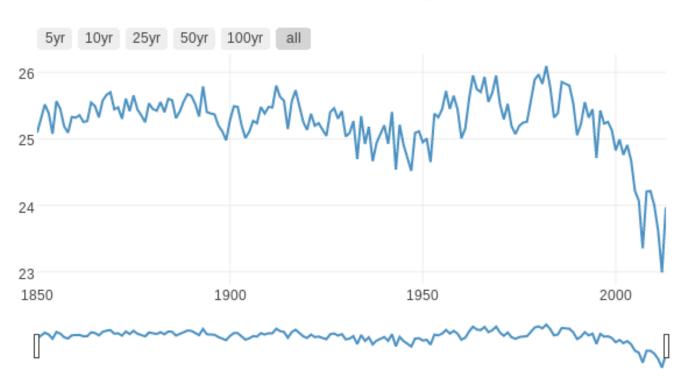
- Yearly averages
- Seasonal averages
- Trend analysis
- Spatio-temporal clustering





Data Visualizations

Yearly Seaice Averages

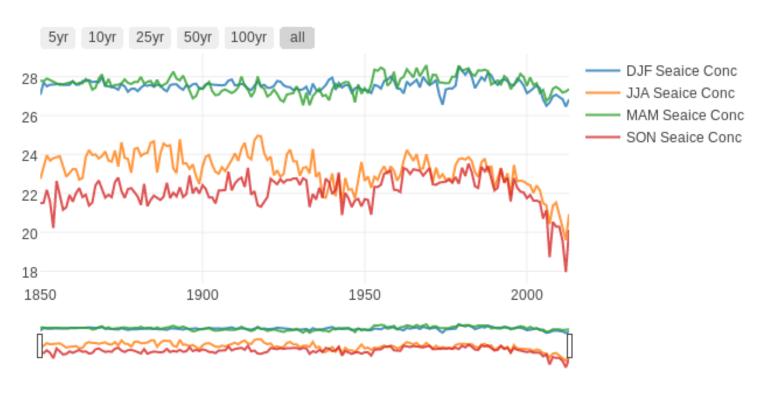




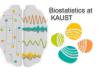


Data Visualizations

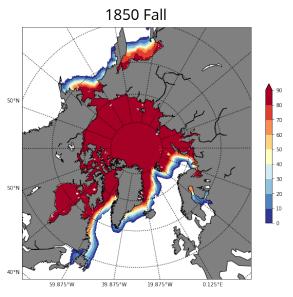
Yearly Seasonal Seaice Averages

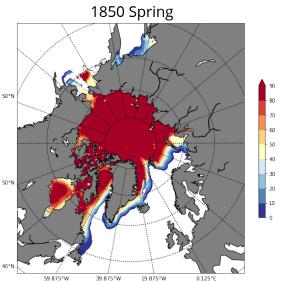


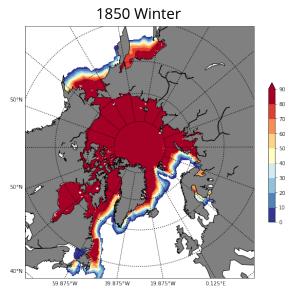


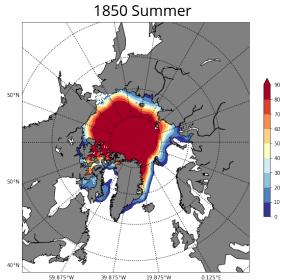


Data Visualizations

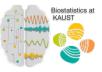




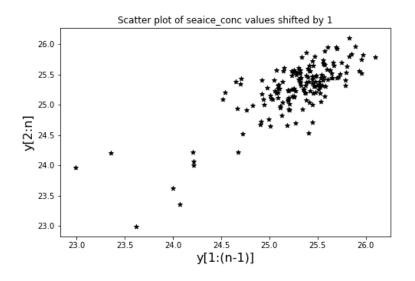


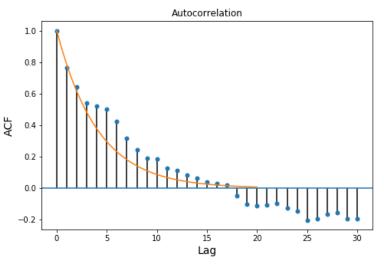






Estimate autocorrelation function from time series data.









Simple trend analysis with linear regression





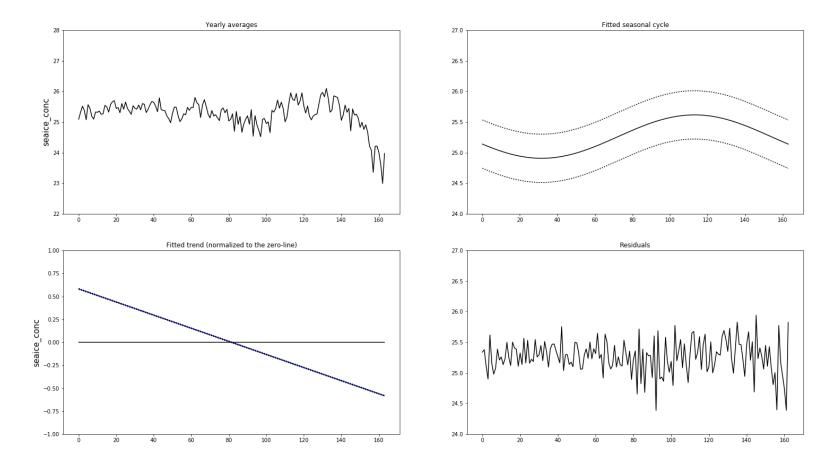
Trend analysis II: Linear regression with AR1 covariance

```
Generalized least squares fit by maximum likelihood
  Model: y ~ t1
  Data: NULL
       AIC
                BIC
                       logLik
  59.81062 72.21009 -25.90531
Correlation Structure: AR(1)
 Formula: ~1
 Parameter estimate(s):
      Phi
0.7662343
Coefficients:
               Value Std.Error t-value p-value
(Intercept) 32.43808 3.655355 8.874128 0.0000
            -0.00373 0.001892 -1.971625 0.0504
t1
 Correlation:
   (Intr)
t1 -1
Standardized residuals:
        Min
                     01
                                Med
                                             03
                                                        Max
-4.40980463 -0.35191840 0.05324723 0.53181699 2.38815117
Residual standard error: 0.4398229
Degrees of freedom: 164 total; 162 residual
```

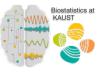




Time series decomposition







Trend analysis II: time series decomposition

```
Generalized least squares fit by maximum likelihood
 Model: v \sim sinf + cosf + t1
  Data: NULL
       AIC
               BIC
                      logLik
  59.08643 77.68563 -23.54322
Correlation Structure: AR(1)
Formula: ~1
Parameter estimate(s):
     Phi
0.723617
Coefficients:
              Value Std.Error t-value p-value
(Intercept) 39.03511 4.553353 8.572829 0.0000
           -0.33283 0.164774 -2.019891 0.0451
sinf
cosf
           -0.12260 0.108425 -1.130770 0.2598
t1
           -0.00714 0.002357 -3.029358 0.0029
Correlation:
     (Intr) sinf cosf
sinf -0.732
cosf -0.001 0.000
    -1.000 0.732 0.000
Standardized residuals:
                            Med
-3.8765989 -0.4946197 0.1250875 0.6931627 2.3477637
Residual standard error: 0.4037854
```

Degrees of freedom: 164 total; 160 residual





Future Work

- Investigate climate analysis and models
- Spatial Analysis
- Interactive map of sea ice concentration
- Spatio-temporal clustering
- Include "data source" into analysis