

Tracking Arctic Sea Ice Concentration: Data Analysis and Visualization

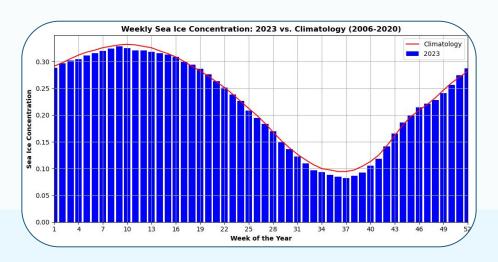
by Tien Ly, Satellite Data Scientist Intern, Summer 2024



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Summer 2024 Internship at NOAA PolarWatch (06/03 - 08/16)

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Overview of Milestones 1 and 2

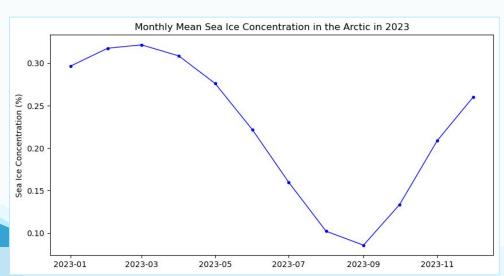
Milestone I: Skills Development

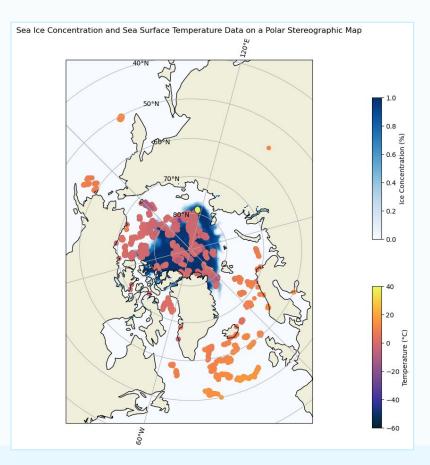
- Understanding satellite products
- Accessing satellite data via ERDDAP
- Using Python for data manipulation
- Git and GitHub for version control and collaboration
- Environment management with conda
- Documentation with Markdown

Milestone 2: Data Analysis and Visualization

- Exploring PolarWatch data catalog and SIC dataset
- Conducting literature review on SIC applications
- Completing Python tutorials and creating Jupyter notebook
- Handling and analyzing data with xarray
- Visualizing temporal and spatial data

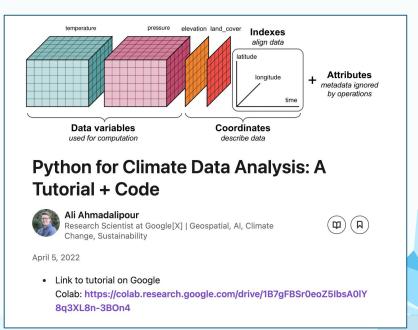
Key Visualizations from Milestone 2





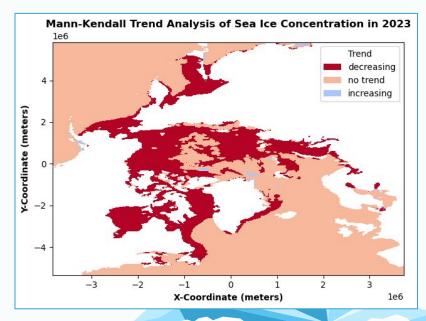
Milestone 3: Weekly Climatology Plots

- Completing tutorials on computing climatology
- Downloading data and computing climatology of sea ice concentration data
- Drafting Jupyter notebook with explanations
- Drafting and presenting data summary

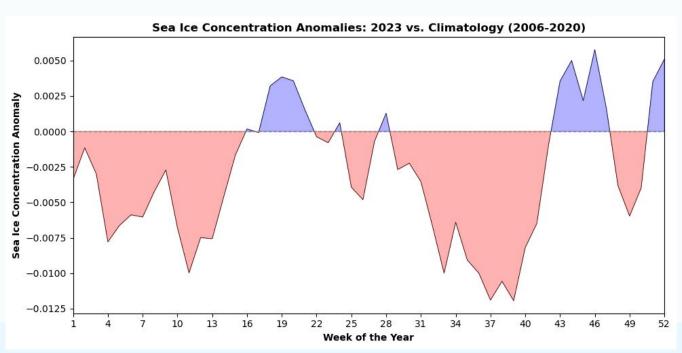


Milestone 3: Weekly Climatology Plots

- Comparing weekly means of sea ice concentration to the current year
- Computing anomalies and trends for 2023
- Identifying weeks with significant differences
- Mapping anomalies and sea ice concentration



2023 Sea Ice Concentration: Identifying Anomalies

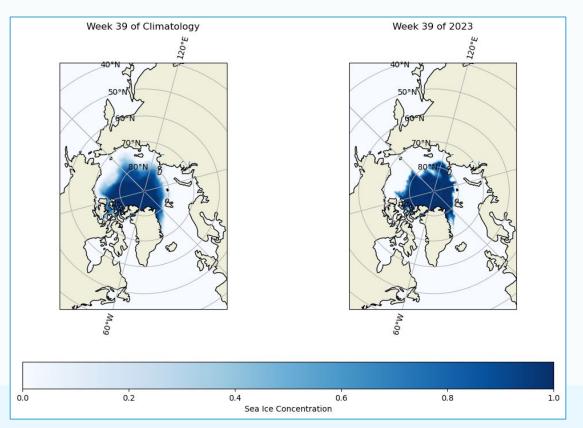


```
Perform two-tailed one-sample t-test and identify significant weeks
                                                                                                                                                                                                         # Initialize lists to store results
    p_values = []
    # cur mean: 2023 weekly means
    # clim mean: climatological weekly means
    # Loop through each week
    for i in range(len(cur_mean)):
      # Calculate the t-statistic
       t stat = (cur mean[i] - clim mean[i]) / (cur std[i] / np.sgrt(7))
       # Calculate the degrees of freedom
       degrees_of_freedom = 7 - 1
       # Calculate the p-value
       p value = (1 - t.cdf(abs(t stat), df=degrees of freedom)) * 2
       # Store the p-value
       p_values.append(p_value)
    # Convert p-values to an array
    p_values = np.array(p_values)
    # Determine significant weeks
    alpha = 0.05
    significant_weeks = np.where(p_values < alpha)[0]
    # Print weeks that show significant differences from the climatology
    if len(significant_weeks) > 0:
       print(f'Weeks showing significant differences from the climatology (alpha = {alpha}):')
       for week in significant_weeks:
           print(f'{week + 1}')
       print('No weeks show significant differences from the climatology.')
Weeks showing significant differences from the climatology (alpha = 0.05):
35
37
```

Identifying weeks with significant differences

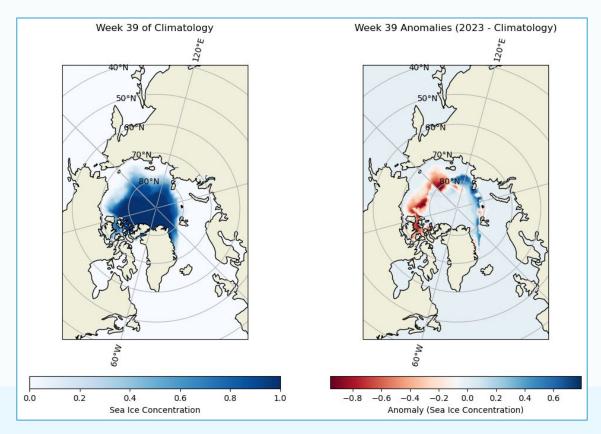
Performing two-tailed one-sample t-test and identify significant weeks

Mapping anomalies





Mapping anomalies



Milestone 4: Modeling and Publishing Sea Ice Data

Milestone 4A (07/29 - 08/09)

- Objective: Model sea ice prediction using deep learning approach
- Activity: Replicate the model introduced in the paper, Sea Ice Forecasting using Attention-based Ensemble LSTM

Milestone 4B (08/12 - 08/16)

 Design and develop a project website in Quarto to showcase all the work completed during the internship

