

# Detecting Trends in Water Temperature in Estuaries of the United States

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# Introduction

- **Goal:** Analyze long term water monitoring data to detect and characterize trends in water temperature for U.S. estuaries
- **Variables of interest** (for trend detection) are water temperature, air temperature, and water temperature with influence of air temperature removed
- Full data set used for initial input, but final analysis broken down by individual estuaries

# Dataset

- 16GB uncompressed, 1.9GB compressed
- historical data with temporal and spatial coverage across the U.S. to study and monitor estuaries
- Features: monitors dedicated to track specific data, such as:
  - water quality, nutrient, meteorological monitors



**National Estuarine Research Reserve System**  
**Centralized Data Management Office**

Image source: <https://cdmo.baruch.sc.edu>

Water Quality Monitors			
Water Temperature	Salinity	Dissolved Oxygen	pH
Turbidity	Specific Conductivity	Chlorophyll-a (optional)	

Nutrient Monitors			
Nitrate (NO <sub>3</sub> <sup>-</sup> )	Nitrate (NO <sub>2</sub> <sup>-</sup> )	Ammonium (NH <sub>4</sub> <sup>+</sup> )	Orthophosphate (PO <sub>4</sub> <sup>3-</sup> )
Total nitrogen (TN)	Total phosphorus (TP)	Silicate (SiO <sub>2</sub> )	

Meteorological Monitors			
Air temperature	Relative humidity	Barometric pressure	Wind speed
Wind direction	Precipitation	Solar radiation	

# Data Preprocessing

- Hadoop MapReduce with Java
- Down-sampled air temperature and water temperature data to get monthly average values
- **Mapper:** Took in full dataset, filtered relevant rows and prepared a key-value pair based on month
- **Reducer:** Aggregated our readings for each key and calculate the average

### **Mapper Input**

StationCode isSWMP dateTimeStamp waterTemperature

acebbwq P 3/3/1995 18:00 14.2

### **Mapper Output (Key-Value Pair)**

Key: acebbwq-199503, Value: 14

---

### **Reducer Input**

Key: acebbwq-199503

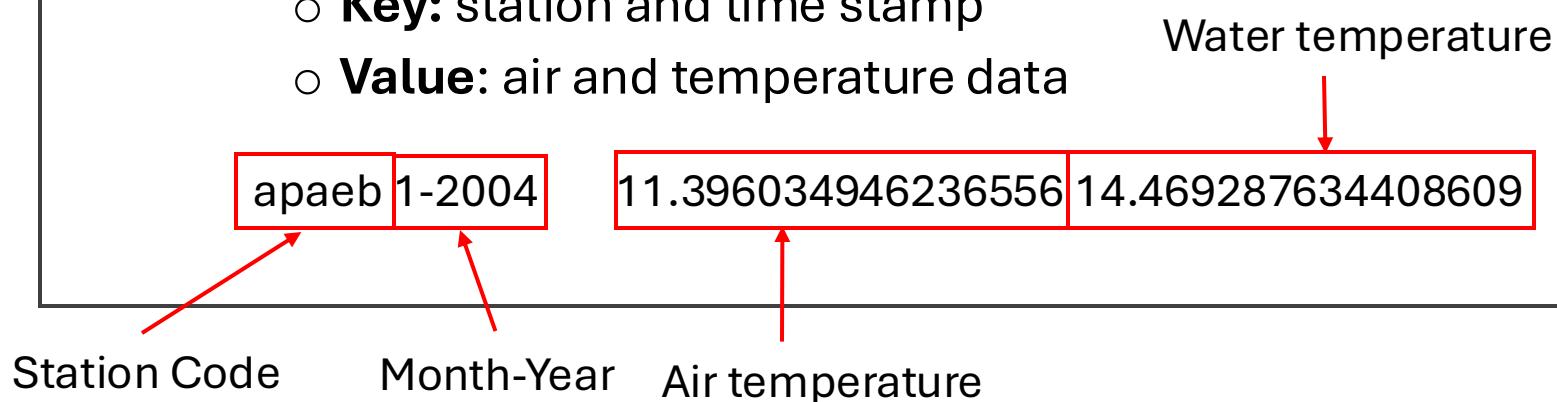
Values: [14, 14, 15]

### **Reducer Output**

Key: acebbwq-199503, Value: 14

# Composite Join

- Hadoop MapReduce with Java
- Joined two separate datasets for **air** and **water** temperature
- Join on **two foreign keys**: station and time stamp
- Map-only job
  - **Key:** station and time stamp
  - **Value:** air and temperature data



# Mapper

```
public static class CompositeMapper extends MapReduceBase implements Mapper<Text, TupleWritable, Text, Text> {  
    private Text outputKey = new Text();  
    private Text outputValue = new Text();  
  
    public void map(Text key, TupleWritable value, OutputCollector<Text, Text> output, Reporter reporter)  
        throws IOException {  
  
        // stationCode and dateTimeStamp  
        outputKey.set(key.toString().trim() + "\t");  
  
        // Obtain air and water data, removing extra spaces  
        String airInfo = ((Text) value.get(0)).toString().trim();  
        String waterInfo = ((Text) value.get(1)).toString().trim();  
  
        outputValue.set(airInfo + " " + waterInfo);  
  
        output.collect(outputKey, outputValue);  
    }  
}
```

## Driver snippet

```
// input  
format      conf.setInputFormat(CompositeInputFormat.  
class);  
  
// specify join type (e.g., inner join)  
// the input format for the data sources  
(KeyValueTextInputFormat)  
// corresponding paths (airPath, waterPath).  
conf.set("mapreduce.join.expr",  
        CompositeInputFormat.compose(joinType,  
        KeyValueTextInputFormat.class, airPath,  
waterPath));
```

# Linear Regression

- Computed linear regression residuals
  - Represent the water temperature minus the impacts of the air temperature
- Apache Spark using the ML library

```
LinearRegression regression = new LinearRegression();
```

```
LinearRegressionModel model = regression.fit(vectorized);
```

```
Dataset<Row> residuals = model.summary().residuals();
```



# Seasonal Mann-Kendall Test

- Calculated for all three variables of interest for each estuary
  - Identifies and characterizes long-term monotonic trends if any
  - Water, air, and isolated water temperatures
- Compares historical data from similar seasons
  - e.g., winter to winter, summer to summer. In our case, we compared values from the same months but from different years.



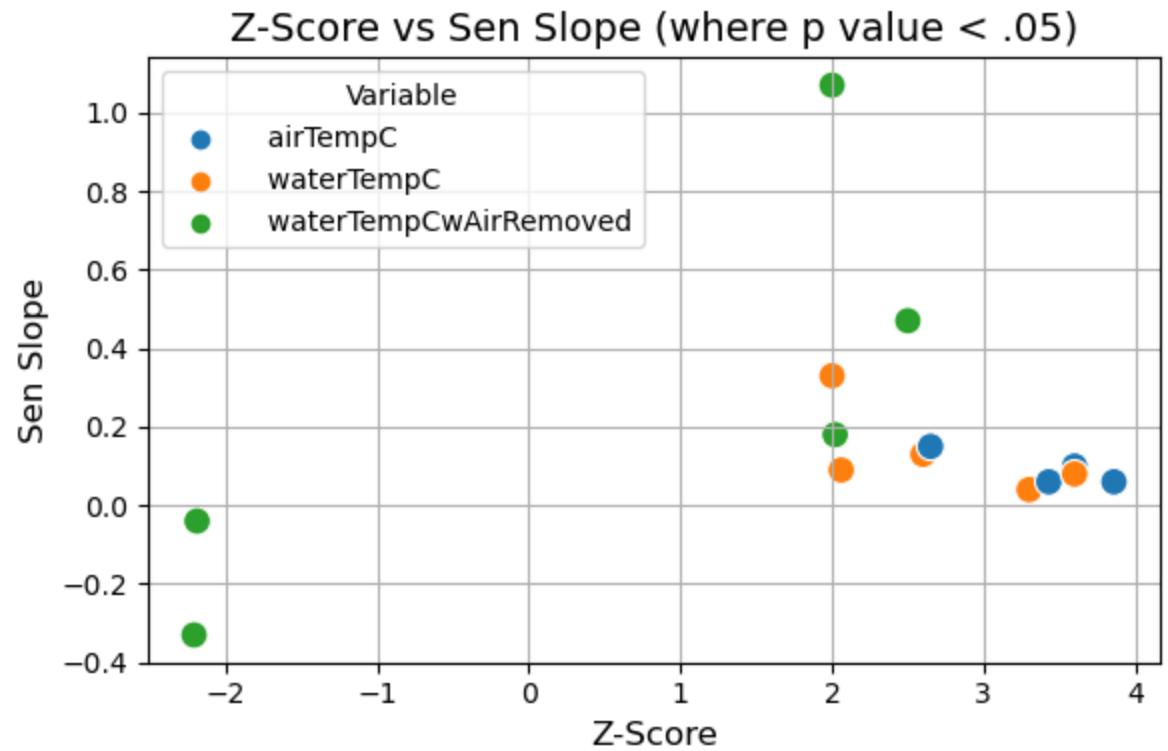
# Seasonal Mann-Kendall Test

- Mann-Kendall Tests
  - Compare data points to all previous datapoints and assigns a -1,+1 or 0. Sums all these to get an S score.
  - Tau normalizes this value between -1 and 1.
  - Z score calculated using the S score and the variance of the data.
  - If a significant trend is detected, a Sen slope estimate can give an estimate of the slope of the trend. It takes the median of all the slopes between all the points.
- Seasonal Mann-Kendall Test
  - Compare only values from like seasons.
  - S score is calculated for each season and then summed.
  - Variance calculated with specific formula and for each season, and then summed together.



# Analysis

- SMK test was used to determine stations with statistically significant results
- 85% of those stations had an increase in water temperatures
- Some stations required more data to be collected before the test could be considered significant



# Conclusion

- Data collected from across the United States
- Distilled and aggregated the water and air temperatures
- Performed the SMK Test, removing statistically insignificant results
- For estuaries with stronger trends detected, the water and air temperature trends were dominantly positive.

# Next Steps & Future Directions

- Determine better methods for determining the trends
- Utilize stations near each other to supplement missing data
  - Air temperature missing from large portions of the dataset
- Tweak hyperparameters for linear regression
  - Improve the accuracy of the residuals
- Remove less data during aggregation step, focusing on more granular timeframes

# Questions?

