

Thursday, November 24, 2020 | **Class #15**

**SciPy (2-D interpolation),  
pandas (correlating, resampling, and  
smoothing data)**

OCEAN 215 | Autumn 2020

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# Calculating correlations using `.corr()`

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Recall that a column of a Pandas DataFrame is a Pandas Series.

We can correlate two Pandas Series objects, `s1` and `s2`, using the following syntax:

```
s1.corr(s2)
```

The result is the standard Pearson correlation coefficient,  $r$ .

More commonly, however, you'll use  $r^2$ . As I introduced in Lesson #14,  $r^2$  represents the proportion of variance in one variable that is explained by the other variable.

API documentation: <https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.DataFrame.corr.html>

# Changing the time resolution using `.resample()`

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We can down-sample the time resolution of a Pandas Series or DataFrame if it has datetimes as its index, i.e. a `DatetimeIndex`.

The function `.resample()` takes a "frequency alias" or "offset alias" as its argument. It behaves similar to `.groupby()`, so after you group the data, you have to apply a function like `.mean()`, `.max()`, or `.sum()`.

Here are the available frequency aliases: [https://pandas.pydata.org/docs/user\\_guide/timeseries.html#offset-aliases](https://pandas.pydata.org/docs/user_guide/timeseries.html#offset-aliases)

Some common options are:

- 'H': hourly frequency
- 'D': daily frequency
- 'W': weekly frequency
- 'MS': monthly frequency (use start of month as resulting date)
- 'YS': yearly frequency (use start of year as resulting date)

API documentation: <https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.DataFrame.resample.html>

# Smoothing data using `.rolling()` averages

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A common technique to smooth a time series (or other data) is to calculate a rolling average, also called a running average, running mean, or moving average:

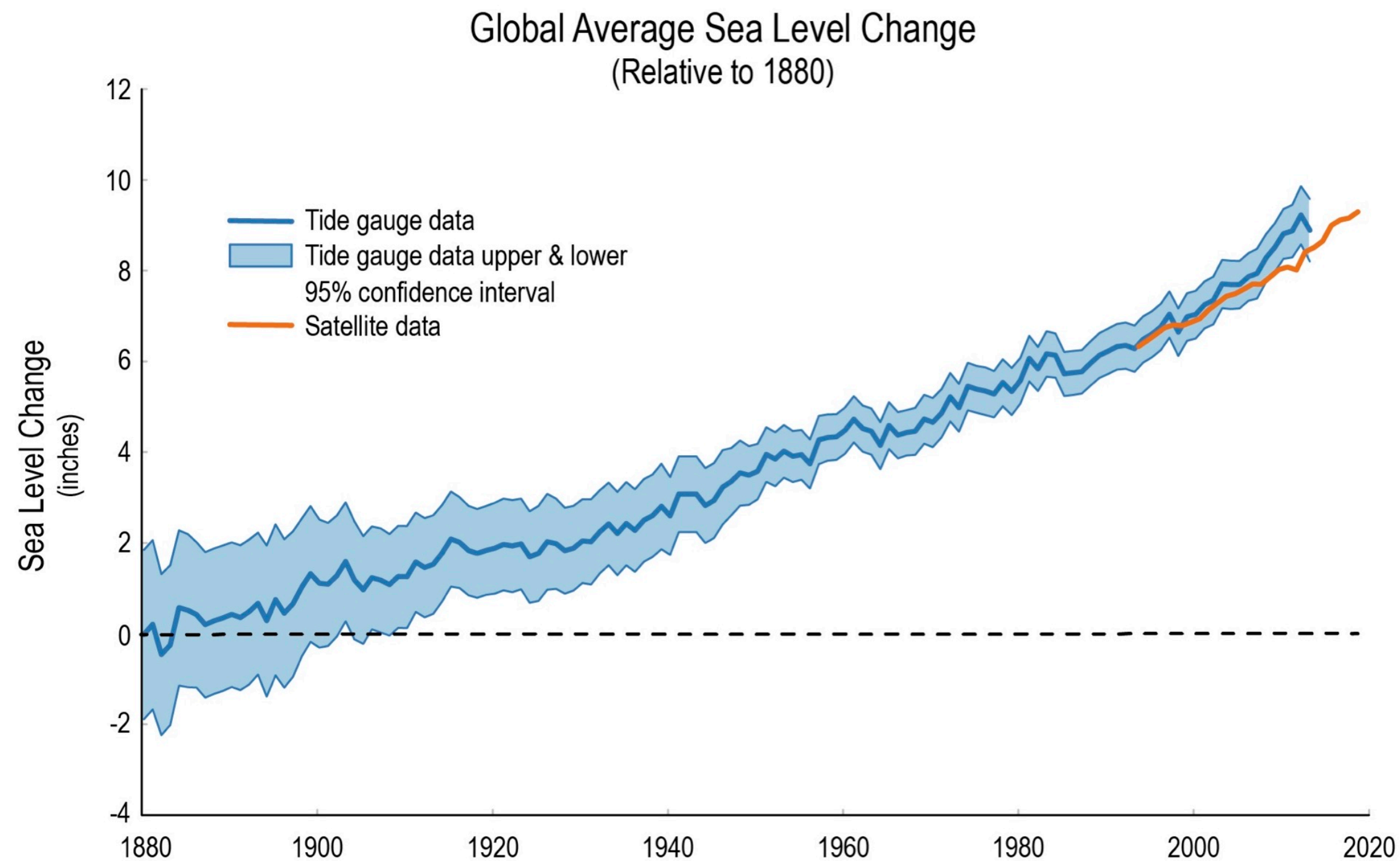
```
.rolling(window,min_periods=1,center=False)
```

- `window` specifies the size of the moving window in number of rows
- `min_periods` specifies the minimum number of rows required to have data (otherwise the result is `np.NaN`); this is important at the start and end of the time series
- if `center` is `True` (recommended), the date will be set to center of each window; if `False`, the date will be set to the end of each window

Again, `.rolling()` behaves similar to `.groupby()` and `.resample()` in that you have to apply a function like `.mean()` or `.median()` to the grouped data to get a result.

API documentation: <https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.DataFrame.rolling.html>

# Activity: detecting sea level rise from Florida tide gauge records



**Source:** U.S. Global Change Research Program  
(USGCRP)



**Image:** Miami Beach faces more frequent sunny-day flooding during king tides due to sea level rise. (South Florida Sun Sentinel)

Google Doc with activities (also accessible from Canvas Modules or Google Drive folder):

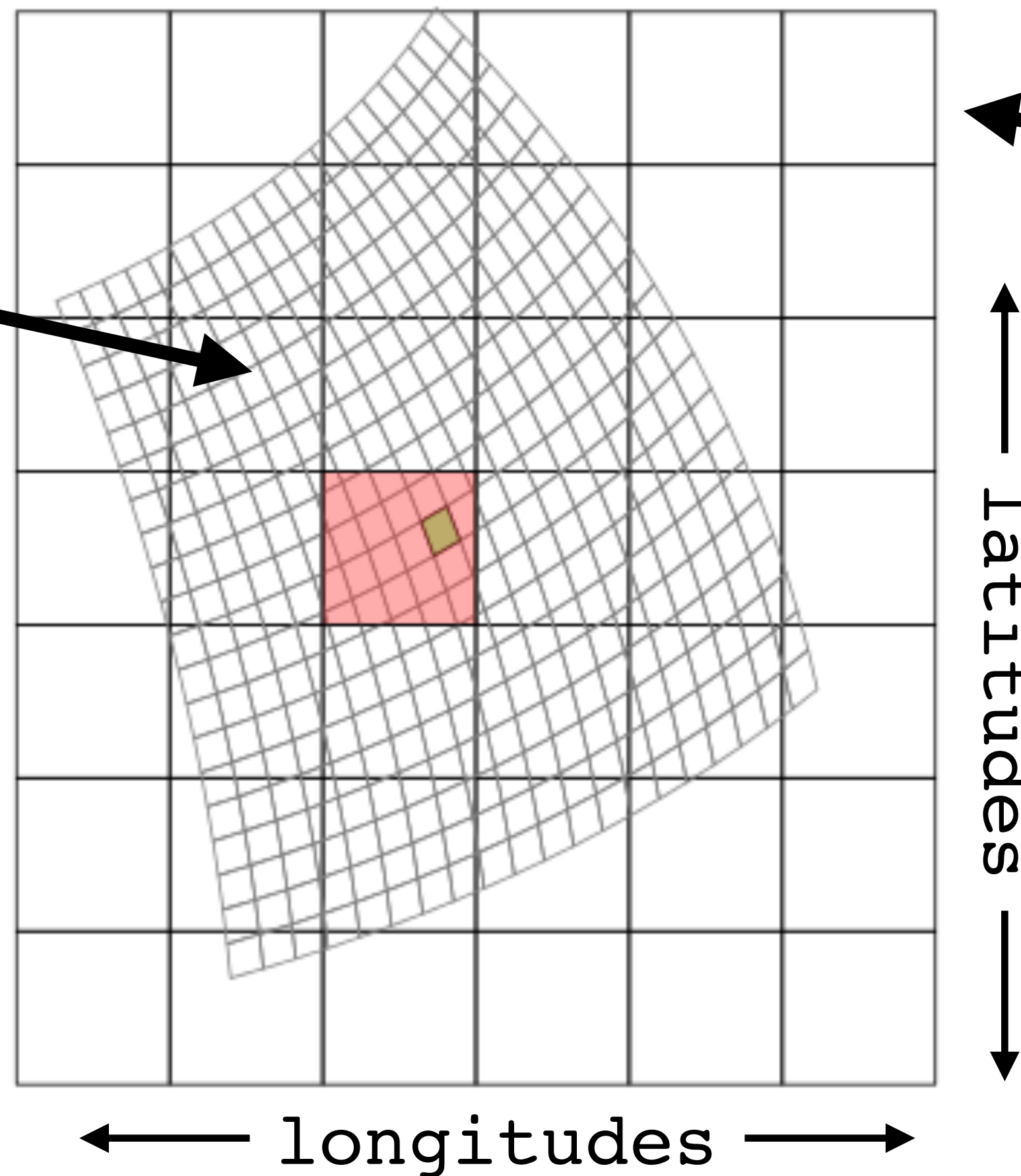
**<https://tinyurl.com/OCEAN215-Class15>**

# 2-D interpolation (a.k.a. 2-D regridding)

## You have:

An irregular grid  
(`lat` and `lon`  
or other coordinates are usually  
2-D arrays)

~~`plt.pcolormesh()`  
`plt.contourf()`  
`xarray's .sel()`~~



## You want:

A regular grid  
(`lat` and `lon`  
can be represented  
as 1-D coordinates)

`plt.pcolormesh()`  
`plt.contourf()`  
`xarray's .sel()`

For more information on regridding, see [Climate Data Guide's "Regridding Overview"](#)

Image credit: [Lu et al. \(2018\)](#)

# 2-D interpolation in SciPy is a three-step process

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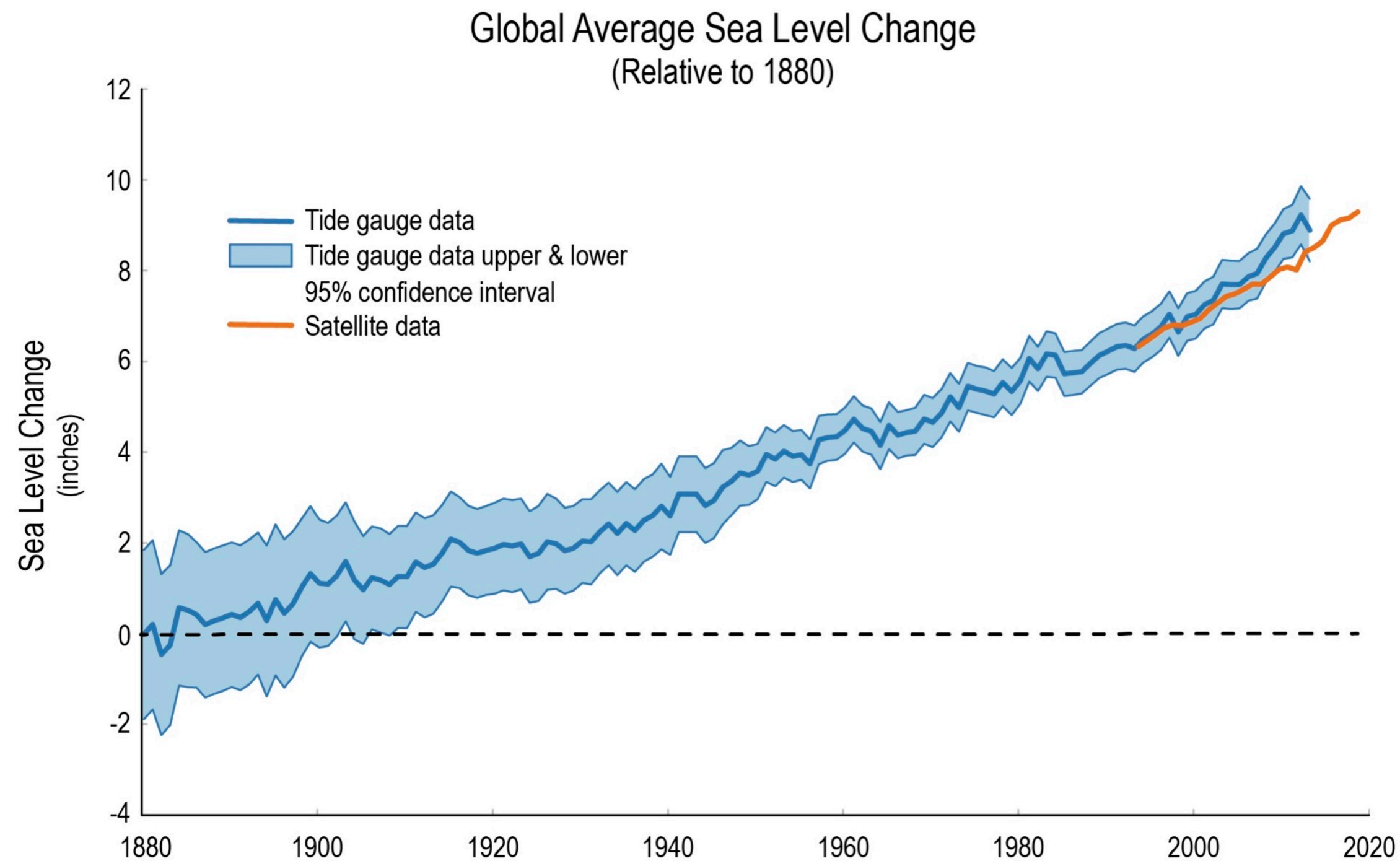
```
x_coord = np.linspace(start, end, num_x_points)
y_coord = np.linspace(start, end, num_y_points)
```

```
x_grid, y_grid = np.meshgrid(x_coord, y_coord)
```

```
z_gridded = interpolate.griddata((x_flat, y_flat),
                                z_flat,
                                (x_grid, y_grid),
                                method='linear')
```

API references: [NumPy meshgrid\(\)](#) and [SciPy griddata\(\)](#)

# Activity: interpolating scattered global tide gauge measurements



*Source: U.S. Global Change Research Program (USGCRP)*



*Image: Miami Beach faces more frequent sunny-day flooding during king tides due to sea level rise. (South Florida Sun Sentinel)*

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