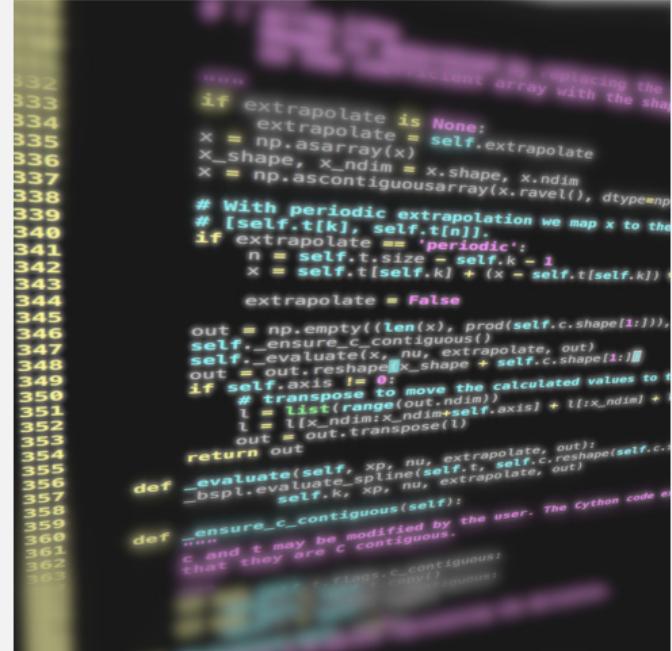


Programming Numerical Data Analysis & Visualization

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```
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```

if extrapolate is None:
 extrapolate = self.extrapolate
x = np.asarray(x)
x_shape, x_ndim = x.shape, x.ndim
x = np.ascontiguousarray(x.ravel(), dtype=np.float64)
With periodic extrapolation we map x to the
[self.t[k], self.t[n]].
if extrapolate == 'periodic':
 n = self.t.size - self.k - 1
 x = self.t[self.k] + (x - self.t[self.k]) % n
 extrapolate = False
out = np.empty((len(x), prod(self.c.shape[1:])))
self._ensure_c_contiguous()
self._evaluate(x, nu, extrapolate, out)
out = out.reshape(x_shape + self.c.shape[1:])
if self.axis != 0:
 # transpose to move the calculated values to the
 # right position.
 l = list(range(out.ndim))
 l[-x_ndim:x_ndim+1] = l[-x_ndim+1:-1]
 out = out.transpose(l)
return out
def _evaluate(self, xp, nu, extrapolate, out):
 bsp1.evaluate_spline(self.t, self.c.reshape(self.c
 .shape[0]),
 self.k, xp, nu, extrapolate, out)
def _ensure_c_contiguous(self):
 """
 c and t may be modified by the user. The Cython code
 expects c and t to be contiguous.
 That's why we check if they are c-contiguous.
 If not, we copy them.
 """

Recap

Reading from file

Dynamic: read from file with name requested by prompt

```
1 fName = input('Input file: ')
2 lines = list()
3 f = open(fName)
4 for line in f:
5     lines.append(line)
```

File formats

Unstructured data

- Text

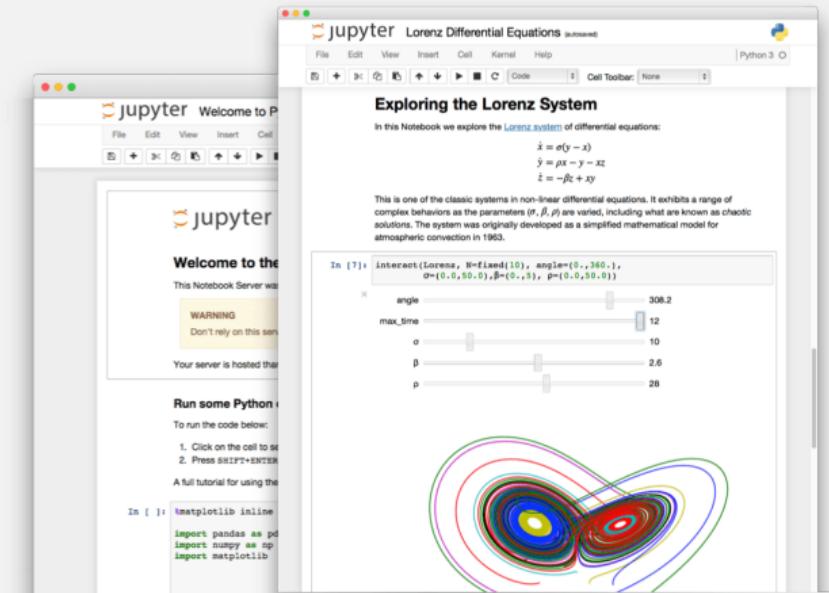
Structured data

- XML
- JSON
- Tables
- Matrices

Jupyter Notebook

Why use Jupyter Notebook in Data Science?

- Simultaneous documentation & analysis
- Step-by-step processing
- Ensures reproducibility

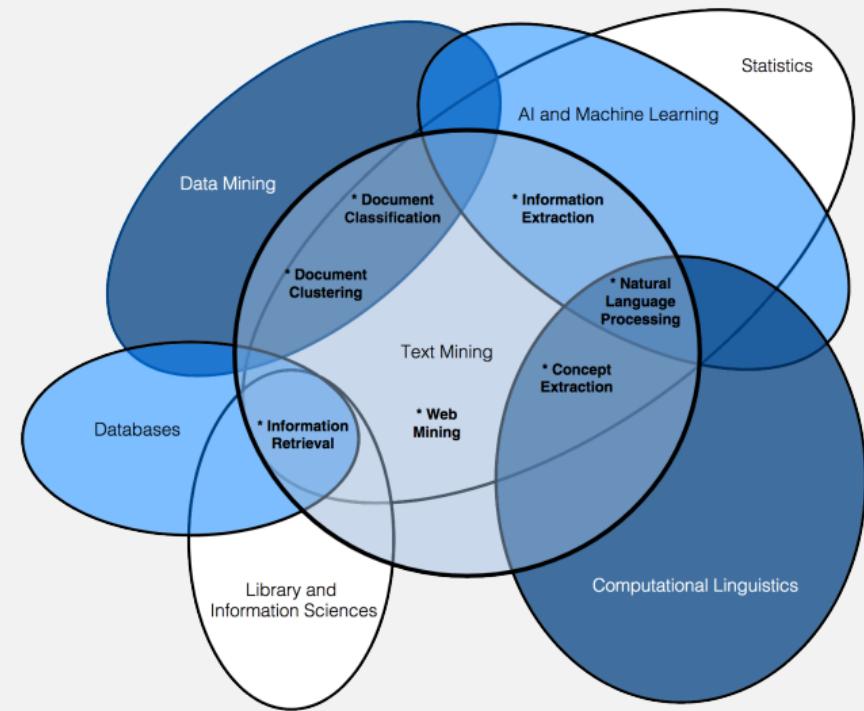


Text mining

Relies on *Natural Language Processing* (NLP)

Main (constitutive) tasks:

- Document summarization, clustering & classification
- Information extraction
- Information discovery



source: Miner, Gary. Practical Text Mining and Statistical Analysis for Non-structured Text Data Applications. 1st ed. Amsterdam: Academic Press, 2012.

**Data
Visualization**

**Numerical
Data Analysis
with NumPy**

**Modeling
Experimental
Data**

Matplotlib: Visualization with Python

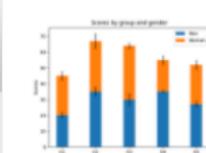
- de-facto standard library for scientific visualizations
- many third party packages built on top of Matplotlib
- comprehensive library for creating static, animated, and interactive visualizations

The screenshot shows the official Matplotlib website at matplotlib.org/index.html. The page features a large logo with the word "matplotlib" and a circular icon. Below the logo, a navigation bar includes links for Installation, Documentation, Examples, Tutorials, and Contributing. The main content area is titled "Matplotlib: Visualization with Python" and describes the library as "a comprehensive library for creating static, animated, and interactive visualizations in Python". It highlights that Matplotlib makes "easy things easy and hard things possible" and provides examples of various plots like a line graph, a histogram, a heatmap, and a 3D surface plot. Three sections below the main text—Create, Customize, and Extend—each list several bullet points about the library's capabilities. A sidebar on the right contains information about the latest release (3.2.1), the last release for Python 2 (2.2.5), and development versions. At the bottom, there's a "Documentation" section and a "Support Matplotlib" button.

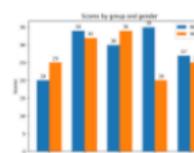
source: <https://matplotlib.org/>

Matplotlib: Visualization with Python

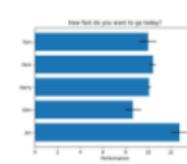
Lines, bars and markers



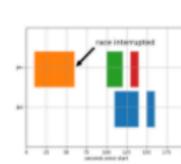
Stacked bar chart



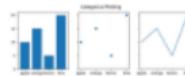
Grouped bar chart
with labels



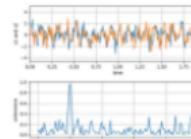
Horizontal bar chart



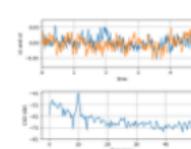
Broken Barh



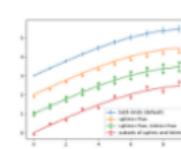
Plotting categorical
variables



Plotting the
coherence of two
signals



CSD Demo

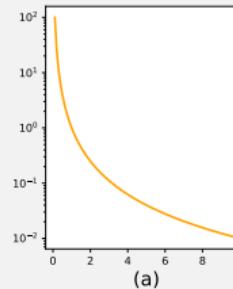


Errorbar limit
selection

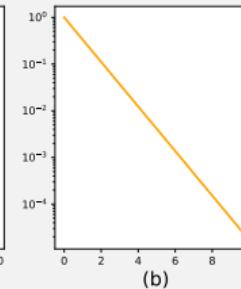
source: <https://matplotlib.org/gallery/index.html>

Quiz

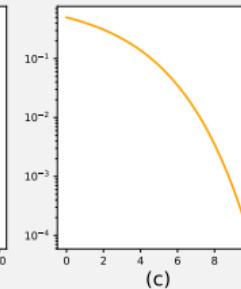
- Assign the decay functions to their shapes when plotted on a logarithmic scale:



(a)



(b)



(c)

exponential

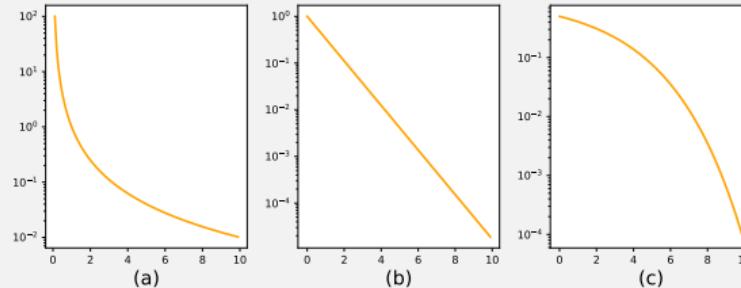
polynomial

super-exponential

- Which of these matplotlib (plt) functions can be used to draw a scatter plot?
`plt.dot()` `plt.plot()` `plt.draw()` `plt.show()`
- Specify the matplotlib (plt) command to draw a dotted blue line defined by lists X and Y.

Quiz

- Assign the decay functions to their shapes when plotted on a logarithmic scale:



exponential (b)

polynomial (a)

super-exponential (c)

- Which of these matplotlib (plt) functions can be used to draw a scatter plot?
`plt.dot()` `plt.plot()` ✓ `plt.draw()` `plt.show()`
- Specify the matplotlib (plt) command to draw a dotted blue line defined by lists X and Y.

```
plt.plot(X, Y, ':', color='blue')
```

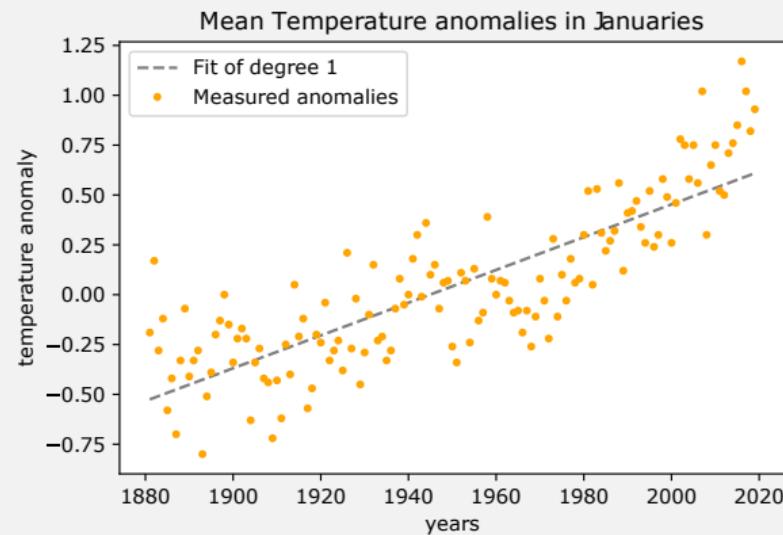
**Data
Visualization**

**Numerical
Data Analysis
with NumPy**

**Modeling
Experimental
Data**

Numerical analysis

- Numerical data: anything measurable
- Methods:
 - Interpolation and regression
 - Solving differential equations
 - Optimization
 - ...



N-dimensional array: numpy.ndarray

Array data structure

- immutable
- n-dimensional
- very storage efficient
- can store only data of same type

Quiz

- Given the following variables: `a = np.array((1, 2, 3))`,
`b = np.array(((4, 5, 6), (7, 8, 9)))`, `c = 10`, specify the following statements use *vectorized operation or broadcasting*:

- `a * b`
- `a ** a`
- `b * c`
- `a / b[1]`

- Determine the data type of the following NumPy arrays:

- `np.array((1,))`
- `np.array((1.0,))`
- `np.array((1, 1.0, (1, 1.0)))`
- `np.array(((1, 1.0), (1, 1.0)))`

Quiz

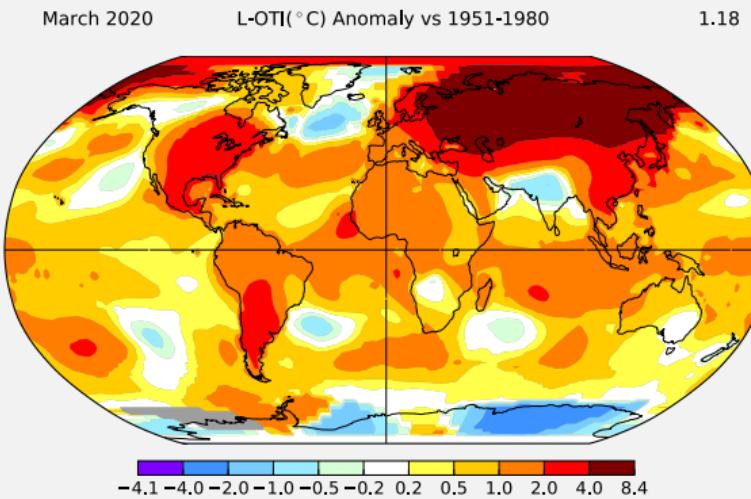
- Given the following variables: `a = np.array((1, 2, 3))`,
`b = np.array(((4, 5, 6), (7, 8, 9)))`, `c = 10`, specify the following statements use *vectorized operation* or *broadcasting*:
 - `a * b` broadcasting
 - `a ** a` vectorized op.
 - `b * c` broadcasting
 - `a / b[1]` vectorized op.
- Determine the data type of the following NumPy arrays:
 - `np.array((1,))` int64
 - `np.array((1.0,))` float64
 - `np.array((1, 1.0, (1, 1.0)))` object
 - `np.array(((1, 1.0), (1, 1.0)))` float64

**Data
Visualization**

**Numerical
Data Analysis
with NumPy**

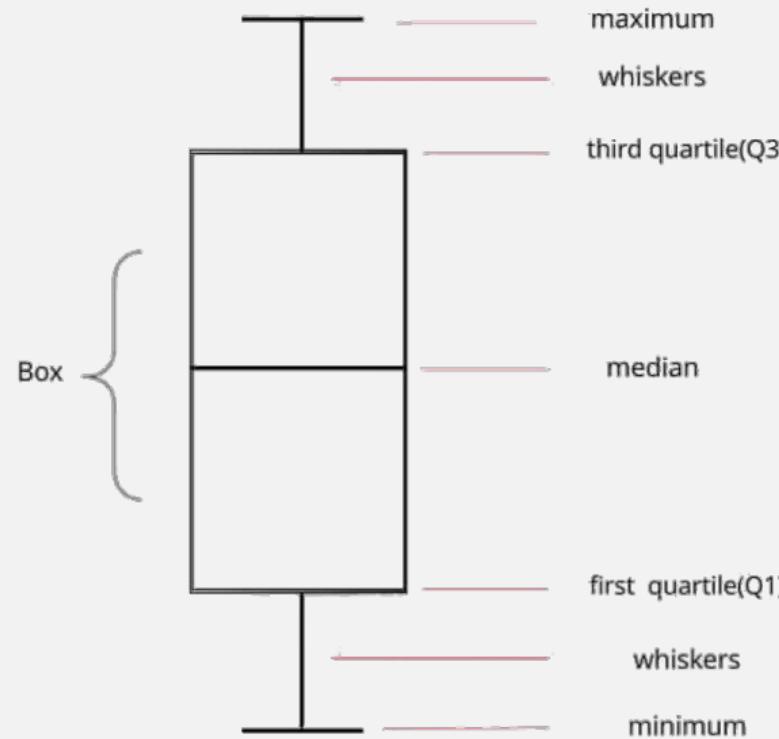
**Modeling
Experimental
Data**

NASA's GISS Surface Temperature Analysis



- ▶ <https://data.giss.nasa.gov/gistemp>
- ▶ Collection of temperature data from thousands of meteorological stations
- ▶ Data represents *anomalies*, i.e., deviations from mean temperature measured in 1951-1980

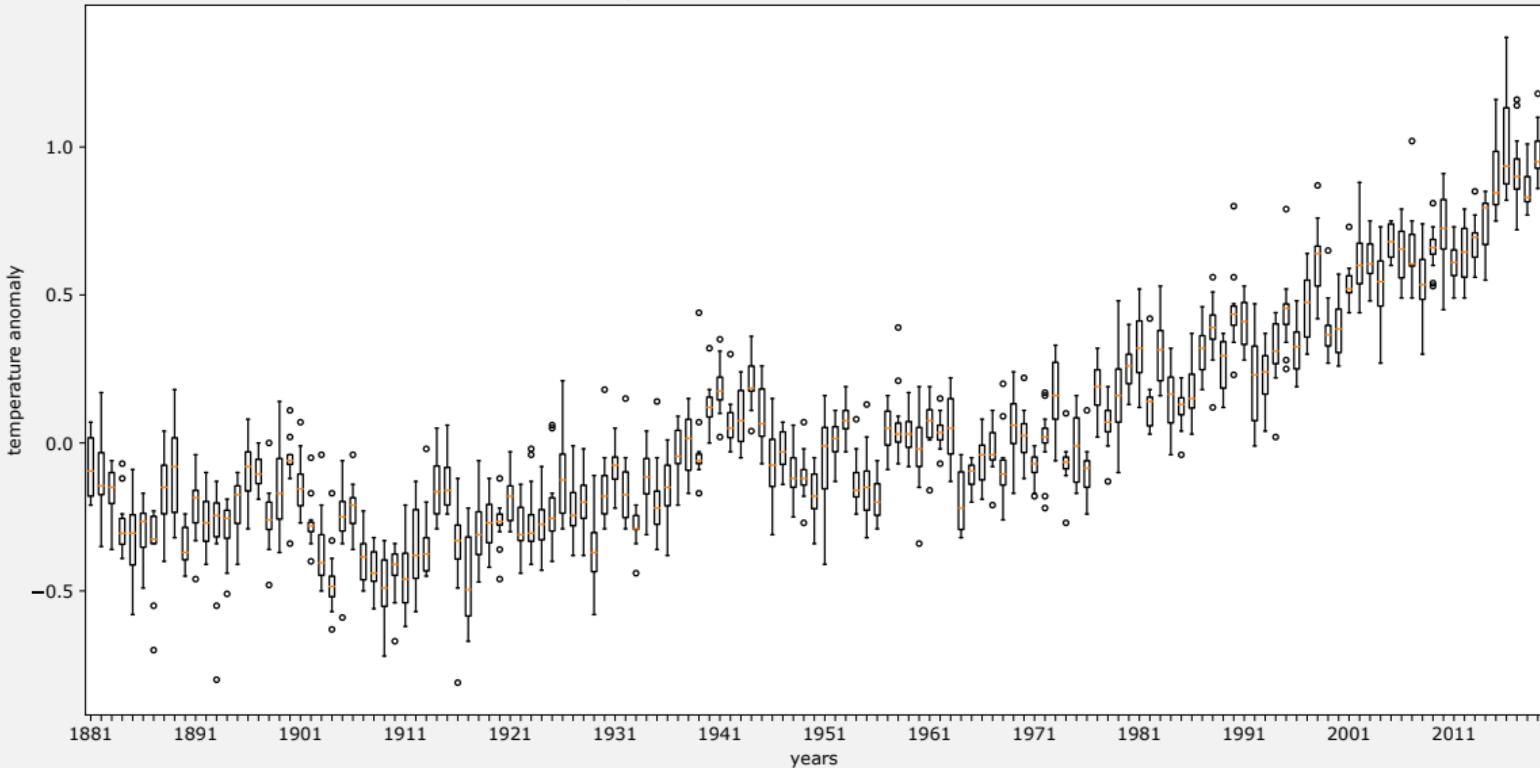
Whisker plot



source: <https://aiaspirant.com/box-plot/>

Whisker plot of GISS data

Temperature anomalies between 1881-2019



Further classes on methods in Data Science

Classes at Bielefeld University:

- 311850, Multivariate Methods / Regression Analysis (winter semester)
- 392140, Introduction to Machine Learning (winter semester)
- 392219 Applied Optimization (winter semester)

MIT Course 6.0002, Lectures on Understanding Experimental Data:

- <https://www.youtube.com/v/vIFKGF11Cn8>
- <https://www.youtube.com/v/fQvg-hh9dUw>

Linear regression

Estimate a *linear number* of parameters from given input, e.g.:

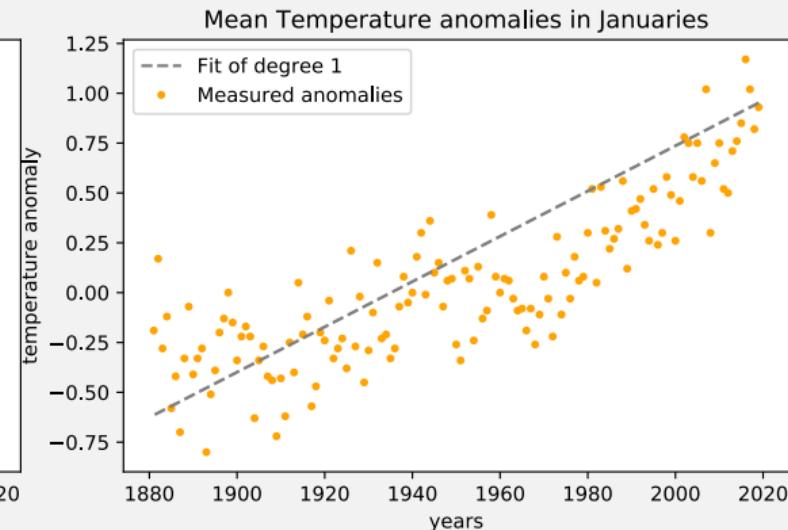
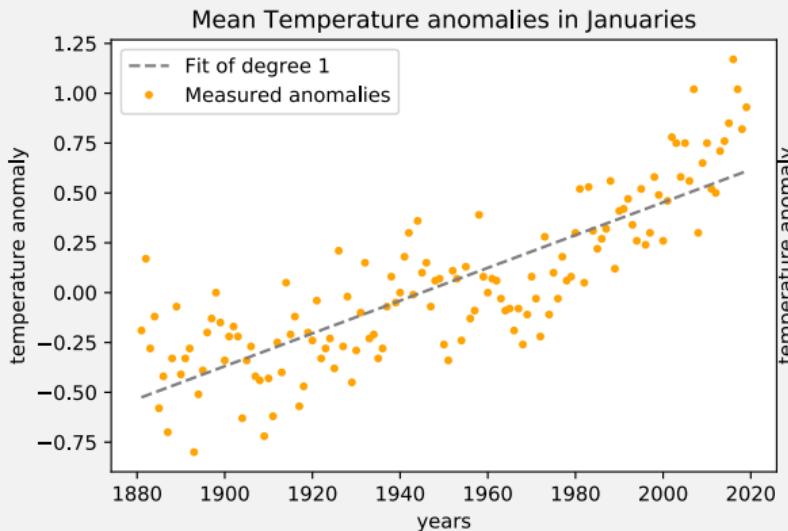
$$X = \begin{pmatrix} X_0 \\ X_1 \\ \vdots \\ X_{N-1} \end{pmatrix}, Y = \begin{pmatrix} Y_0 \\ Y_1 \\ \vdots \\ Y_{N-1} \end{pmatrix} \rightarrow \alpha = \begin{pmatrix} \alpha_0 \\ \alpha_1 \\ \vdots \\ \alpha_{N-1} \end{pmatrix}$$

Estimator:

$$\hat{Y} = \alpha_0 + \alpha_1 X + \alpha_2 X^2 + \cdots + \alpha_{N-1} X^{N-1}$$

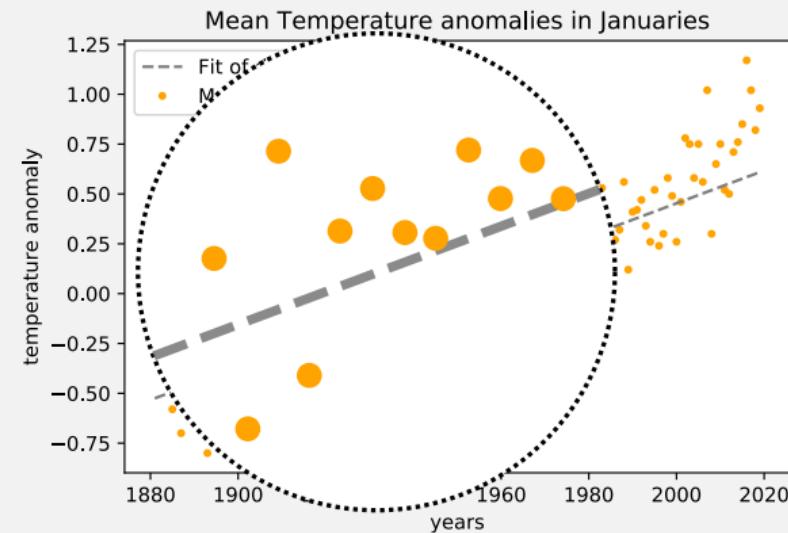
Simple linear regression: Estimate line, i.e., estimate α_0, α_1 and set $\alpha_2 = \cdots = \alpha_{N-1} = 0$

What criterion to optimize?



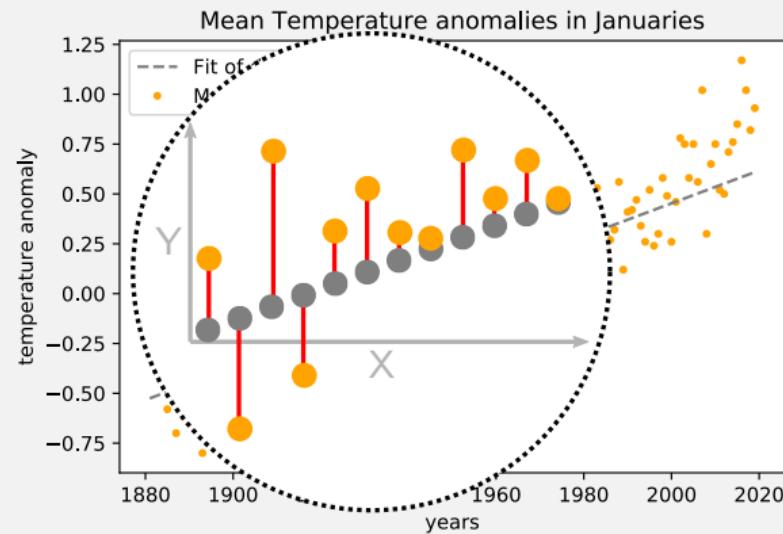
Optimization criteria

- Residual: difference predicted/observed $|Y_i - \hat{Y}_i|$
- Possible minimization criteria:
 - Sum of residuals
 - Maximum
 - Variance of residuals
- $Var_{res} := \frac{1}{N} \sum_i (Y_i - \hat{Y}_i)^2 = E[(Y - \hat{Y})^2]$
- Minimize Var_{res} = ordinary least squares optimization



Optimization criteria

- Residual: difference predicted/observed $|Y_i - \hat{Y}_i|$
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- Minimize Var_{res} = ordinary least squares optimization



Coefficient of determination R^2

How to measure quality of fit?

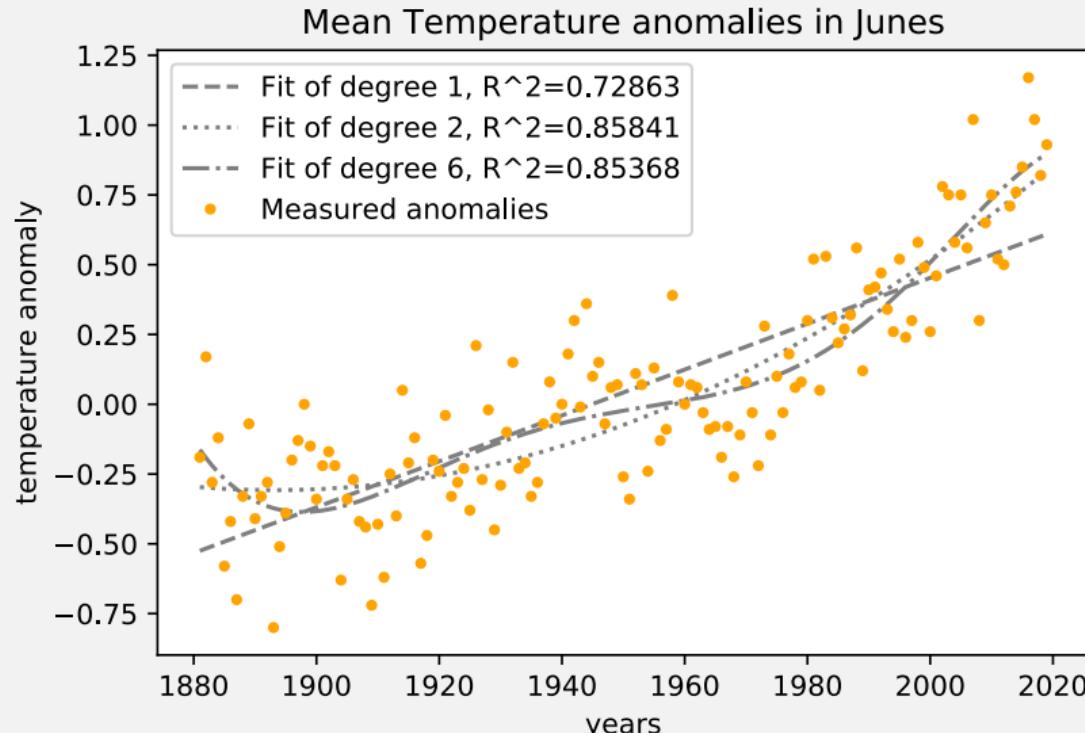
Recall: Ordinary Least squares optimization minimizes Var_{res}

R^2 is a normalized measure thereof:

$$R^2 := 1 - \frac{\text{Var}_{\text{res}}}{\text{Var}(Y)}$$

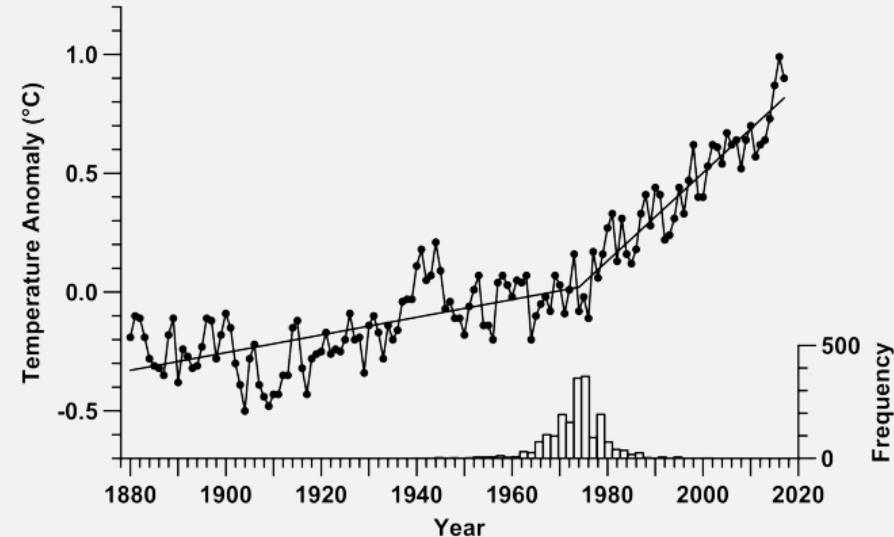
i.e., $R^2 \rightarrow 0$ bad fit, $R^2 \rightarrow 1$ good fit

Separated training from testing



The “break” model

- Combination of two linear functions
- Estimated break around year 1974 ± 5.9

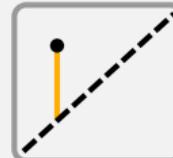


source: Mudelsee, M. (2019). Trend analysis of climate time series: A review of methods. *Earth Science Reviews*, 190, 310–322.

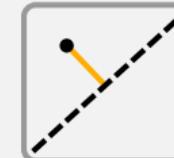
Quiz

True or false?

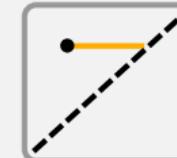
- The residual is the distance between an observed and its predicted data point
 - Linear regression always minimizes the variance of residuals
 - Linear regression is the task of fitting a line to a set of data points
 - Ordinary least squares always minimizes the variance of residuals
- How does linear regression measure the distance between an observed and its predicted data point?



(a)



(b)

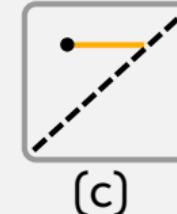
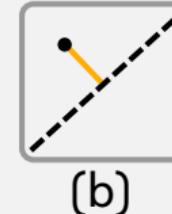
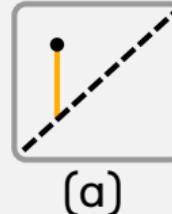


(c)

Quiz

True or false?

- The residual is the distance between an observed and its predicted data point true
 - Linear regression always minimizes the variance of residuals false
 - Linear regression is the task of fitting a line to a set of data points false
 - Ordinary least squares always minimizes the variance of residuals true
- How does linear regression measure the distance between an observed and its predicted data point? (a)



Recap

Summary

- Plots with `matplotlib`:
 - Line- and scatter plot
 - Histogram
 - Whisker (box) plot
- Numpy
 - `ndarray` data type
 - Vectorized operations, broadcasting
 - Curve fitting: `polyfit()`
- Realistic data analysis: climate trends

What comes next?

- ▶ Draw your first plots with `matplotlib`
- ▶ Further reading about NumPy: Chapter 2 of the “Python Data Science Handbook”:
<https://jakevdp.github.io/PythonDataScienceHandbook/>
- ▶ Due date for this week’s exercises is **Sunday, May 24, 2020**.

Next lecture: Pandas, applied machine learning, etc. ...