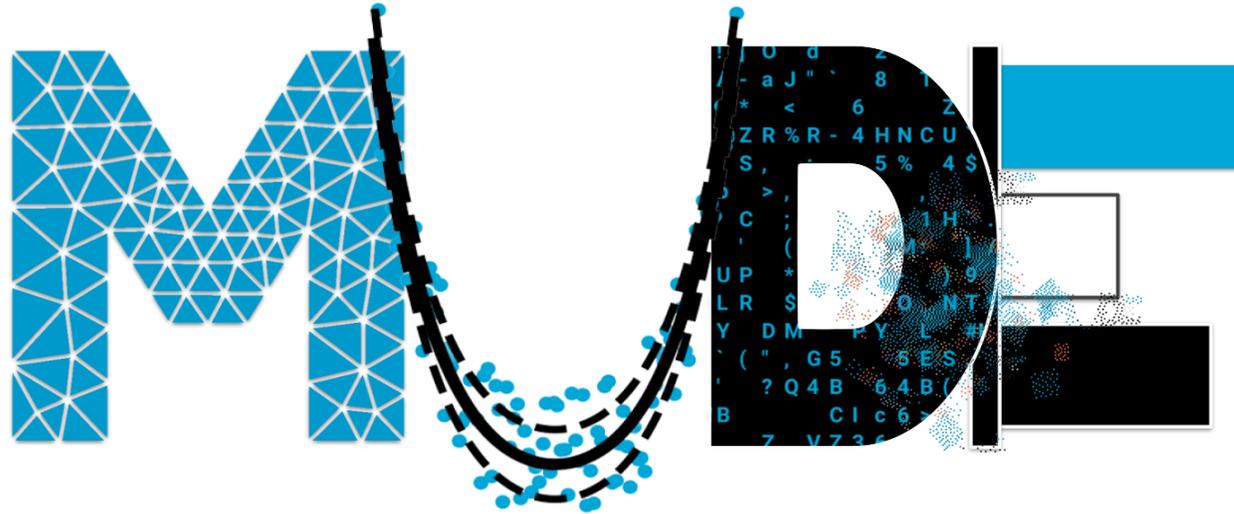


Welcome to...



Modelling, Uncertainty, and Data for Engineers

WEEK 2!

Today's Lecture

- Recap of last week (activities)
- Demonstration (software)
- In-class activity: modelling concepts and an uncertain future
- Recap of last week (topics) + a quick look ahead
- Week 1.2 Topic: Uncertainty Propagation

What you should have done last week

- Find MUDE Website, online textbook, MUDE Files. Review course policies.
- Wednesday in-class session:
 - README.md instructions
 - Workshop (WS 1.1): Able to use conda and VS Code on your computer
 - Programming Assignment (PA 1.1): can run a notebook

>>> Not critical to install everything exactly as in book; as long as you can activate `mude-base`, run a notebook

>>> If you are still having issues, come to office hours!
- Friday in-class session:
 - You have a group assigned in Brightspace and met them Friday
 - Group Assignment (GA 1.1): nothing has to be turned in; solution online via MUDE Files page
 - BuddyCheck: released Friday, due Monday before 11:00. Just click the link and do it (more info on Wed)

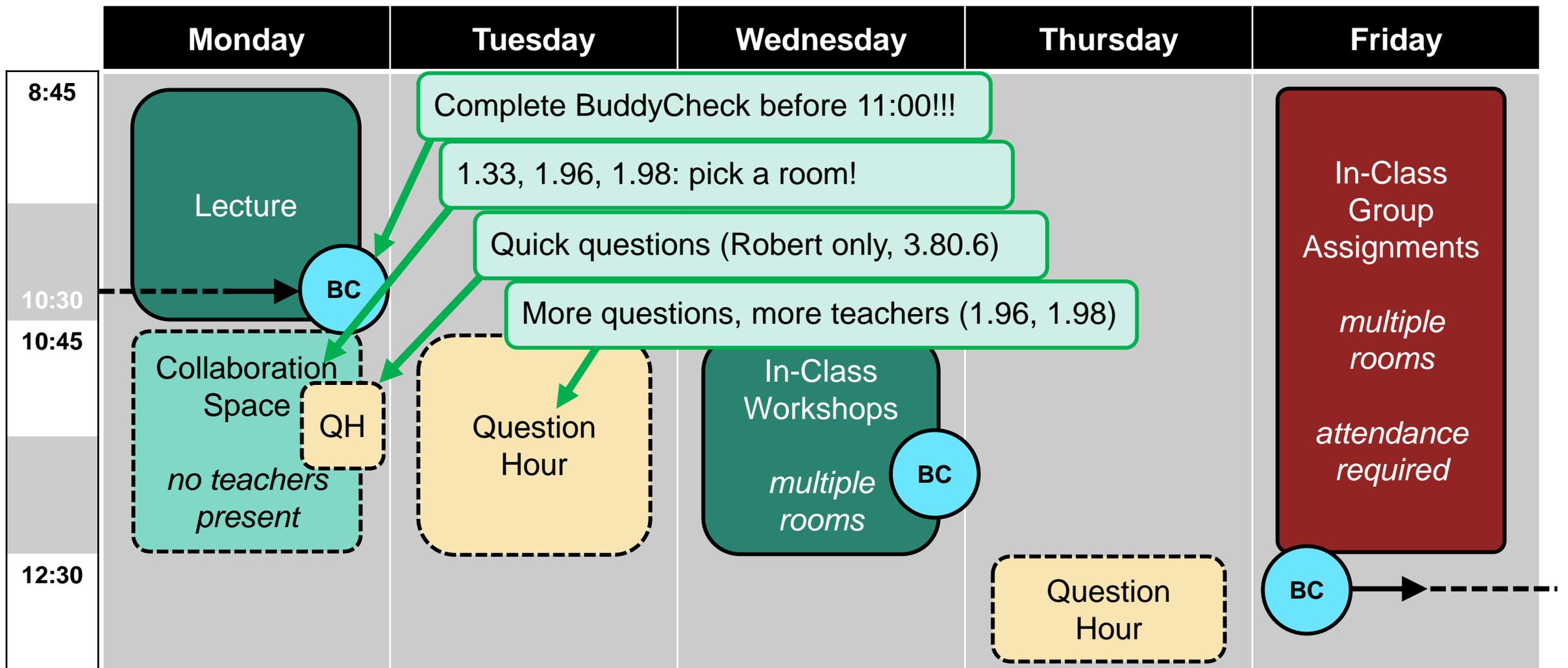
>>> No group? Not sure who is in your group? Other questions? Email MUDE-CEG@tudelft.nl

Demonstration

- Python, conda, etc
- VS Code
- Jupyter Notebooks

Demonstration

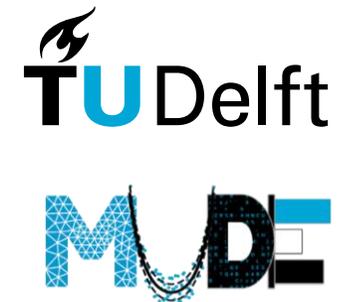
- MUDE Website, Book, Files
- Setup of computer and Python
 - Working directory setup
 - Checking conda: `conda --version`
 - Listing environments: `conda env list`
 - Looking at file `environment.yml`
- Some fun conda stuff
 - `conda list`
 - `conda info`
- Setup of VS Code
 - Opening a new folder / workspace
 - File viewer, Extensions
 - Terminal: setting defaults, checking conda
 - Activating an environment
- Jupyter Notebooks
 - Overview of structure
 - Cell types
 - Running a cell, selecting environment
 - Restarting a kernel
 - Saving outputs (or not)
- Some fun with AI?



Programming Assignment: any time during the week, but... **Finish before Friday!**

BC = BuddyCheck: opens Fri (closes Mon); review results Wed with group

Question Hours (optional): Mon 11.00-12.00, Tue 10:45-12:30, Thu 12:30-13:30



In-Class Activity

Useful links:

- mude.citg.tudelft.nl/2024/book/modelling/gof.html
- mude.citg.tudelft.nl/2024/files/Week_1_2/

- We will use the file `In_Class_Activity` (located in MUDE Files, Week 1.2)
→ Read directly via the browser; can also download zip and run notebook (not needed in class)
- Part 1-3: import data, preliminary analysis, create a model. **Validation: Goodness of Fit.**
- Part 4: confidence intervals. **Validation: would you use it?.**
- Part 5: a non-linear model. **Validation: Goodness of Fit. Which model is better?**

Read quickly over Parts 1-3 (stop at Part 4), then we will answer these questions (together):

General: what is the “model?”

3.1: What is the coefficient of determination? Is the model accurate?

3.2: Is the visual GoF consistent?

3.3: What is RMSE telling us in general? What does it tell us about the model?

3.4: What about *rbias*?

Task 3.1

Coefficient of determination $R^2 = 0.151$

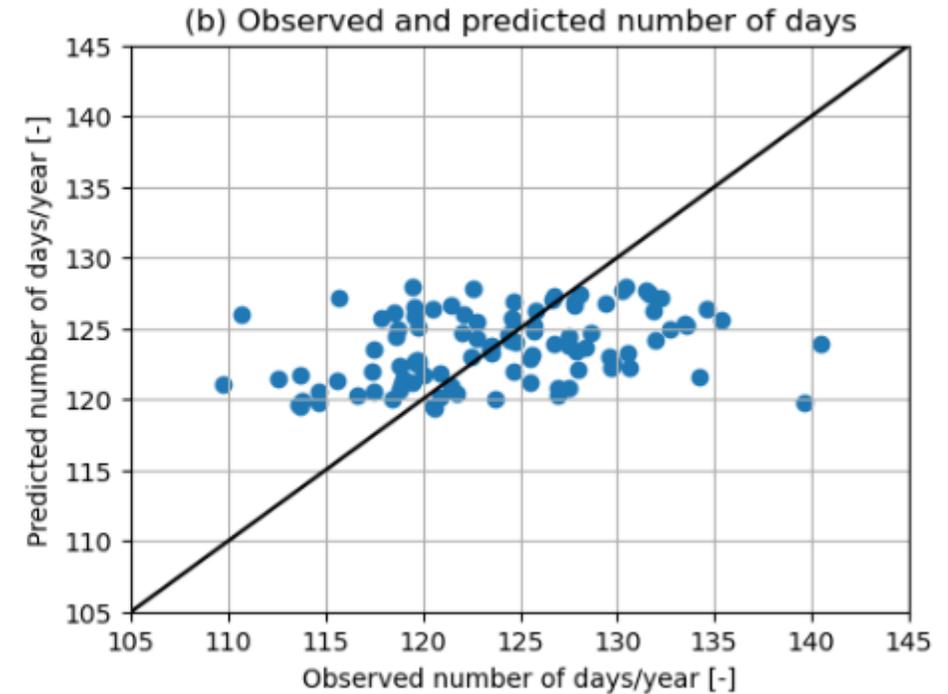
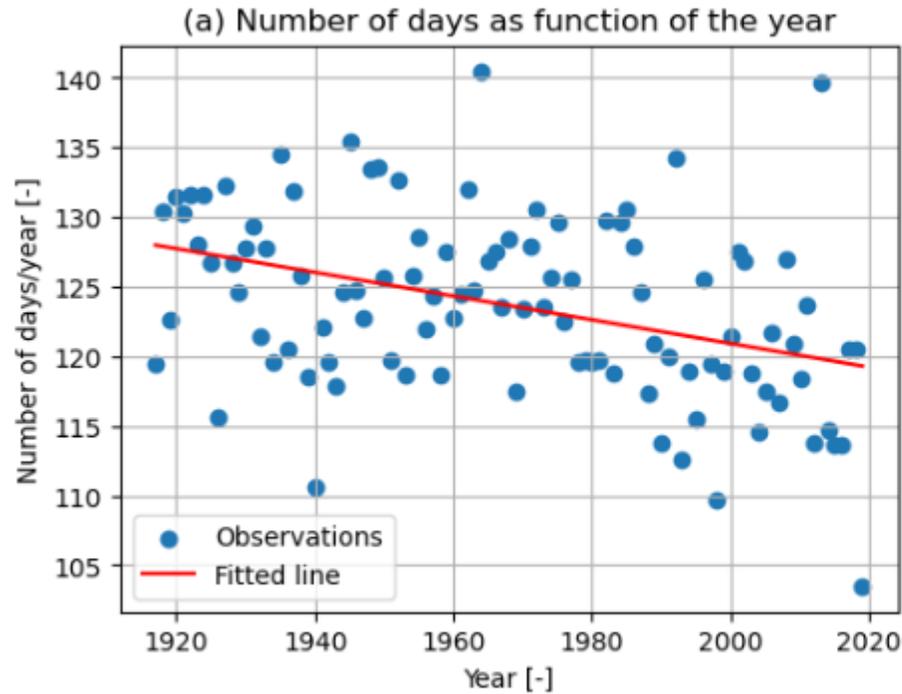
Intercept $q = 291.239$

Slope $m = -0.085$

Solution

1. Coefficient of determination R^2 measures the percentage of the variance in our observations explained by the model. Thus, the higher, the better. As we can see, the value of R^2 is quite low. Only $\approx 15\%$ of the variance is explained by the model, which is very low. Therefore, the linear model is not able to explain the scatter in our observations.
2. Based on the answer to the previous question, the linear model is not an accurate model. Whether this low level of accuracy is good enough or not, depends on the use we want to give to the model.

Task 3.2



Solution

In the plot (a), we observe that the observations have a high scatter around the fitted line, and visually it looks consistent with the R^2 value found above.

In the plot (b), we compare the observations with the predictions of the model. The perfect fit would correspond to the 45-degree line in black. Thus, the model has a poor performance as we already quantified using the coefficient of determination. Both results are aligned.

Task 3.3, 3.4

$$\text{RMSE} = 6.003$$

Solution

1. *RMSE* represents the mean error between the observations and the predictions of the model. This means that the mean error is ≈ 6 days. Therefore, the linear model is not able to explain the scatter in our observations.
2. Based on the previous interpretation, the linear model is not accurate. Whether this low level of accuracy is good enough or not, depends on the use we want to give to the model.

$$\text{rbias} = 0.002$$

Solution

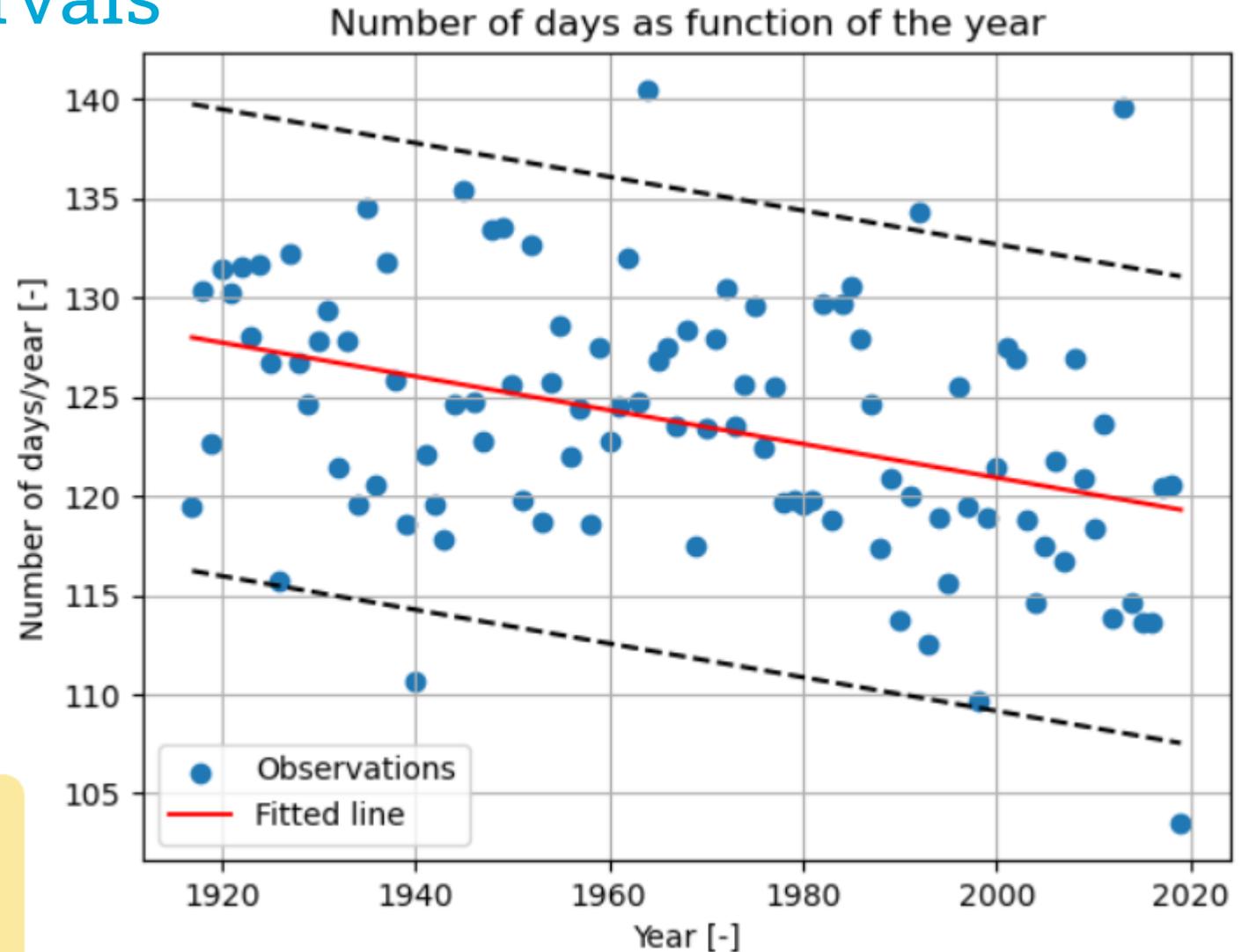
1. `rbias` provides an standardized measure of the systematic tendency of our model to under- or over-prediction. It is very low for our model, so it does not have a clear tendency to under- or overestimate and, thus, does not seem to be biased.

Part 4: Confidence Intervals

- Study the code and try to identify what the analysis accomplishes.
- What can you conclude from the figure?
- Would you bet \$3? What about \$3000?

Part 4: Confidence Intervals

- What can you conclude from the figure?
- Would you bet \$3? What about \$3000?



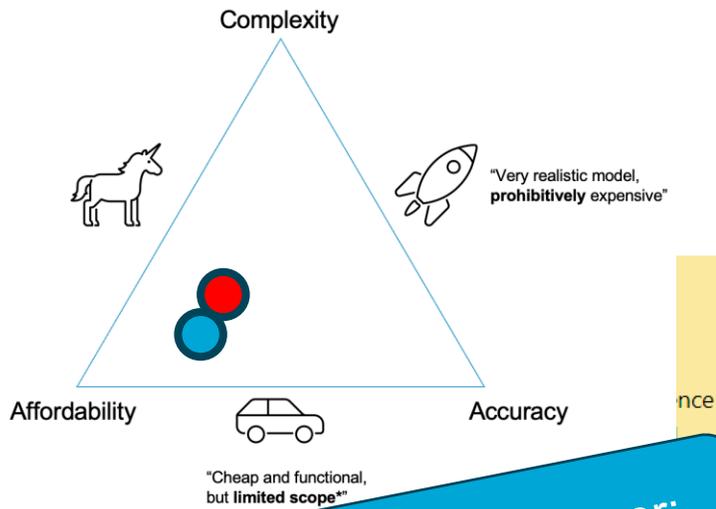
Solution: If you consider that you need to place a bet with not only the day but also the hour and minute at which the ice would break, the model is not accurate enough. You can see that the confidence interval spans almost 20 days!

Part 5: Non-linear Models

- Study the code and try to identify what the analysis accomplishes.
- What is the “model?”
- 5.1: is the quadratic model better than the linear one?

Part 5: Non-linear Models

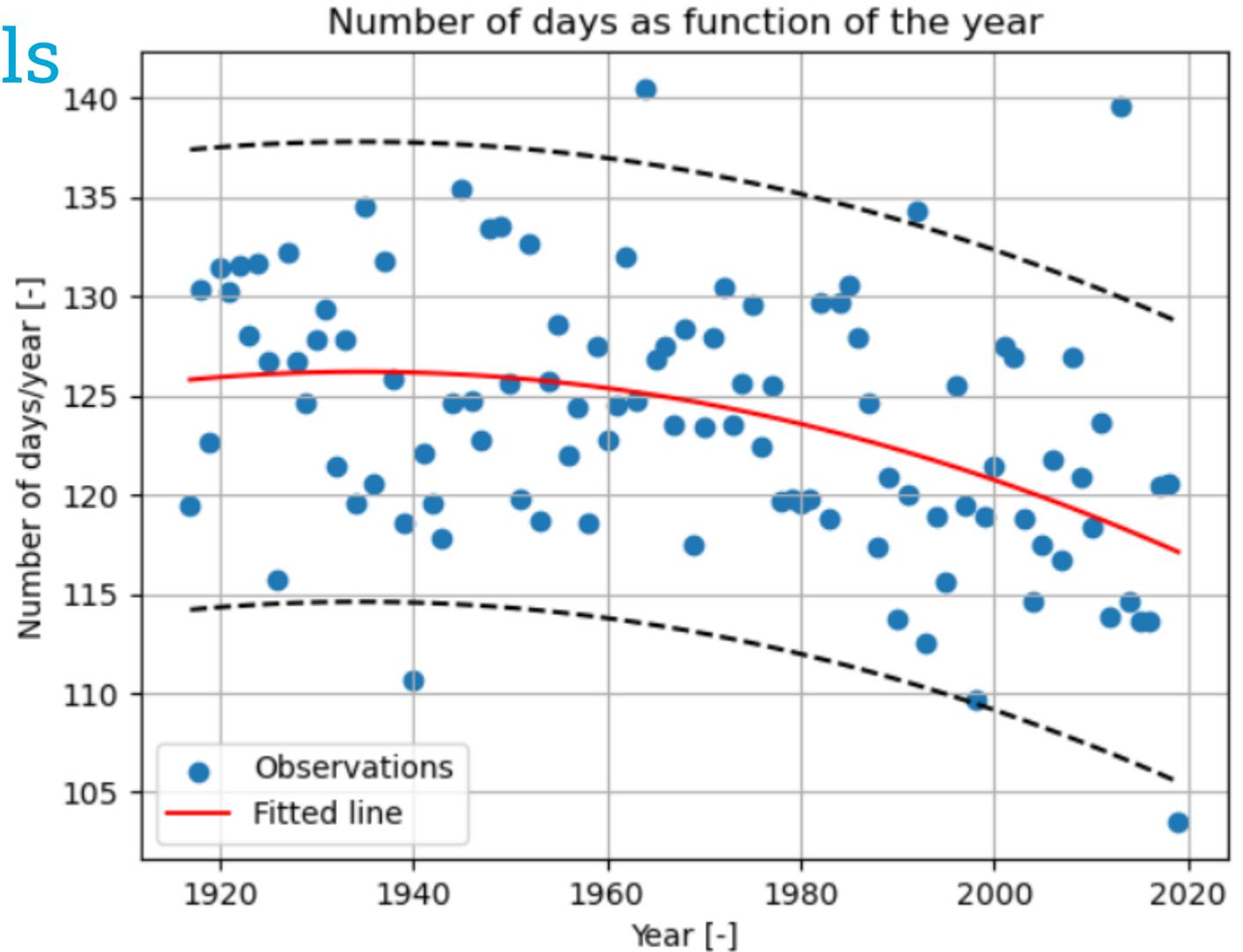
- What is the “model?”
- 5.1: is the quadratic model better than the linear one?



Solution: No! The GOF m complicated (e.g., it includ

At the same time we can bounds. Moreover, the 'm number of days. This varia (e.g., number of heatwaves), the question is then, w related explanatory variable to get a b whether you are int

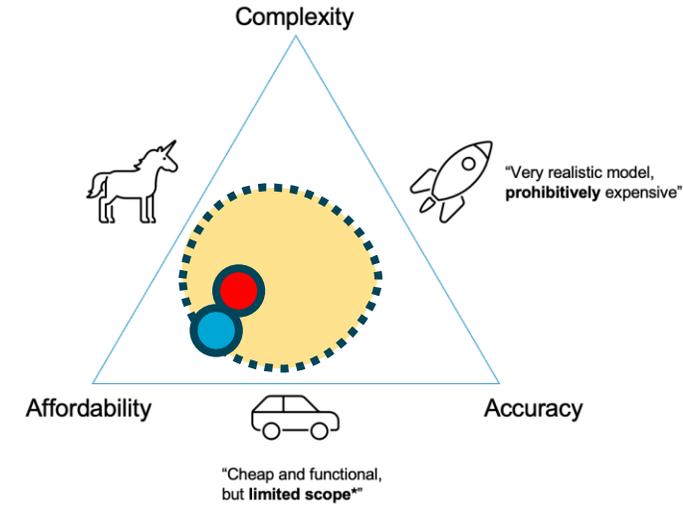
See solution on MUDE Files...short answer:
No! (added complexity not worthwhile)



Linear Model

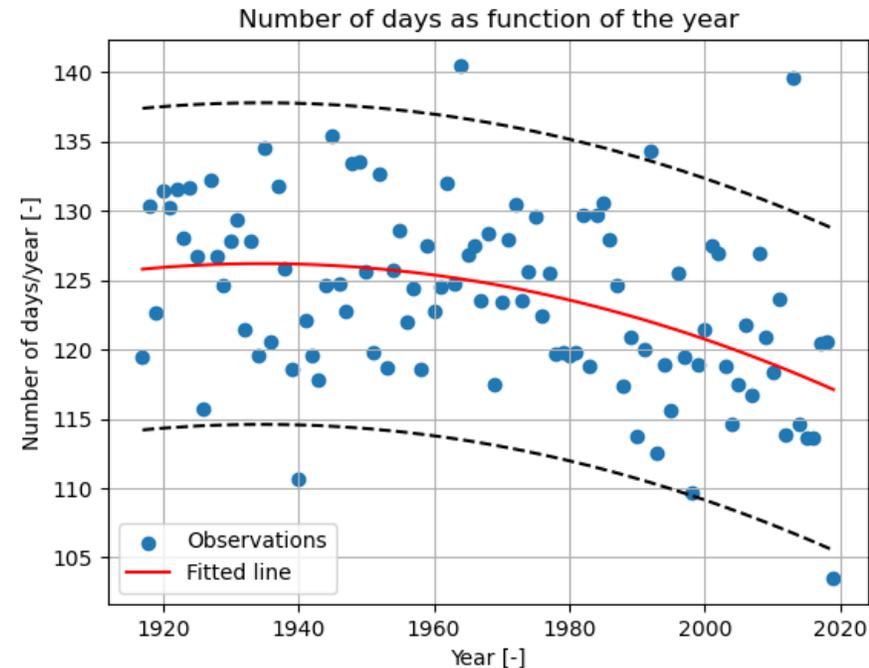
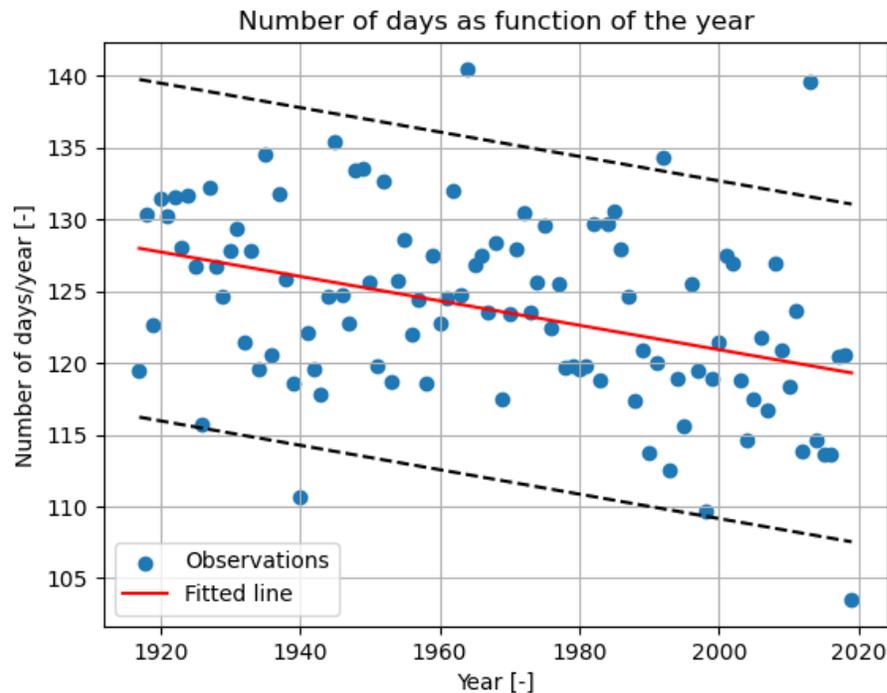
Quadratic Model

Were our models good enough?



	No. Parameters	RMSE	R ²	rbias
Line	2	6.00	0.15	0.002
Parabola	3	5.92	0.18	0.002

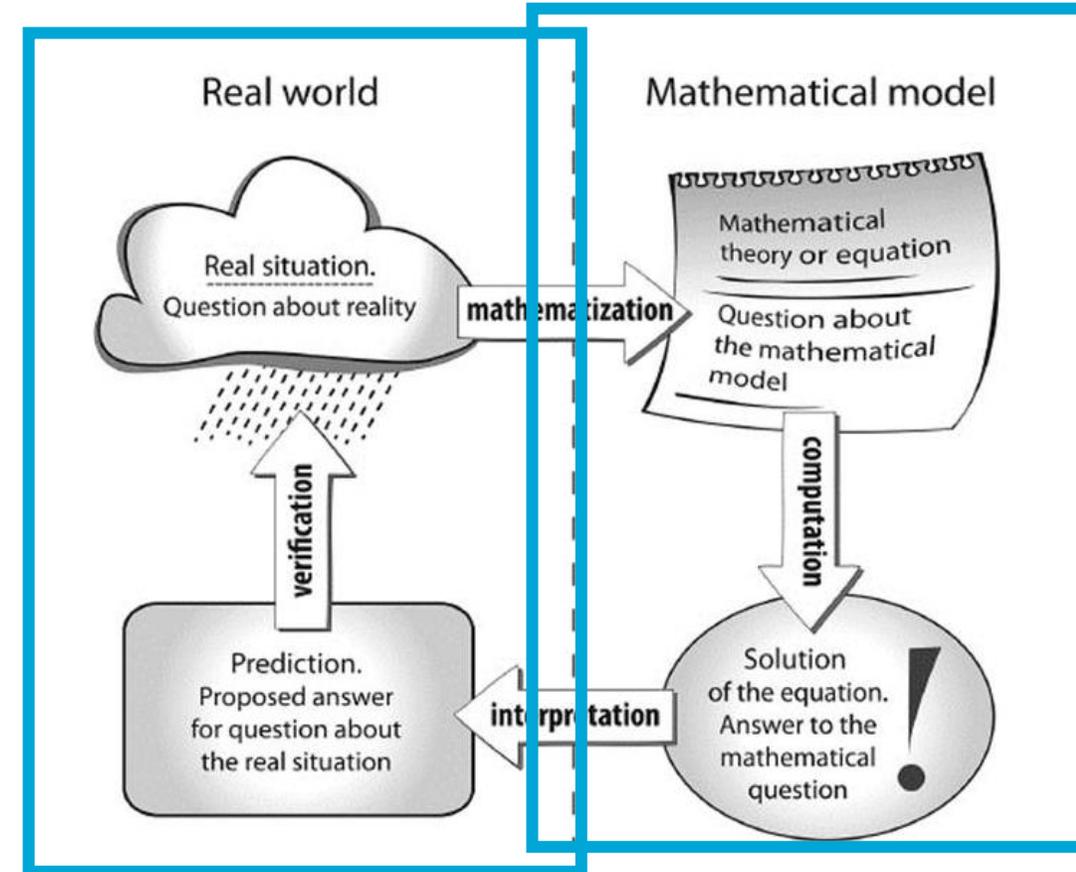
-  Linear Model
-  Quadratic Model
-  Let's go here!



Recap of Week 1 Topics

1. GA 1.1, Task 1: affordability-complexity-accuracy diagram
2. GA 1.1, Task 2, 3: easy to connect variables; process harder →
3. In-Class Activity: simple model is easy, not accurate
4. Real life is an iterative approach:

data / phenomenological / mechanistic



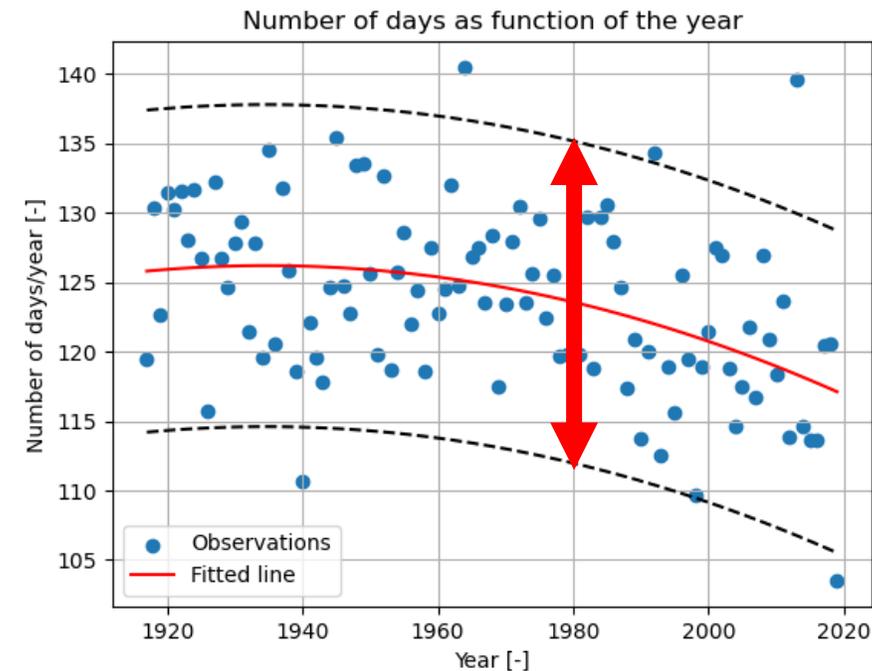
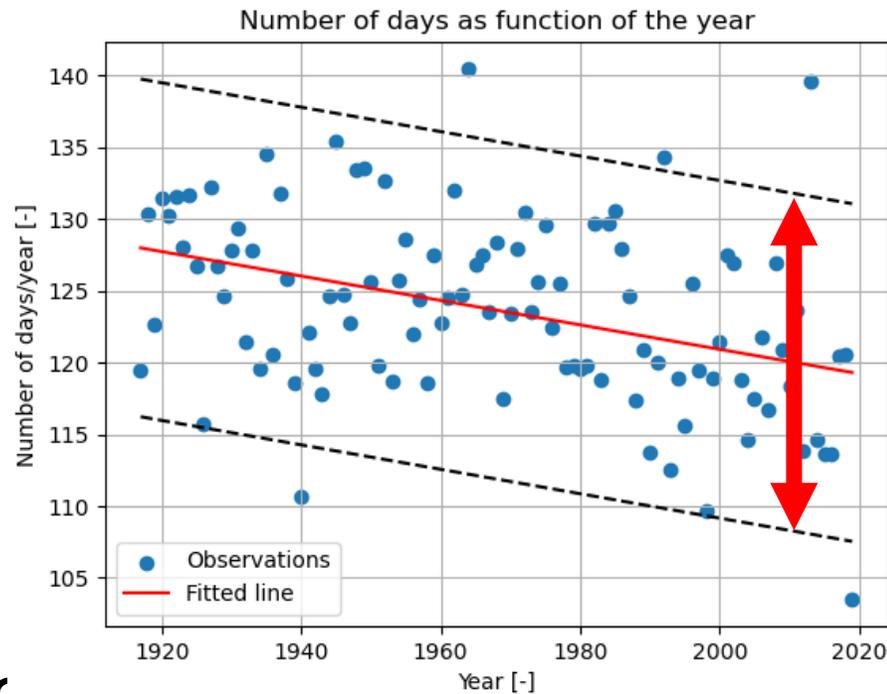
Where we are going next

- Next few weeks (2-4): data-driven, with a hint of phenomenological
 - Weeks 5-6: mechanistic
 - Week 7: stochastic (probabilistic)
 - Week 8: combination!
-
- Linear: useful for learning, understanding
 - Non-linear: reality!
-
- First, we need a few tools to help quantify uncertainty (Week 2).

Let's revisit last week to see what we need...

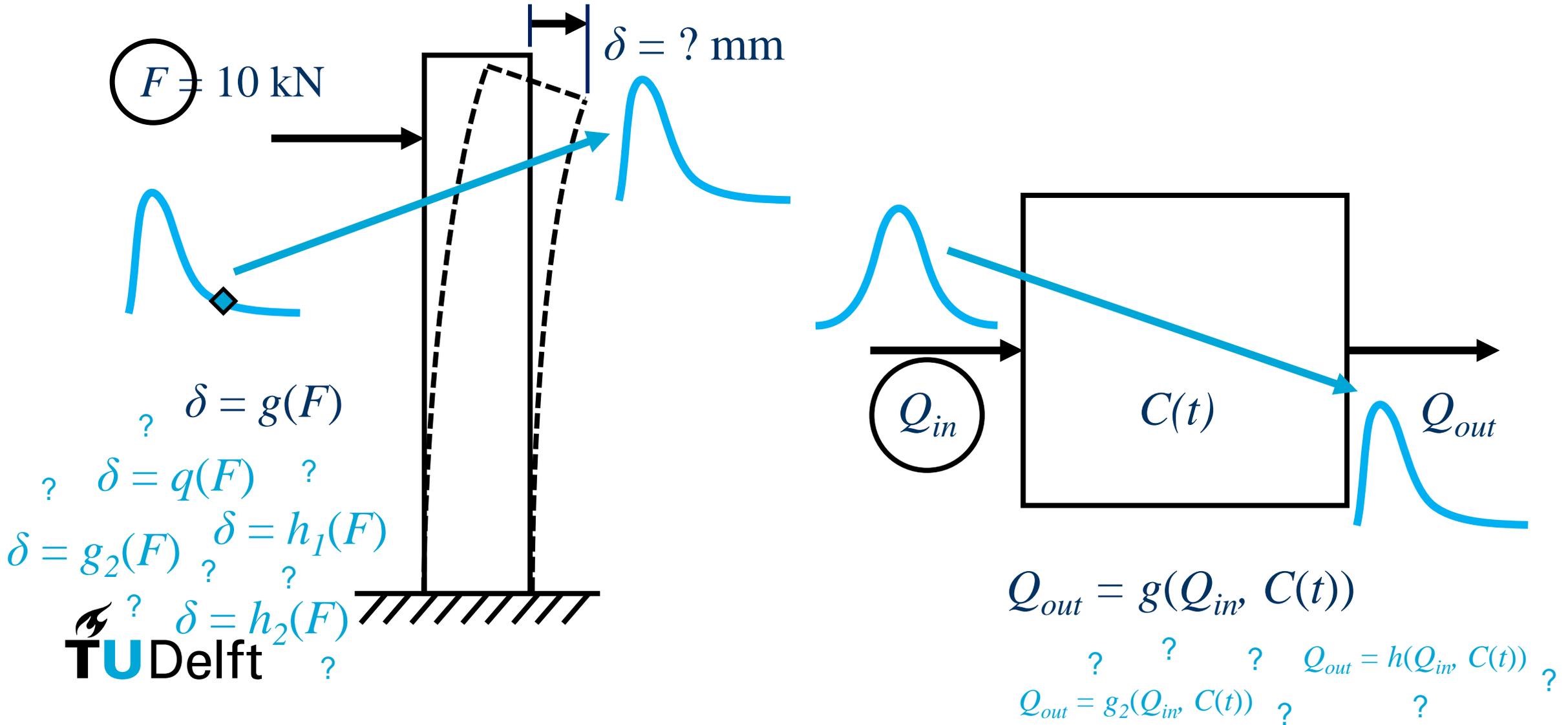
Can we make some sense of the large CI?

	No. Parameters	RMSE	R ²	rbias
Line	2	6.00	0.15	0.002
Parabola	3	5.92	0.18	0.002



Uncertainty – What is it?

Deterministic design (pre-MUDE) – parameters given

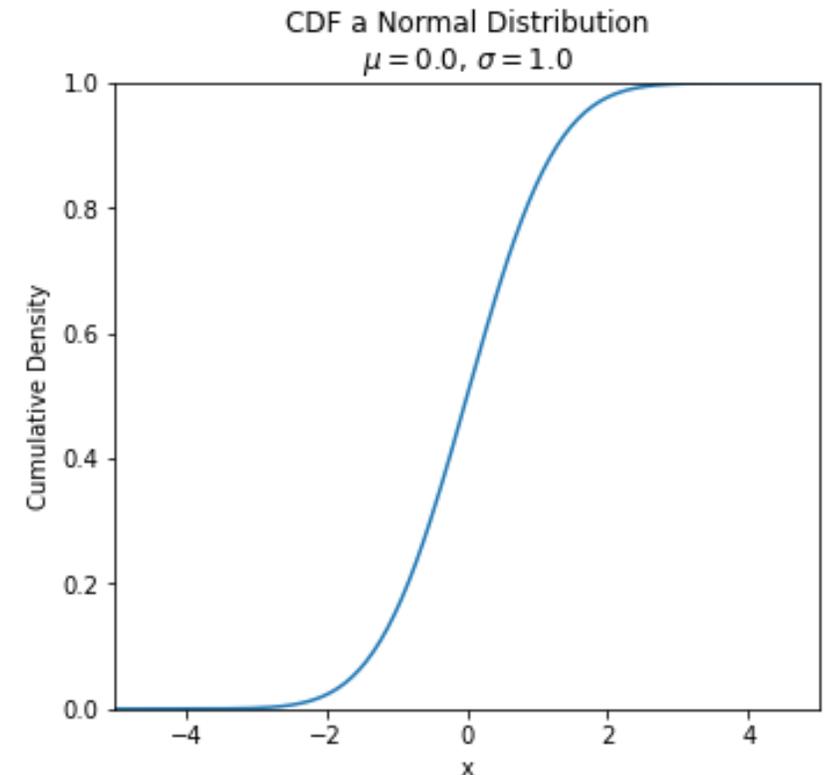
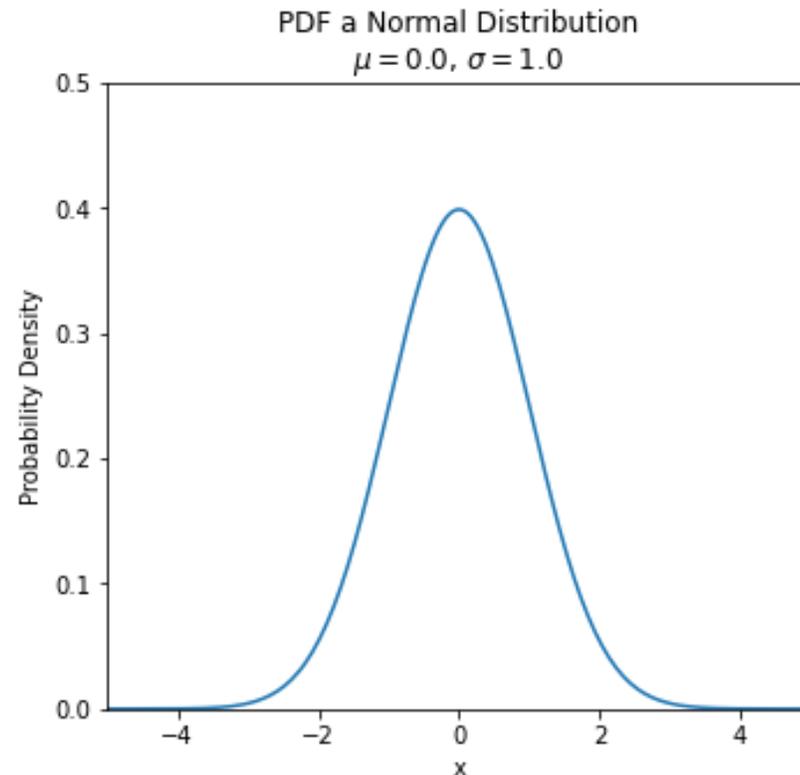


Uncertainty – What is it?

- “Uncertainty Quantification” ... have you seen this term?
 - We avoid it in MUDE: often presented as a “black box”; assumptions can be unclear / used inconsistently.

Instead, we will focus on theoretical concepts: fundamentals of probability and statistics

- We will rely on:
 - Expectation, Variance
 - Covariance / Dependence
 - Probability Distributions: continuous, parametric
 - Set Theory



Uncertainty Classification – Introduction

Aleatoric

- intrinsic phenomenon; typically associated with variations that occur in nature

Epistemic

- lack of knowledge; often called model uncertainty

Error

- deficiency in any stage of modelling/simulation not due to lack of knowledge

Break-up Day Uncertainty

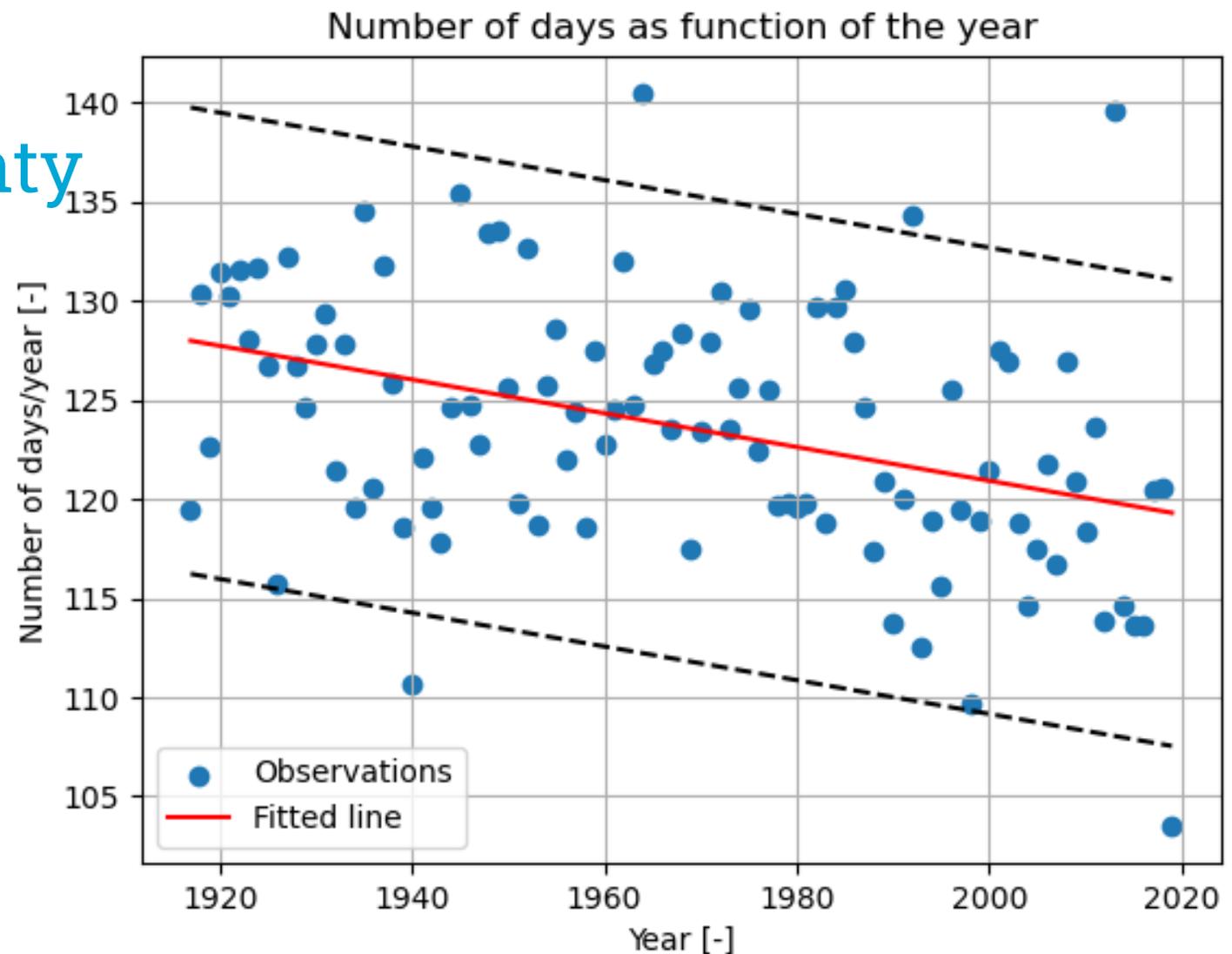
- Days, D , as a function of year, Y :

$$D = q(Y) \quad \gg \gg \quad D = m \cdot Y + b$$

$\gg \gg$ Use a distribution to represent uncertainty

Sources of uncertainty?

- Variation of natural phenomena
 - Aleatoric
- Model uncertainty?
 - Epistemic; not relevant
- Observations?
 - Error (accuracy / precision of historic points)
 - Epistemic (“model” of break-up / definition)



Confidence Intervals

- Days, D , as a function of year, Y :

$$D = q(Y) \quad \gg \gg \quad D = m*Y + b$$

```
k = conf_int(data[:,1], line, 0.05) ←
ci_low = line - k
ci_up = line + k

#plot
plt.scatter(data[:,0], data[:,1], label = 'Observations')
plt.plot(data[:,0], line, color='r', label='Fitted line')
plt.plot(data[:,0], ci_low, '--k')
plt.plot(data[:,0], ci_up, '--k')
plt.ylabel('Number of days/year [-]')
plt.xlabel('Year [-]')
plt.grid()
plt.legend()
plt.title('Number of days as function of the year')
```

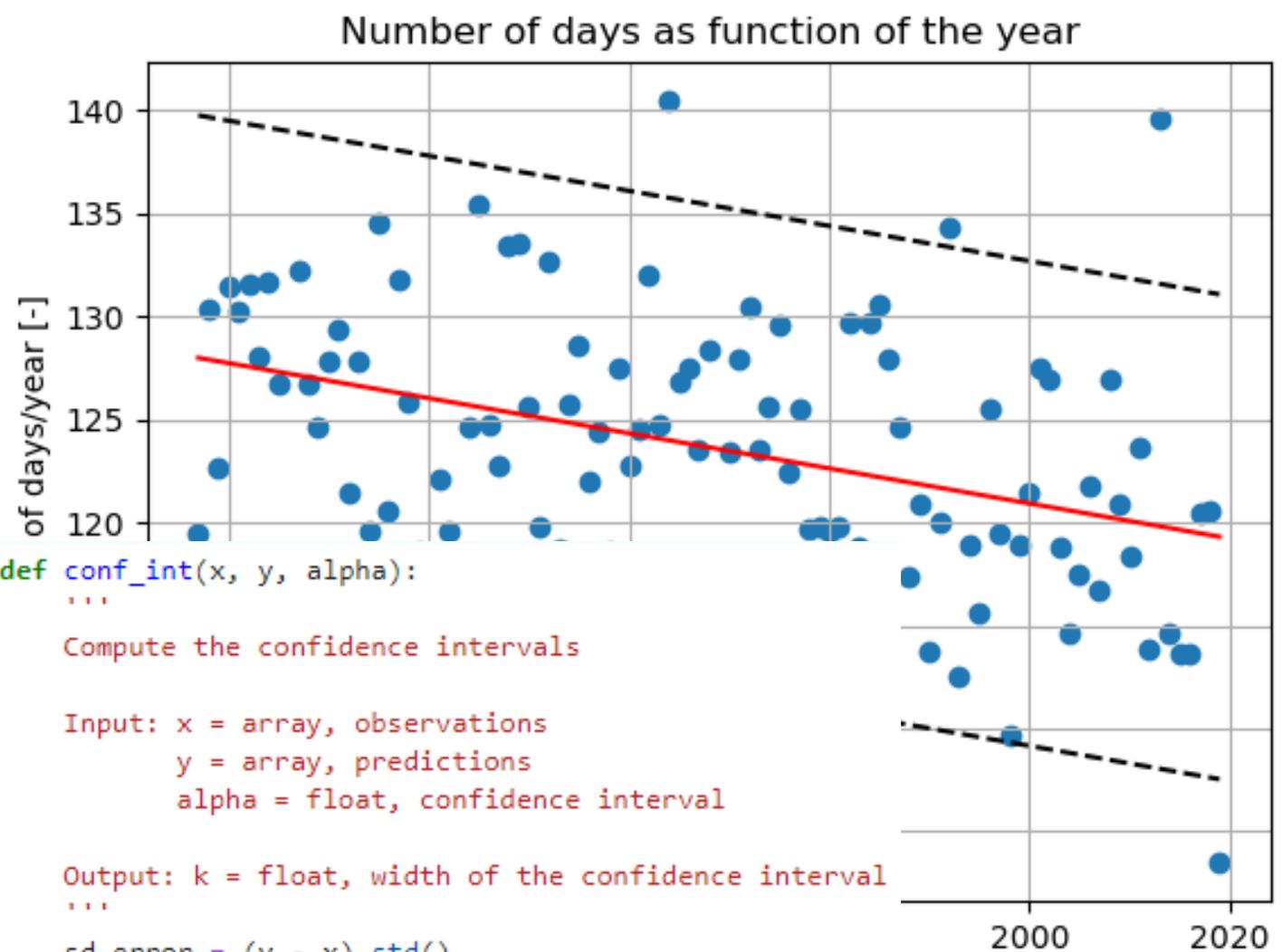
```
def conf_int(x, y, alpha):
    ...
    Compute the confidence intervals

    Input: x = array, observations
           y = array, predictions
           alpha = float, confidence interval

    Output: k = float, width of the confidence interval
    ...

    sd_error = (y - x).std()
    k = sci.norm.ppf(1-alpha/2)*sd_error

    return k
```



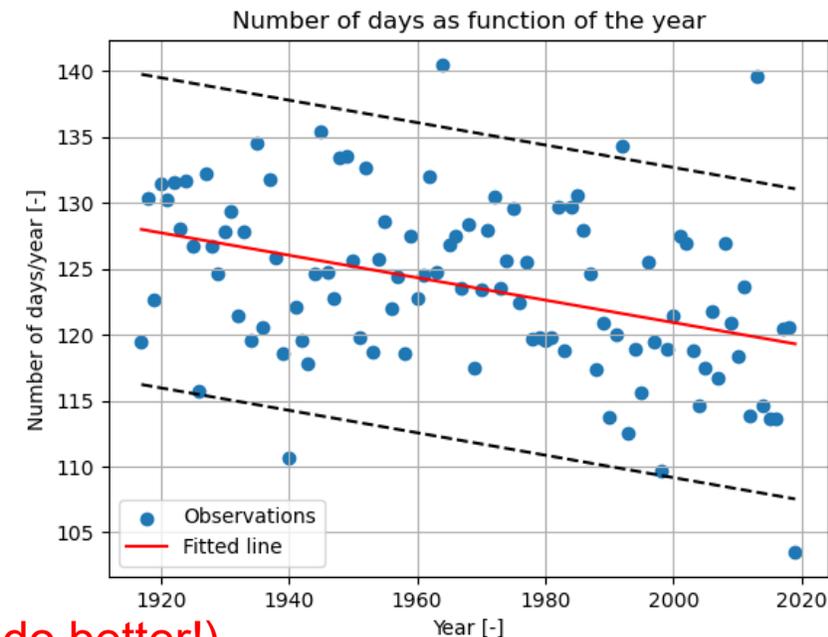
Confidence Intervals

- What did we do?
 - Assume that the uncertainty in D has the Normal distribution $D \sim N(\mu, \sigma)$
 - Mean defined by best estimate of D
 - Standard deviation defined by the standard deviation of error (constant over D)
 - Find region $D \pm \Delta D$ such that probability is 95%: $\gg \gg P(D - \Delta D \leq D \leq D + \Delta D) = 0.95$
- k is the distance that defines this region
- What does this Normal distribution of D mean? (the interpretation part)
 - Uncertainty in the “true” value of D
 - Uncertainty in our model of $D = q(Y)$
 - High variance of past observations, D_i

Uncertainty type: Error

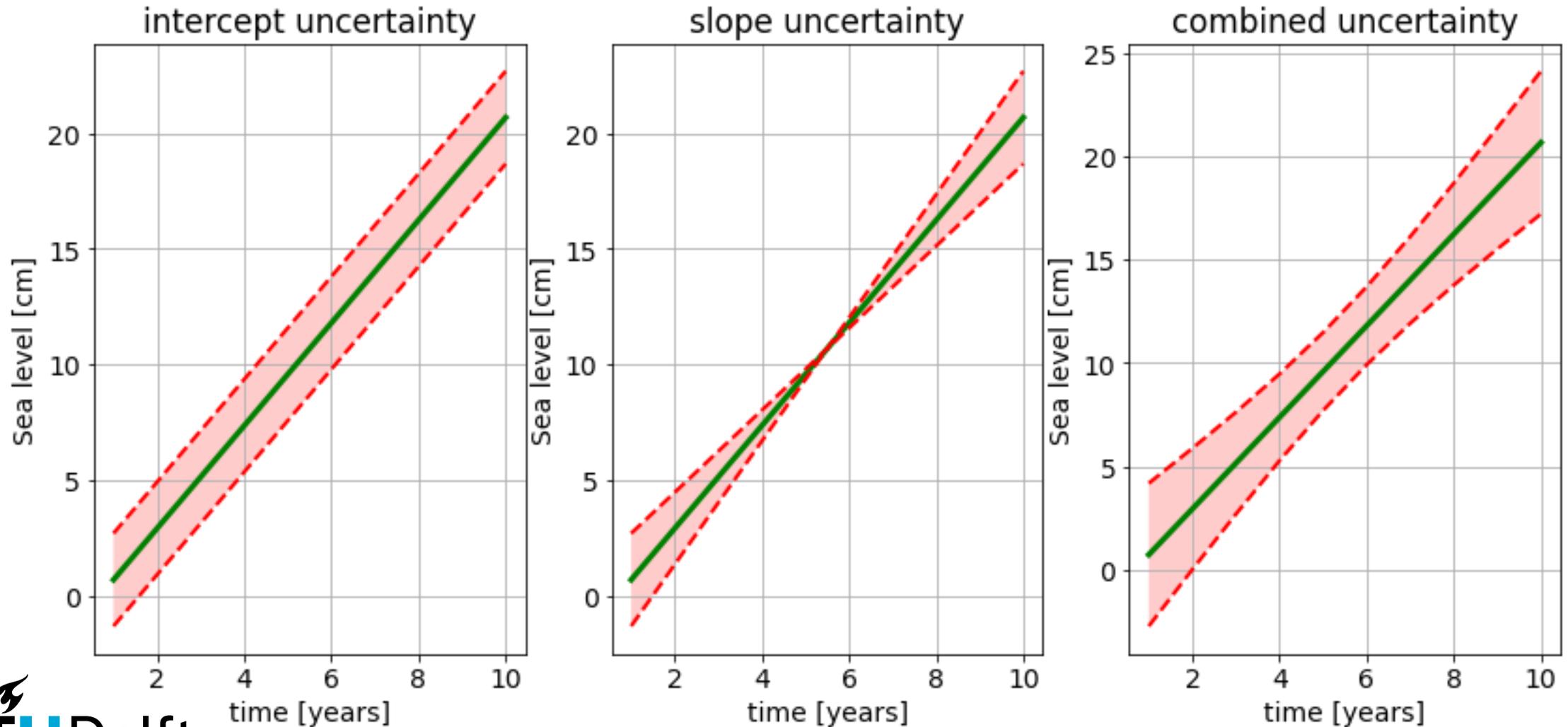
 Answer: mostly Error, but implicitly “most of the above”

Model has 2 parameters (m, b)
Unclear what contributes to σ_D
Variance not uniform in Y

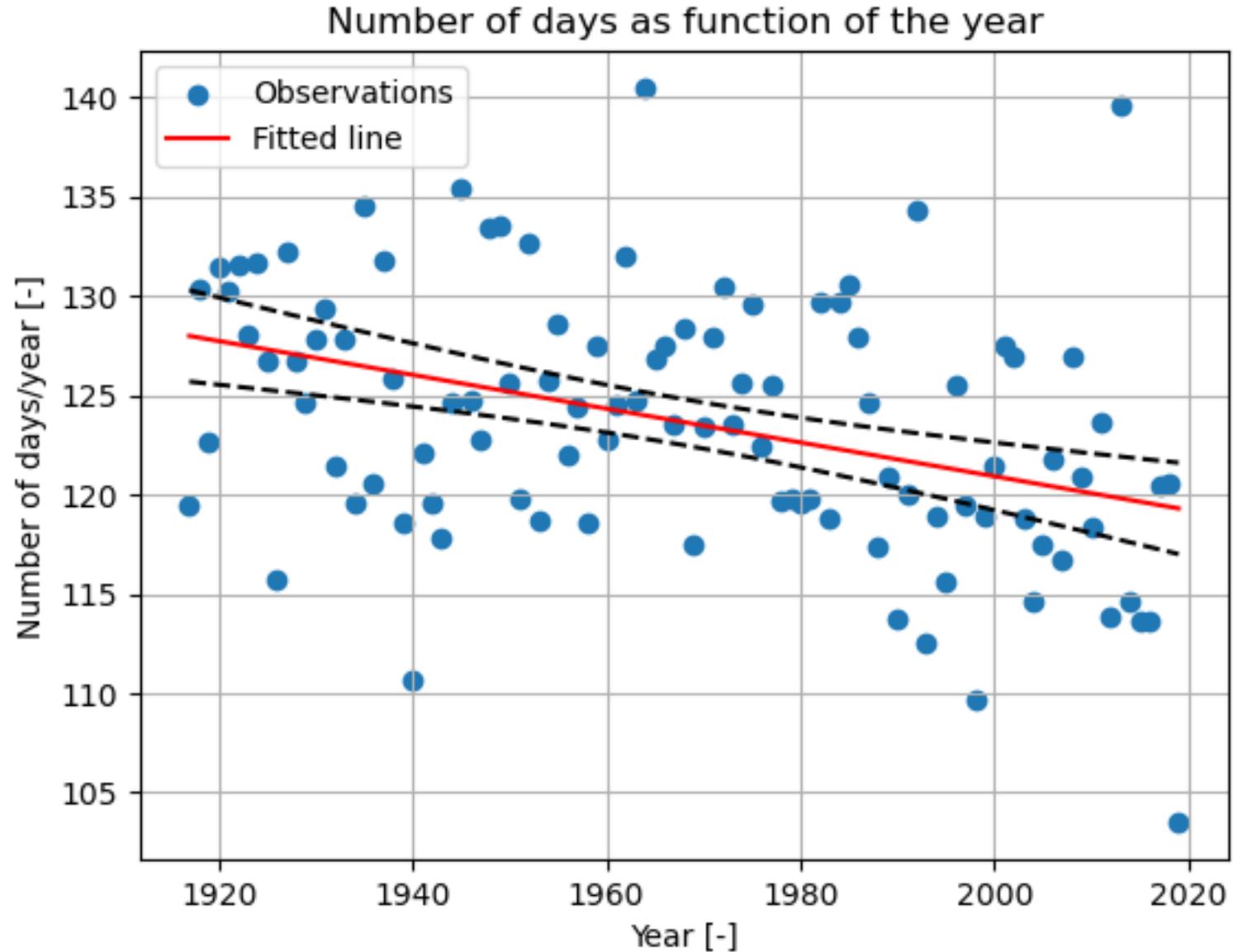


This approach = a simplification (not incorrect...but we can do better!)

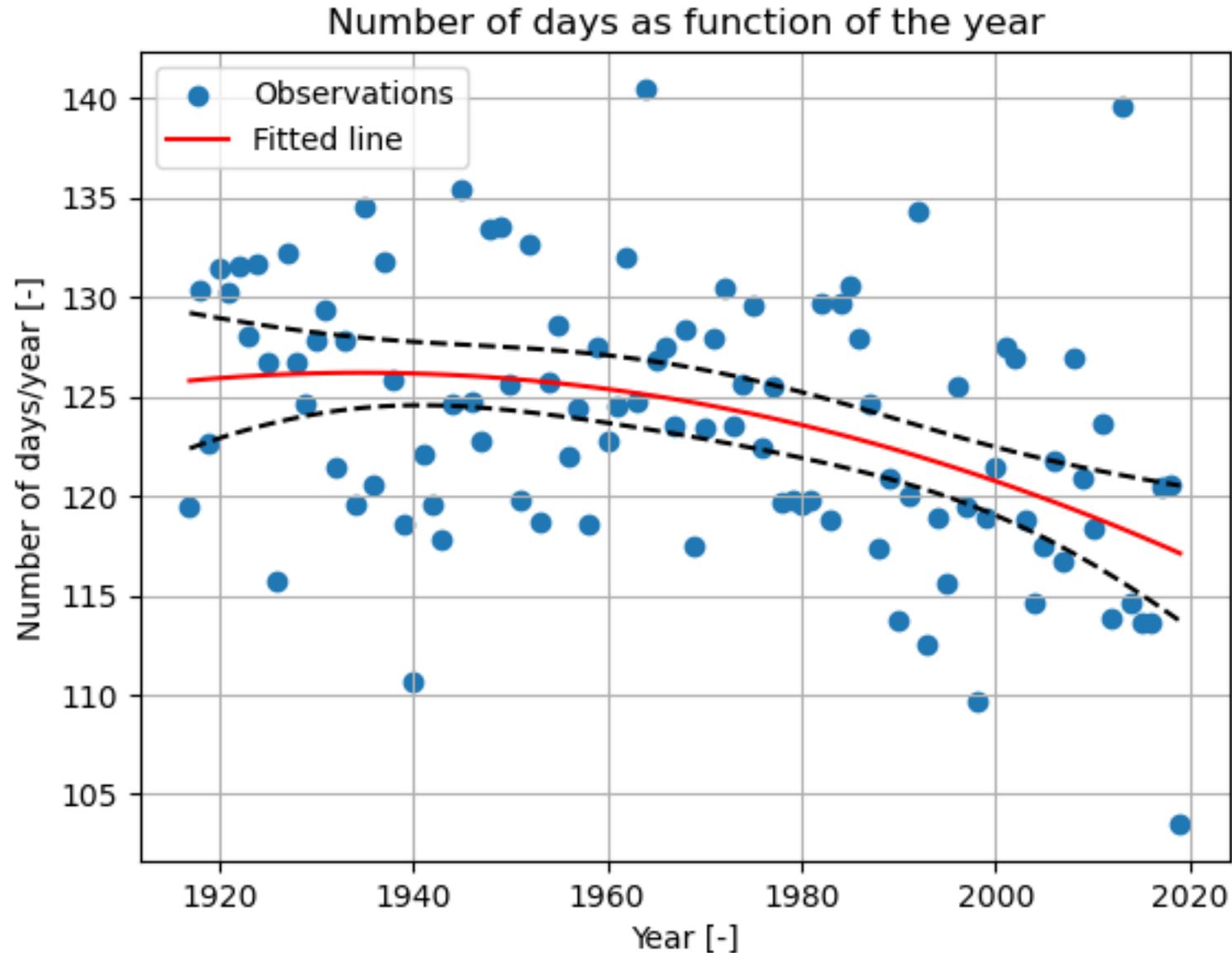
Confidence Intervals – What we will do next



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Confidence Intervals – What we will do next



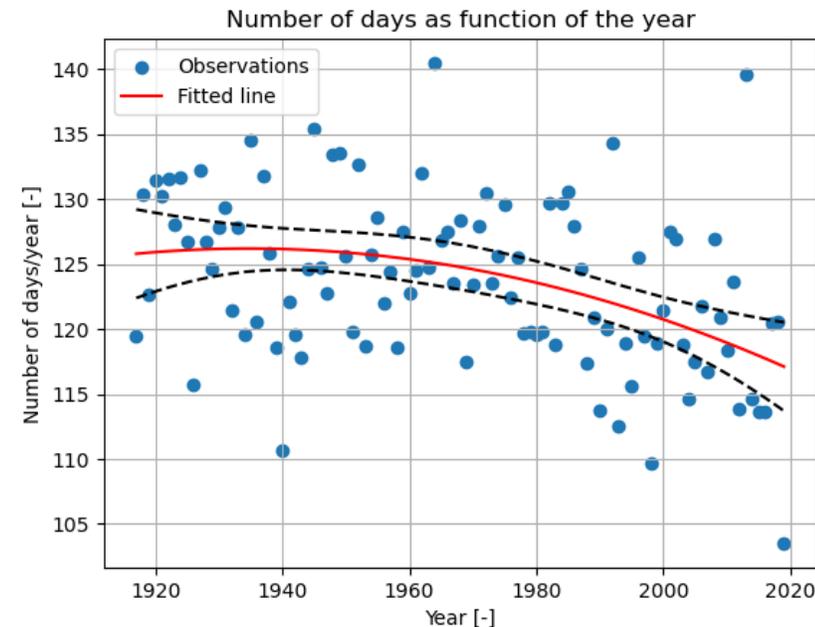
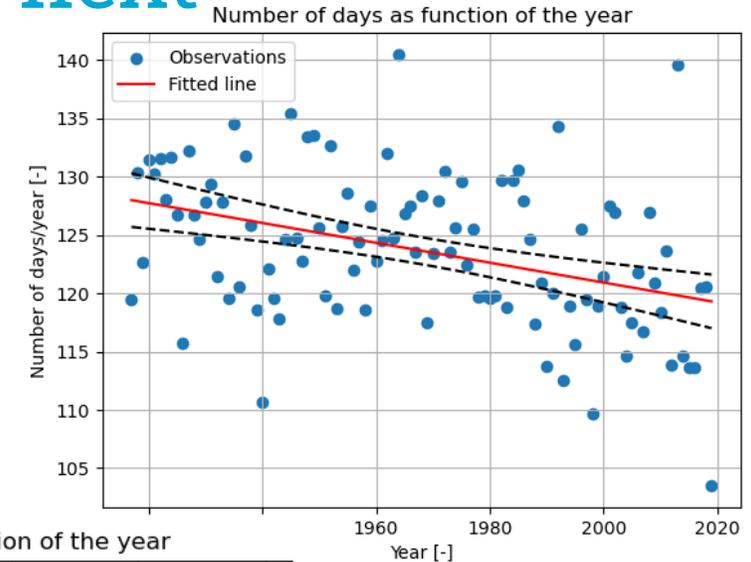
Confidence Intervals – What we will do next

Propagation of Uncertainty

- Considers imprecise measurements (μ , σ for each point)
- Estimate μ , σ for output, D , as a function of inputs, Y

>>> epistemic uncertainty, error

In week 7, 8 we will focus more on aleatoric uncertainty



Week 2 Topic: Propagation of Uncertainty

- Chapter 2 in textbook
 - Videos & PDF Slides: supplementary to the text.
 - Interactive page, multivariate normal: try it!
- Fundamentals chapters in textbook (background knowledge / reference material)
- This week in Programming (PA 1.2): numpy, linear algebra, matplotlib, statistics; "data" handling

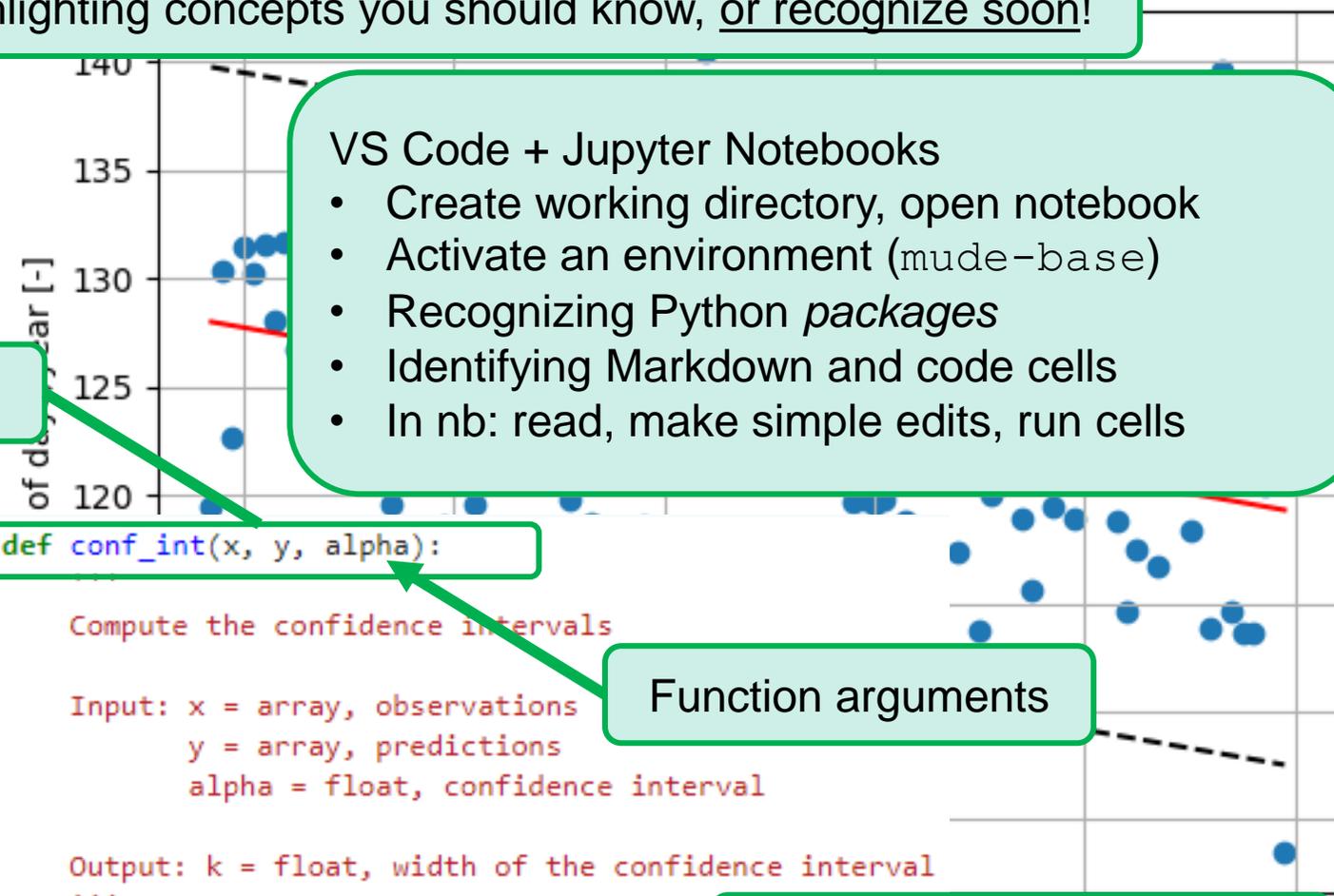
>>> PA 1.2 online (MUDE Files)

>>> Struggling with programming?

Week 2 Programming

Review of previous slide (Confidence Intervals)

Highlighting concepts you should know, or recognize soon!



- ### VS Code + Jupyter Notebooks
- Create working directory, open notebook
 - Activate an environment (mude-base)
 - Recognizing Python *packages*
 - Identifying Markdown and code cells
 - In nb: read, make simple edits, run cells

Indexing/slicing
Numpy ndarray

$$D = m \cdot Y + b$$

Function

```
k = conf_int(data[:,1], line, 0.05)
ci_low = line - k
ci_up = line + k
```

```
def conf_int(x, y, alpha):
```

Function arguments

```
    Compute the confidence intervals
    Input: x = array, observations
           y = array, predictions
           alpha = float, confidence interval
    Output: k = float, width of the confidence interval
    ...
    sd_error = (y - x).std()
    k = sci.norm.ppf(1-alpha/2)*sd_error
    return k
```

Methods of an object (OOP)

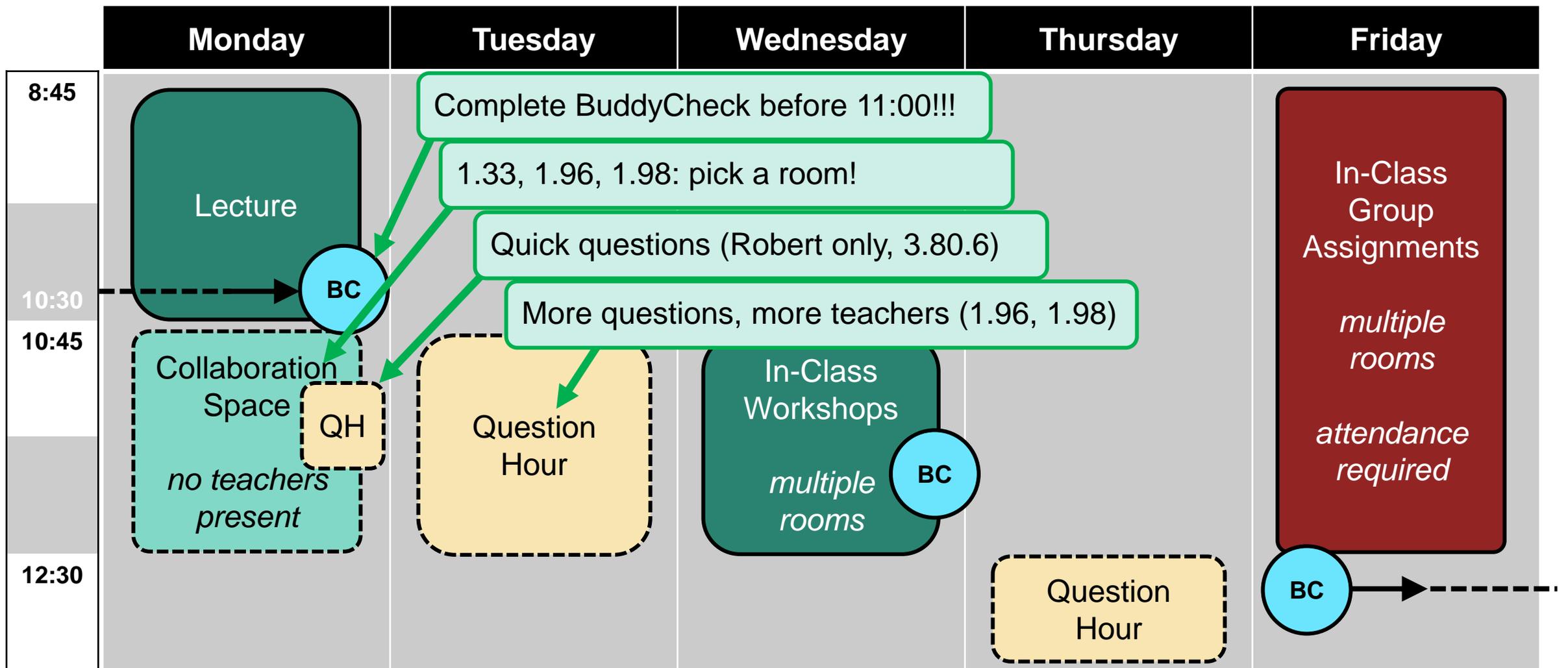
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plt.ylabel('Number of days/year [-]')
plt.xlabel('Year [-]')
plt.grid()
plt.legend()
plt.title('Number of days as function of the year')
```

Matplotlib Plot

What is k?

- ### Lost?
- Come to office hours!
 - Refer to teachbooks.github.io/learn-python





Programming Assignment: any time during the week, but... **Finish before Friday!**

BC = BuddyCheck: opens Fri (closes Mon); review results Wed with group

Question Hours (optional): Mon 11.00-12.00, Tue 10:45-12:30, Thu 12:30-13:30

Closing

- Complete BuddyCheck (NOW!)
 - Read the book (Chapter 2)
 - Work on Programming Assignment (PA 1.2)
 - Struggling with Programming? Don't miss question hours!
- >>> Don't worry about setting up VSC with Copilot now
(we will help you later)
- >>> CLI: Note that PowerShell is different than Command Prompt



Delta Quick Reaction Student Squad ^
Now!