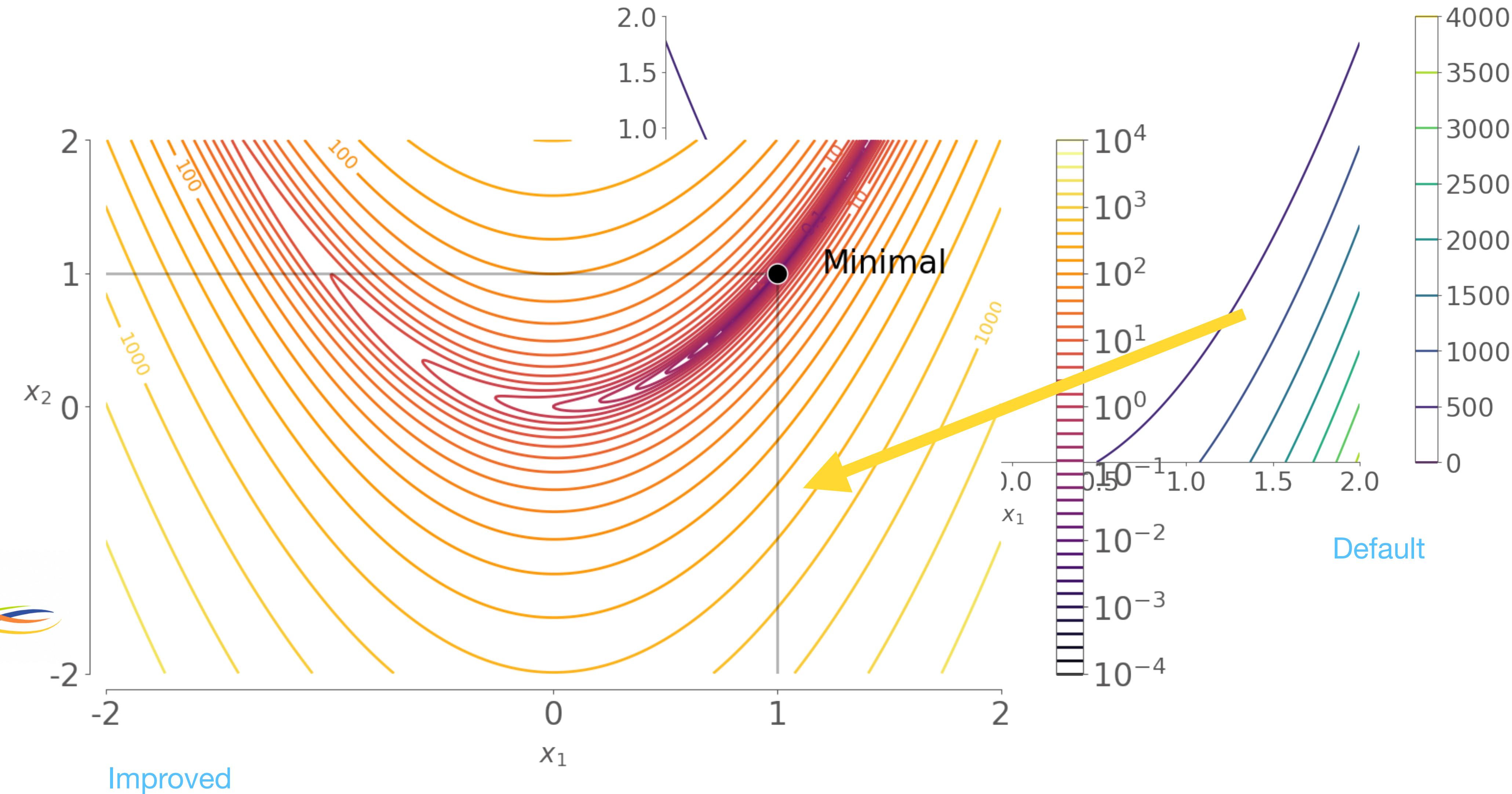


# Generating nice figures for your research

**Sicheng He**  
PostDoc Associate

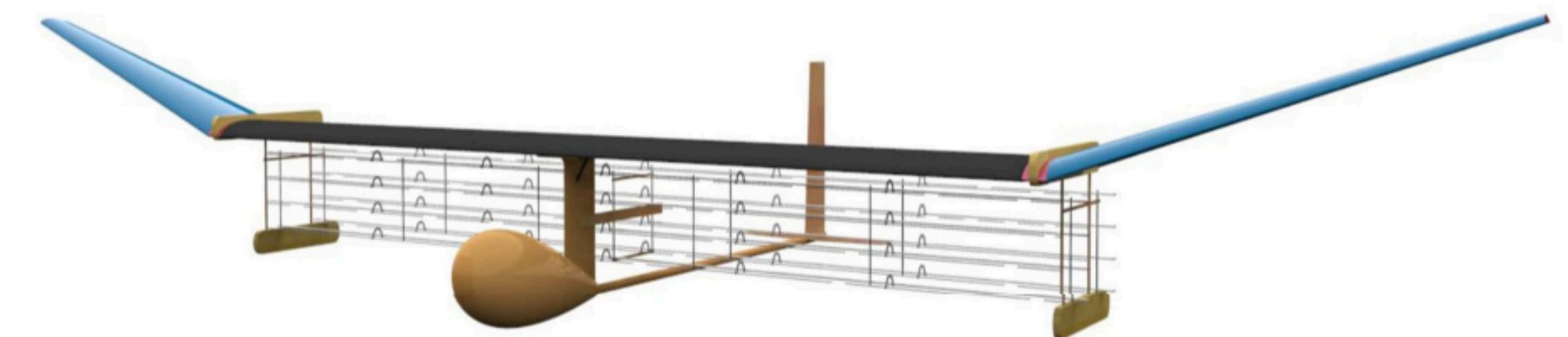
Knowledge and  
Expertise Sharing Seminar  
Oct 14, 2022  
LAE, MIT



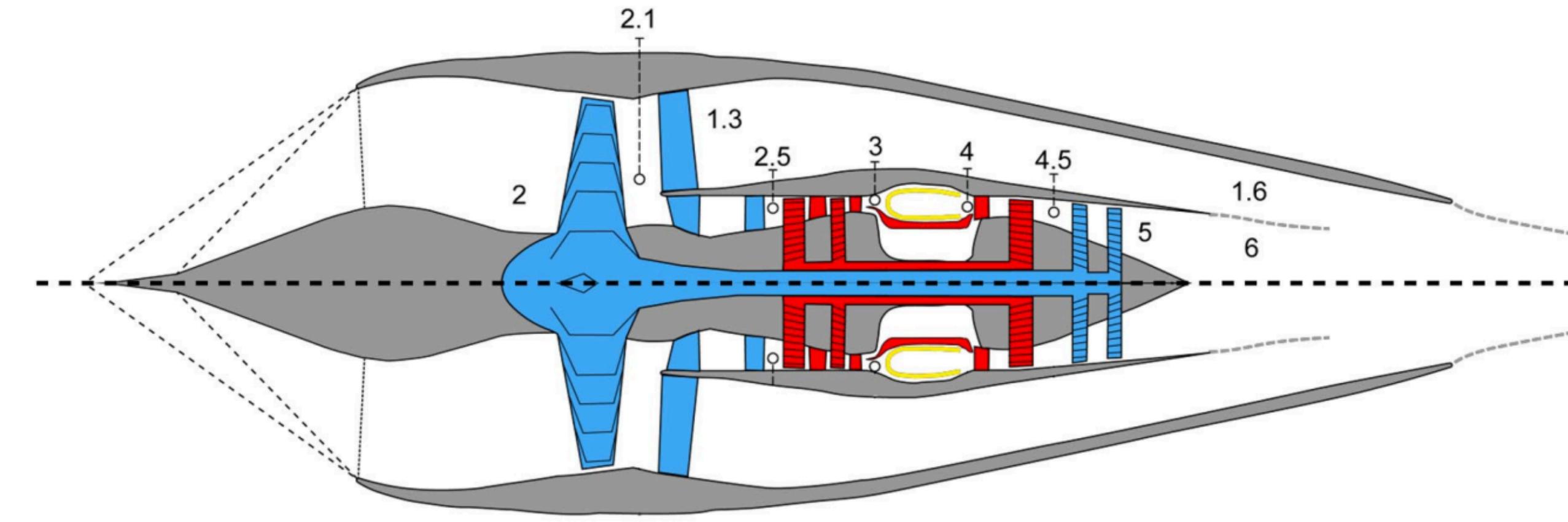
# Why does it matter?

- It reduces effort for people to understand your work.
- It forces you to think clearly and do better research.

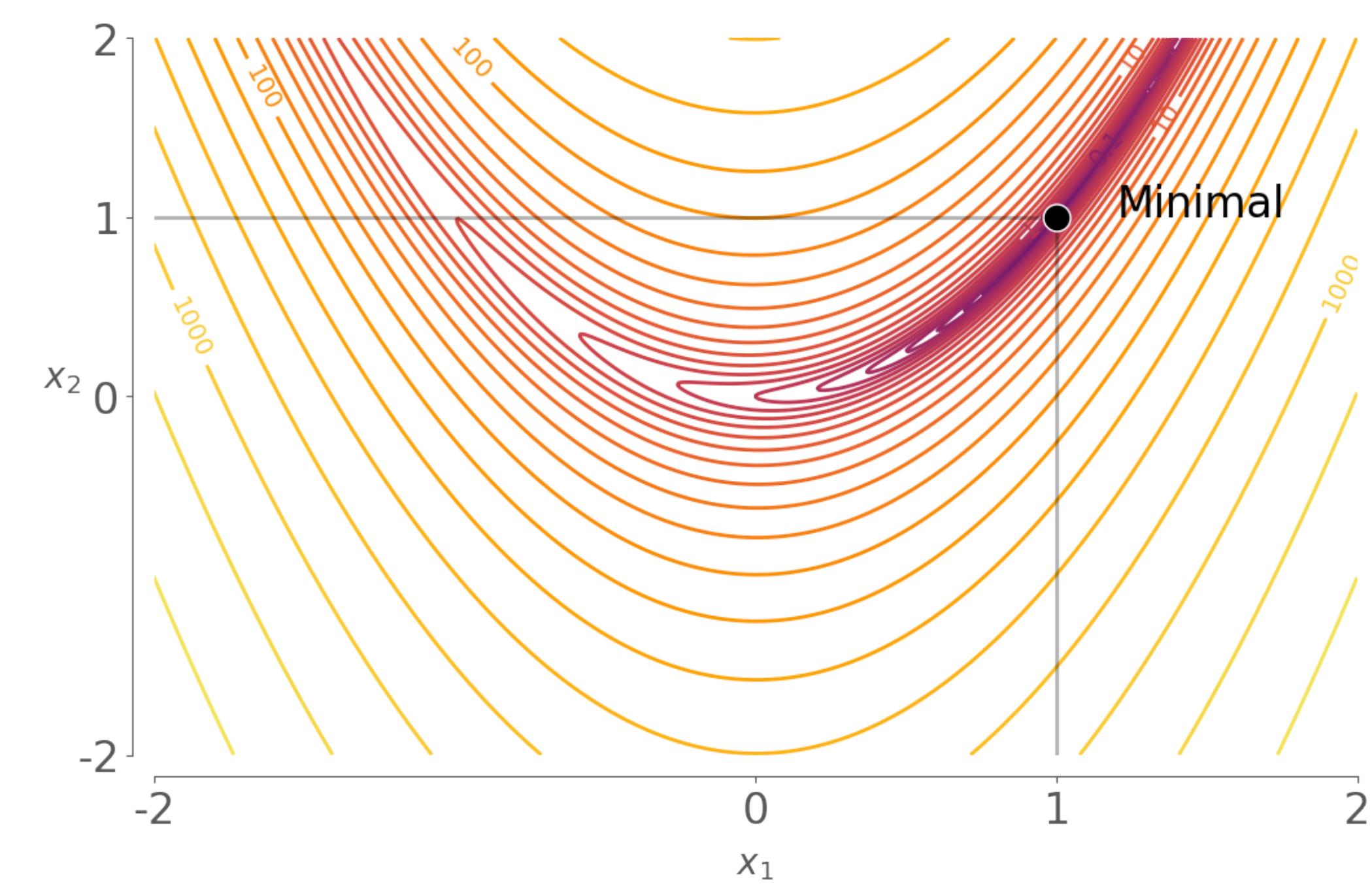
# Different types of figures ...



Xu et al. Nature 2018

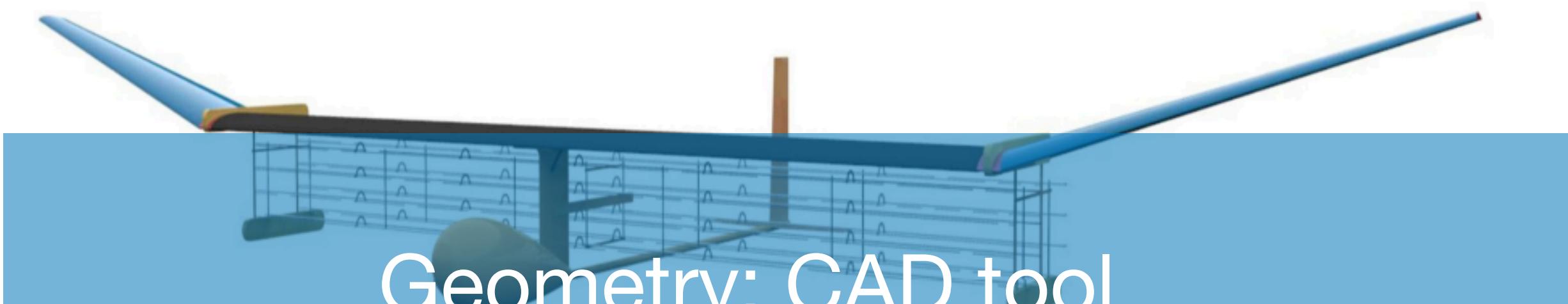


Voet et al. Scitech 2021



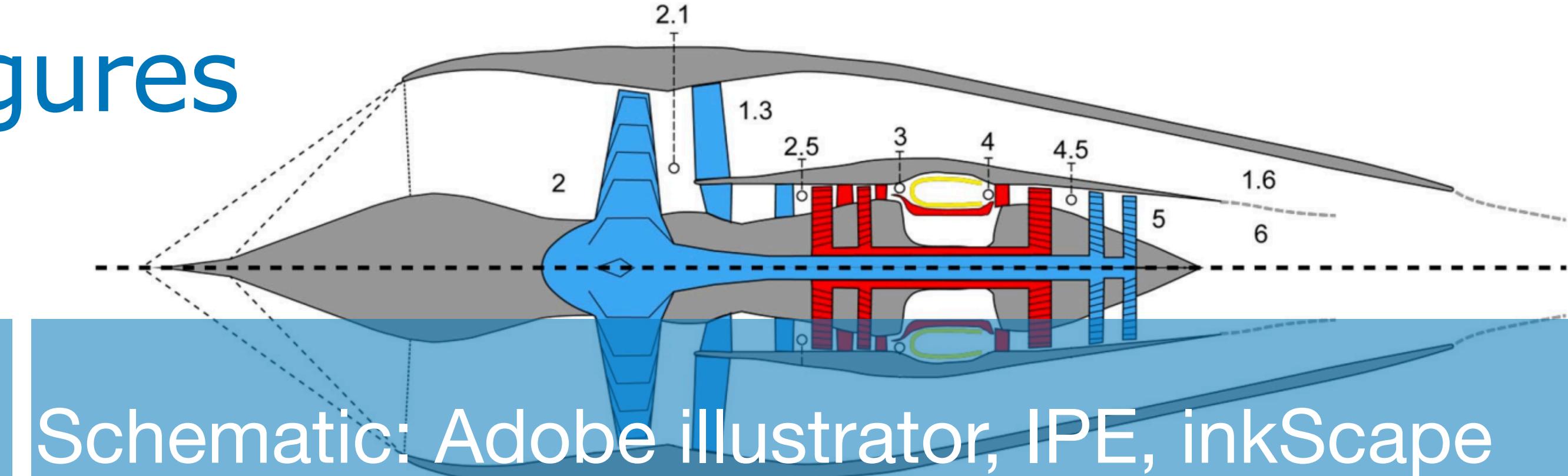
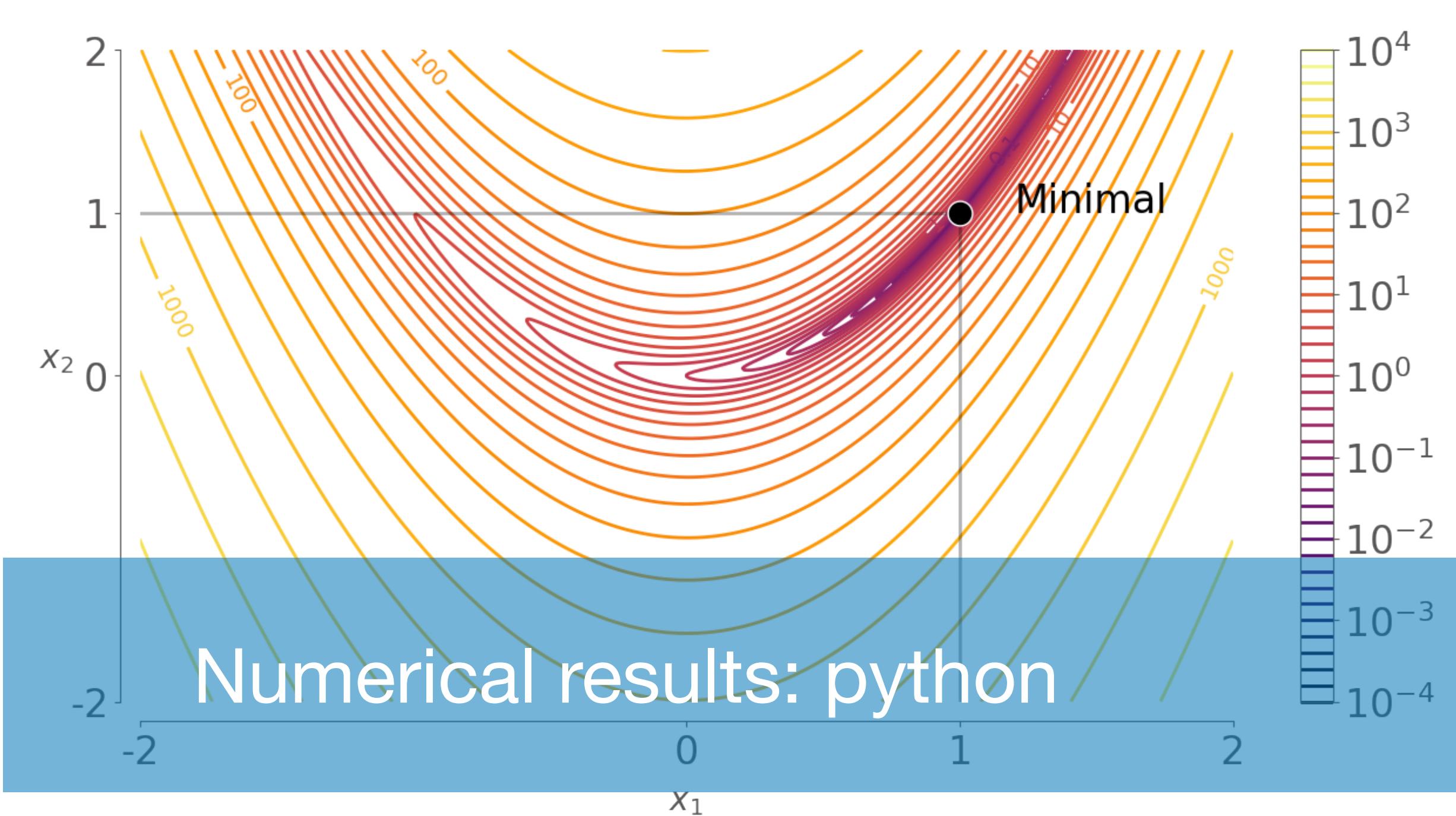
Meijer et al. Environmental Research Letters 2022

# Different tools for different figures



Geometry: CAD tool

Xu et al. Nature 2018



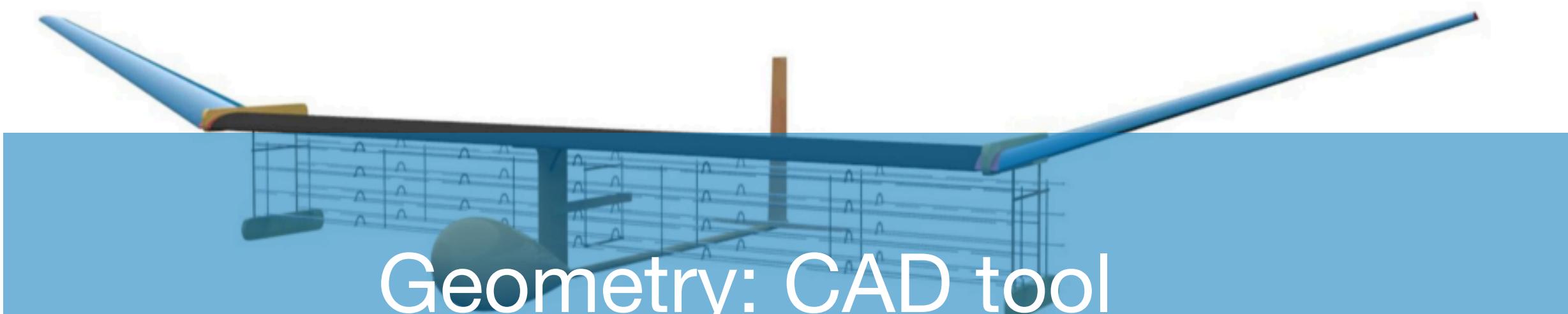
Schematic: Adobe illustrator, IPE, inkscape

Voet et al. Scitech 2021



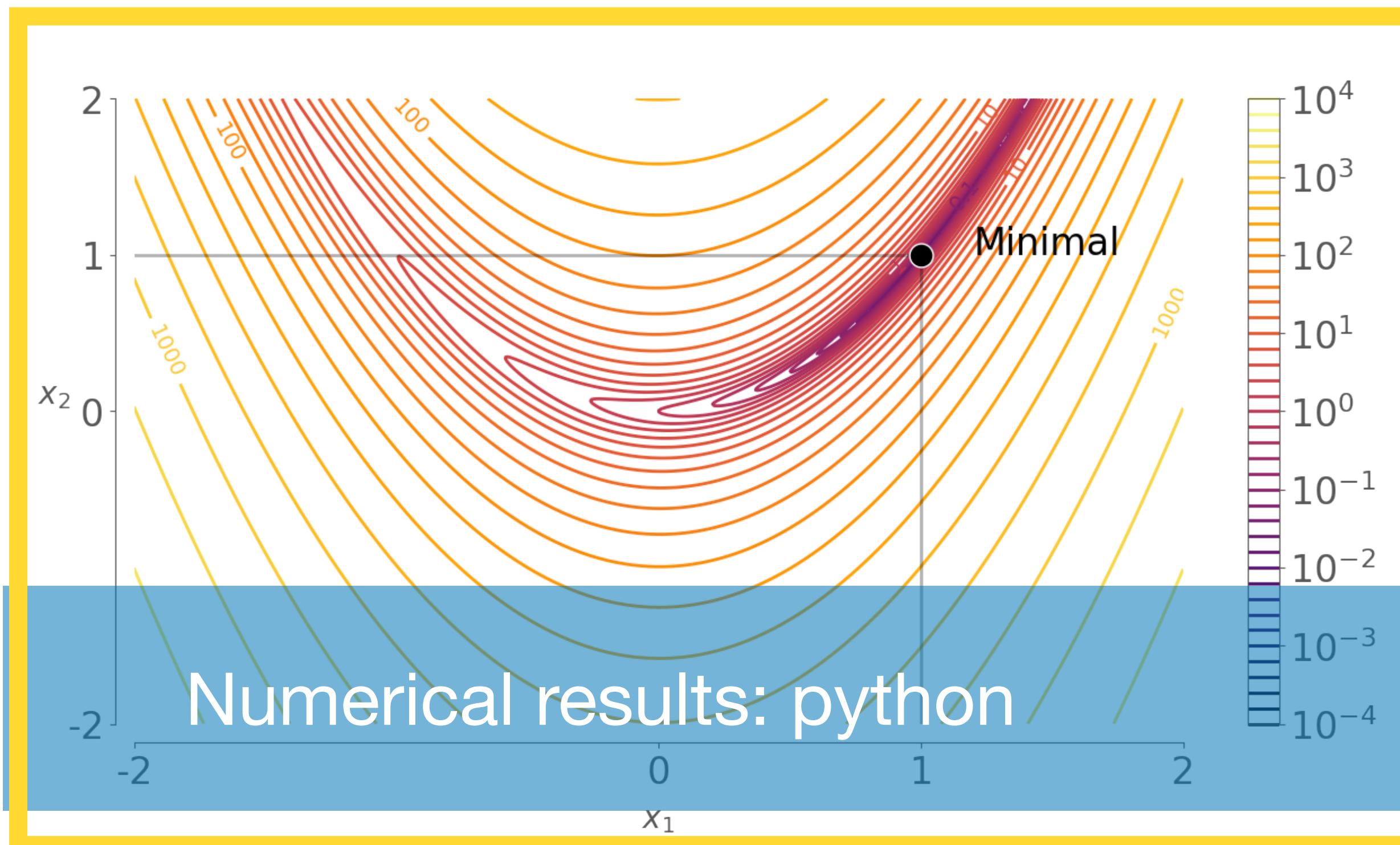
Specific numerical results: python packages

# Different tools for different figures



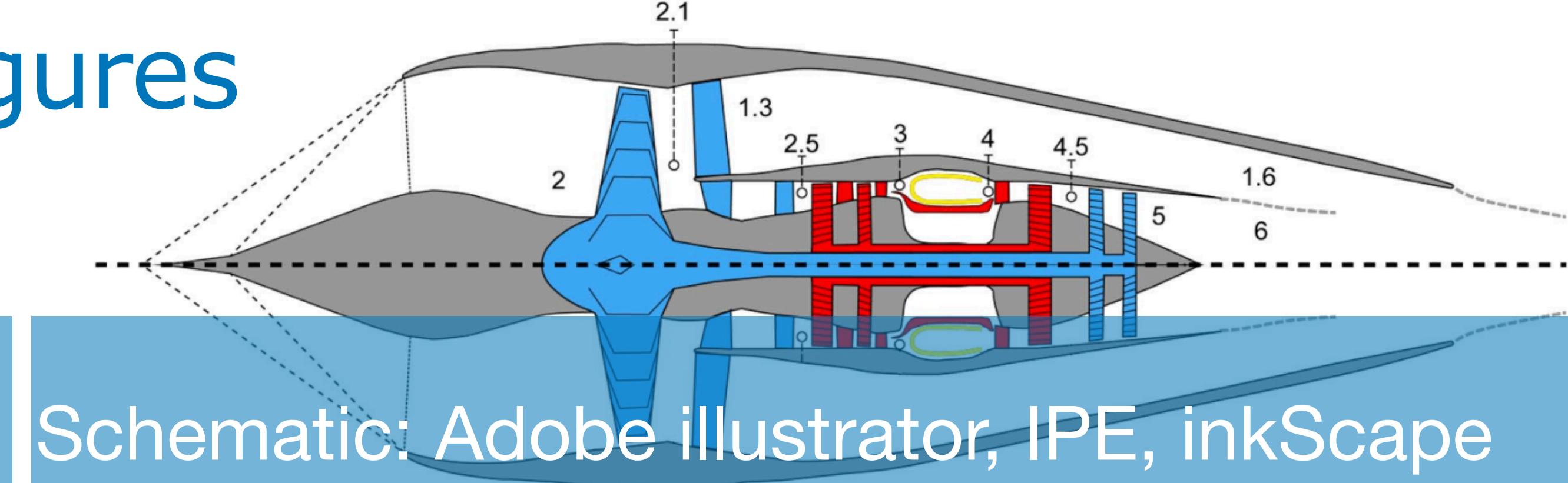
Geometry: CAD tool

Xu et al. Nature 2018



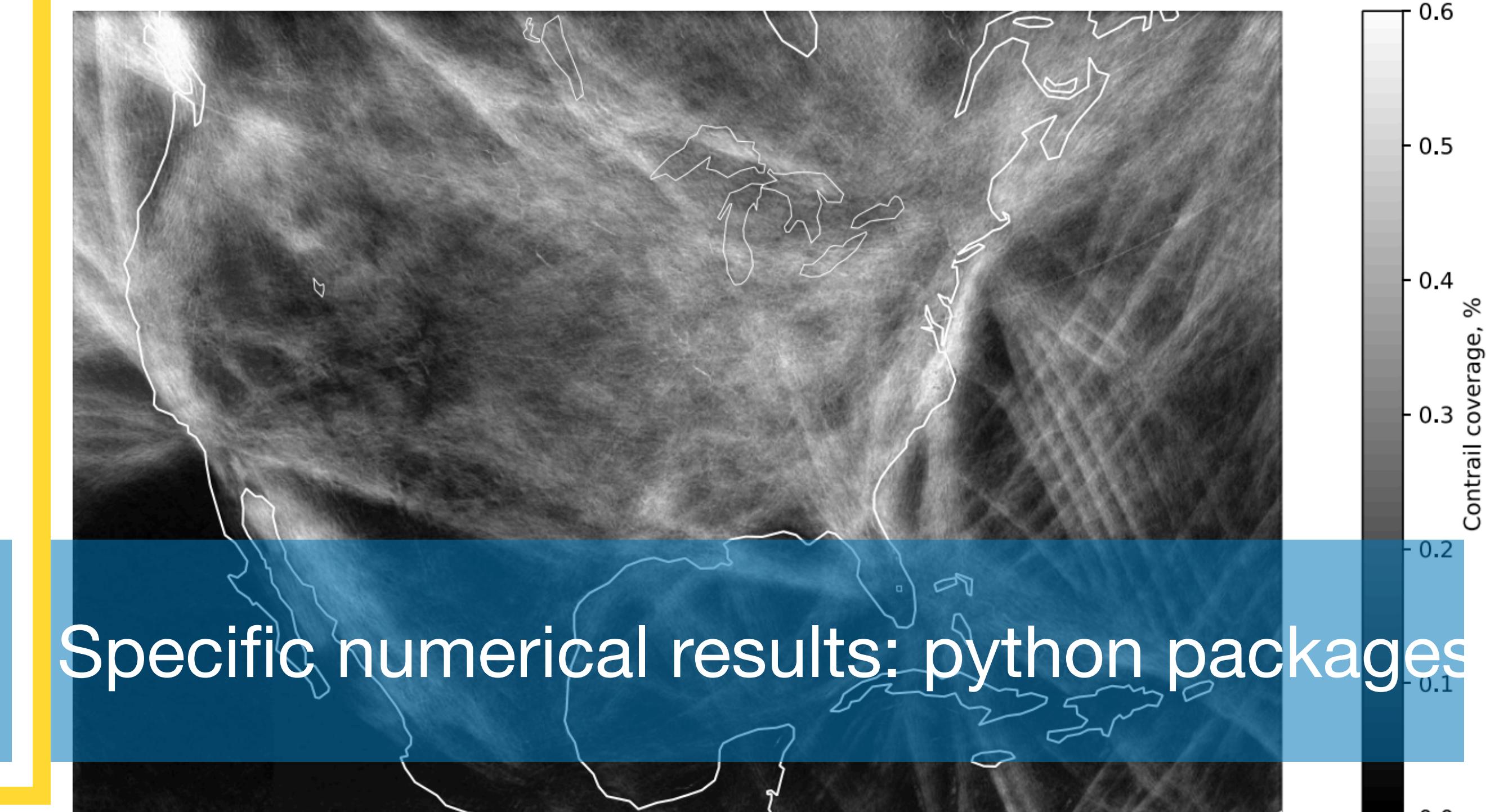
Numerical results: python

Focus of today



Schematic: Adobe illustrator, IPE, inkscape

Voet et al. Scitech 2021



Specific numerical results: python packages

Meijer et al. Environmental Research Letters 2022

# A promise ...

You will learn how to generate publication quality  
curve and contour plot using python and matplotlib.

# Outline

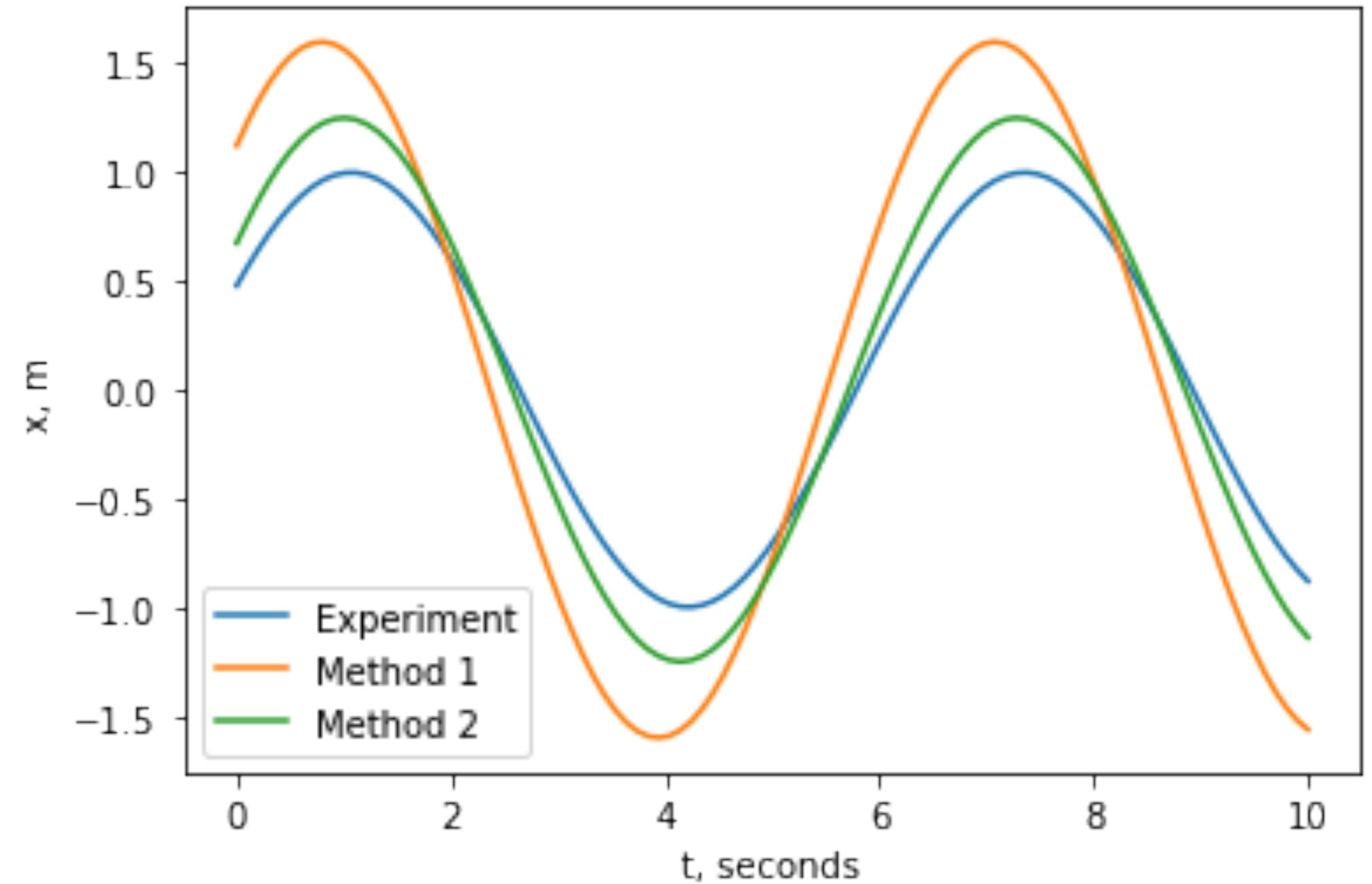
- ▶ Module 1: What and how?
- ▶ Module 2: Try it yourself
- ▶ Module 3: Conclusion

# Module 1: What and how?

## Define nice plots ...

- Accurately convey  $n$  pieces of critical information (usually  $1 \leq n \leq 3$ ).
- High information / noise ratio.
- Good style.

# Case study 1: Curve plot

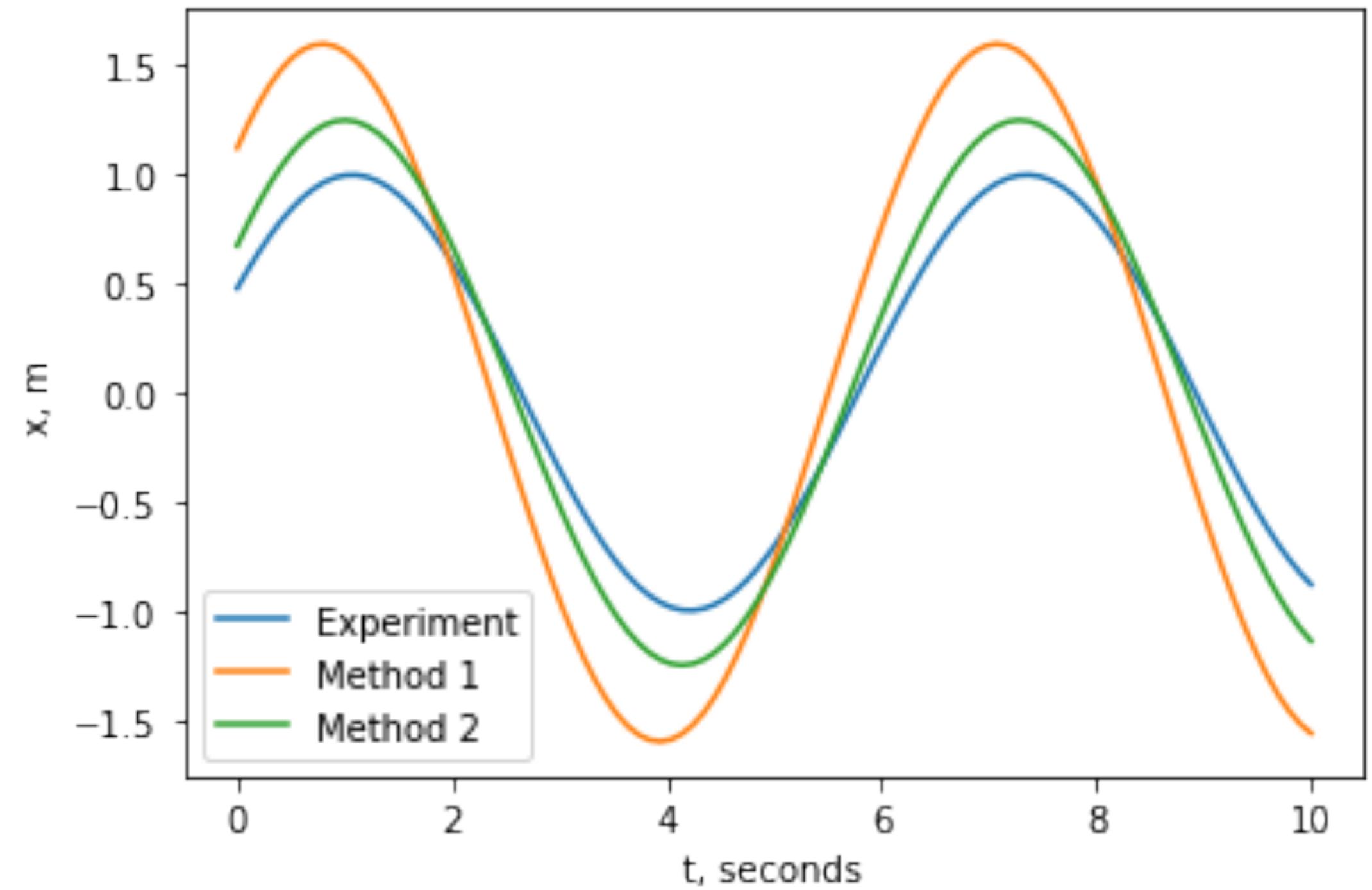


Message:

Method 2 outperforms Method 1 because it is closer to the experimental data.

```
1 # Initialize the figure
2 fig, ax = plt.subplots()
3
4 # Lines and legends
5 ax.plot(x, yref, label = "Experiment")
6 ax.plot(x, y1, label = "Method 1")
7 ax.plot(x, y2, label = "Method 2")
8
9 # x, y labels
10 ax.set_ylabel('x, m')
11 ax.set_xlabel('t, seconds')
12
13 # plot
14 ax.legend()
15 plt.show()
```

# What are the issues?



Message:

Method 2 outperforms Method 1 because it is closer to the experimental data.

```
1 # Initialize the figure
2 fig, ax = plt.subplots()
3
4 # Lines and legends
5 ax.plot(x, yref, label = "Experiment")
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8
9 # x, y labels
10 ax.set_ylabel('x, m')
11 ax.set_xlabel('t, seconds')
12
13 # plot
14 ax.legend()
15 plt.show()
```

# An algorithm

---

**Algorithm 1.1** Figure improving algorithm

---

**Input** Message set, original figure

**for** message ∈ message set **do**

**if** message unclear **then**

Fix message.

**end if**

**end for**

set = {labels, legends, boxline, words, lines, ticks, colorbar, ...}

**for** feature ∈ set **do**

**if** feature unclear **then**

Fix feature.

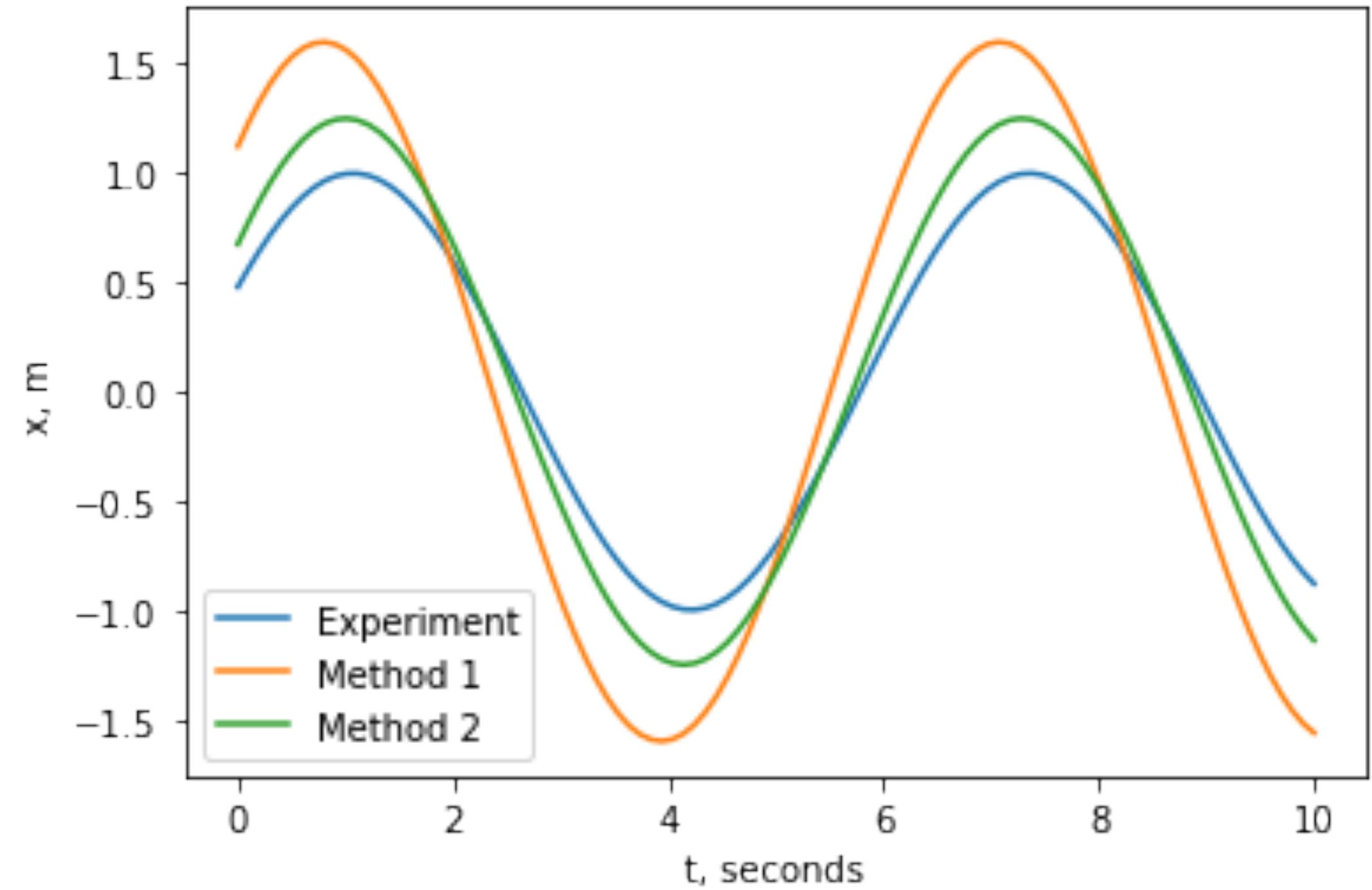
**end if**

**end for**

**Return** Improved figure

---

# What are the issues?



---

**Algorithm 1.1** Figure improving algorithm

---

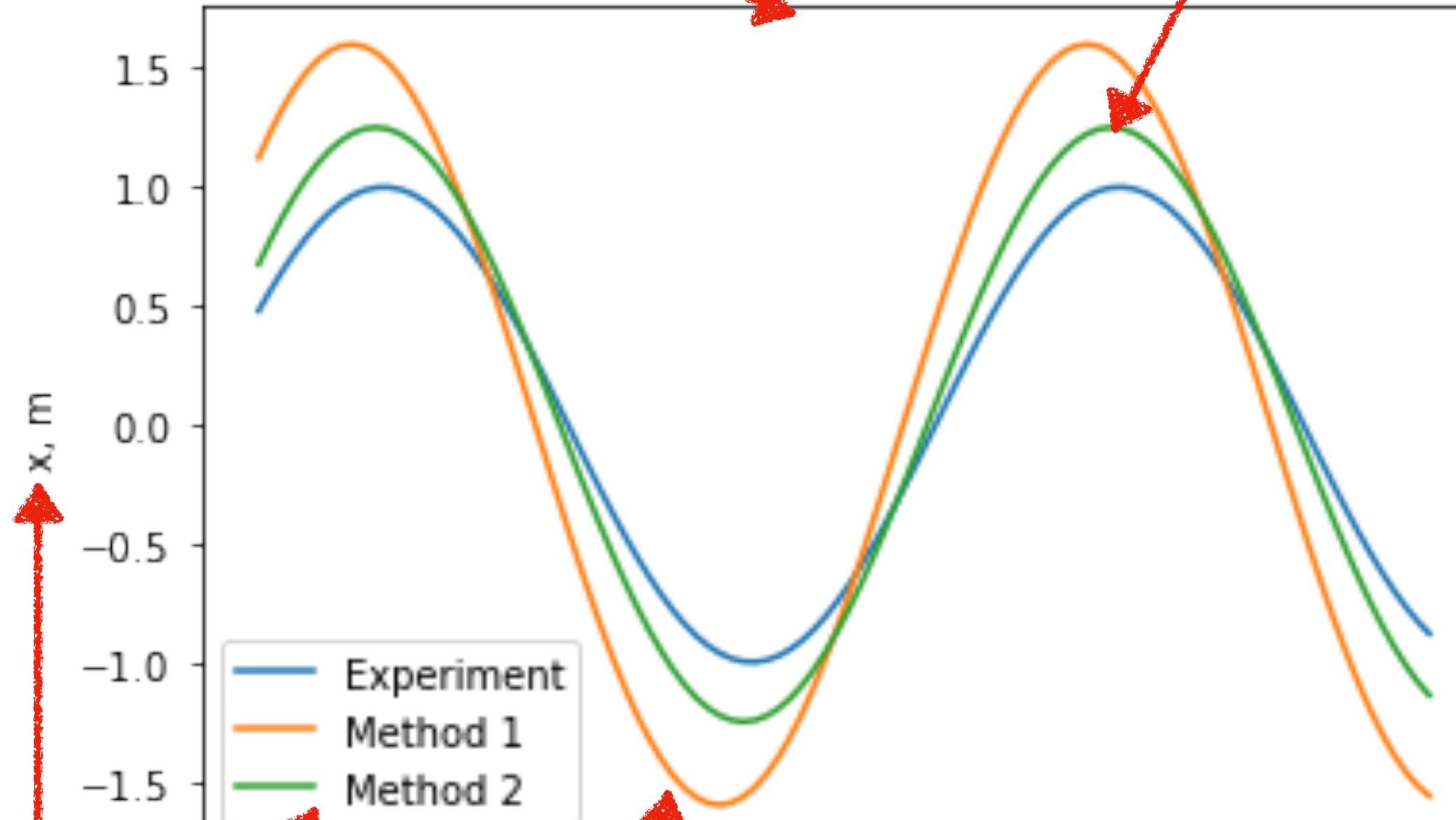
```
Input Message set, original figure
for message ∈ message set do
    if message unclear then
        Fix message.
    end if
end for
set = {labels, legends, boxline, words, lines, ticks, colorbar, ...}
for feature ∈ set do
    if feature unclear then
        Fix feature.
    end if
end for
Return Improved figure
```

---

# What are the issues?

1. OK, 1 is better than 2, but by how much?

7. Do we need to box it? Noise!



4. Shall we rotate our head to read?

2. Did we do experiments for the whole curve?

3. Constantly switch back and forth  
from the legends to the curves?

6. Math symbol, use Latex!

5. Do we need so many ticks points? Noise!

---

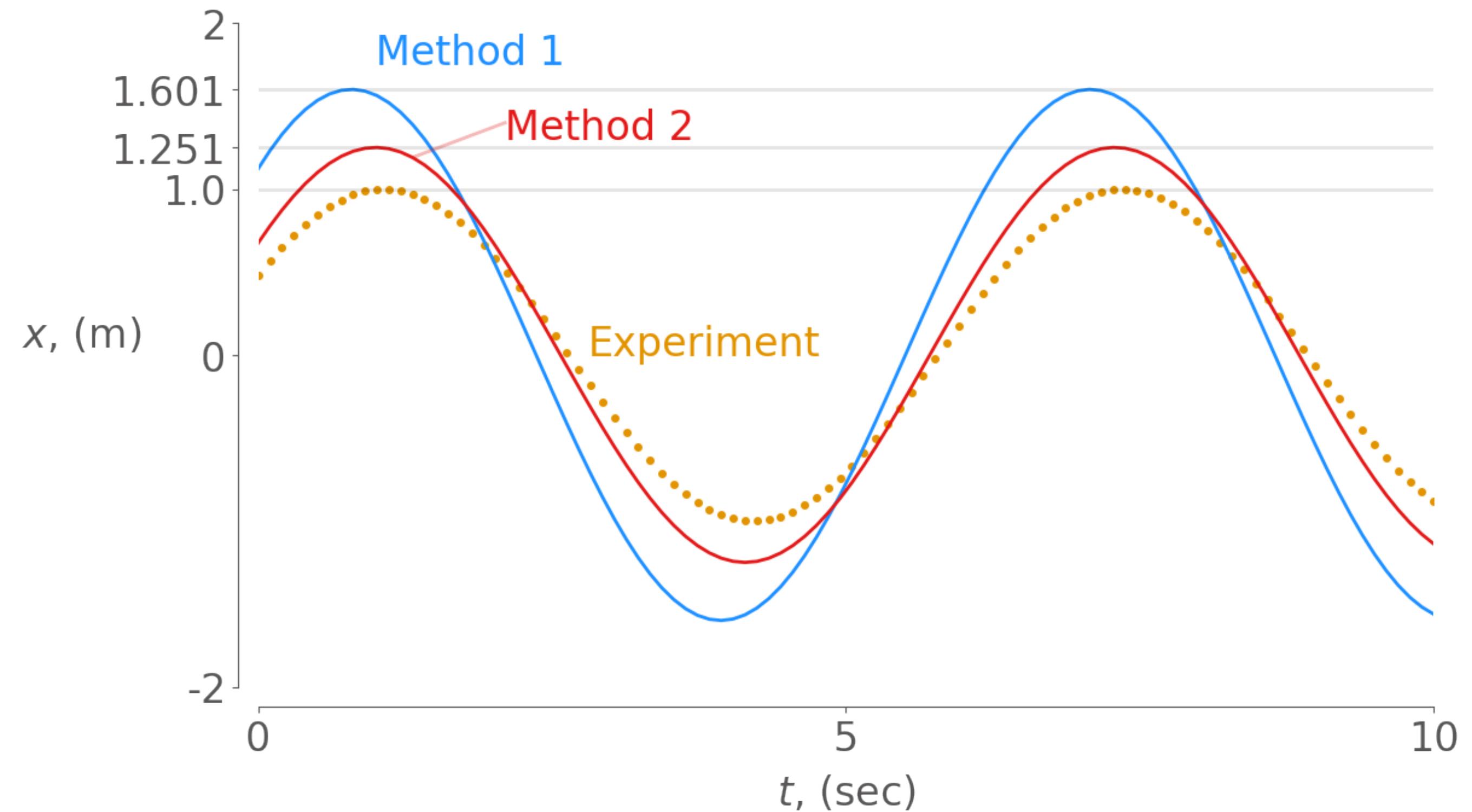
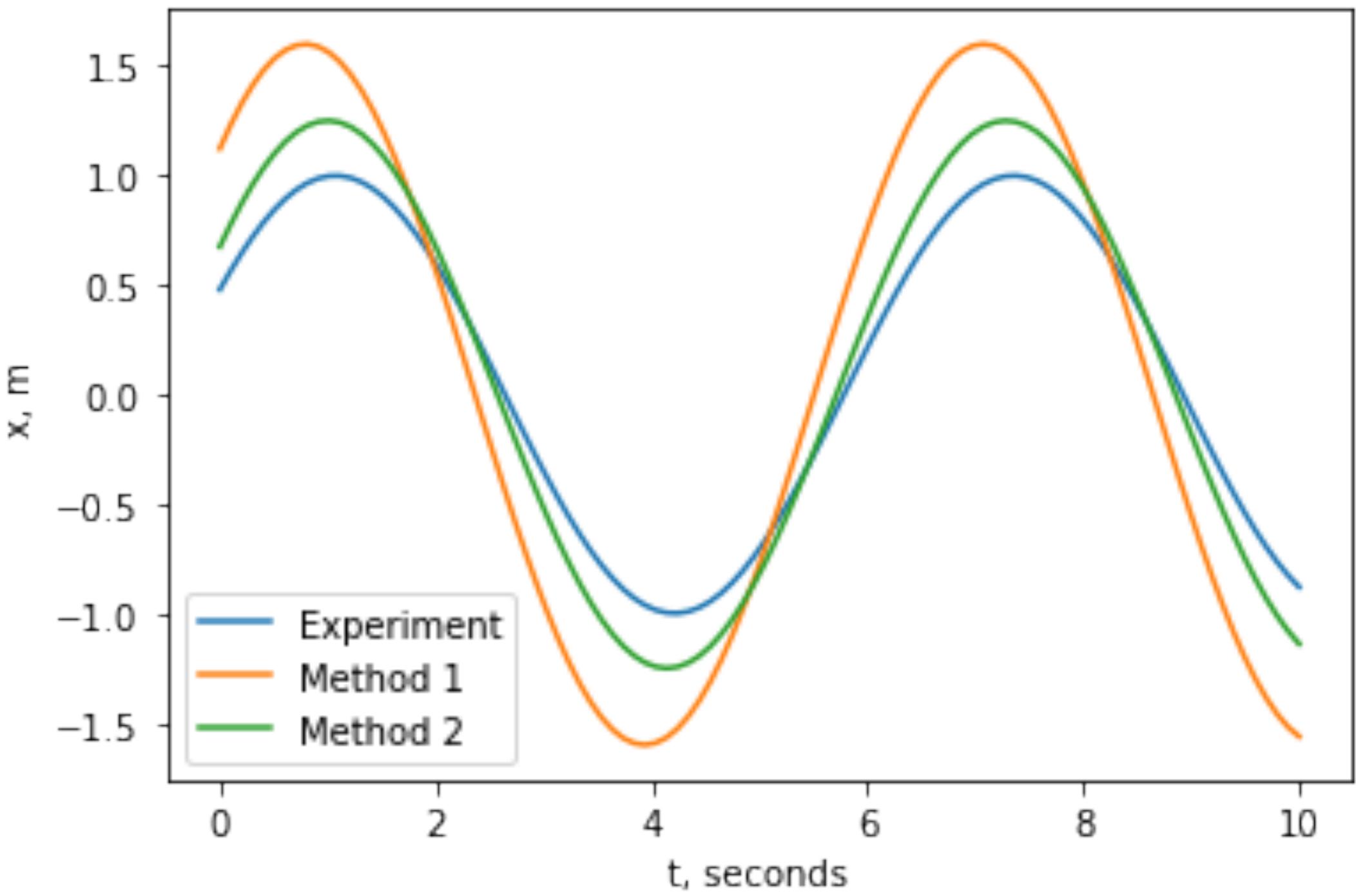
**Algorithm 1.1** Figure improving algorithm

---

```
Input Message set, original figure
for message ∈ message set do
    if message unclear then
        Fix message.
    end if
end for
set = {labels, legends, boxline, words, lines, ticks, colorbar, ...}
for feature ∈ set do
    if feature unclear then
        Fix feature.
    end if
end for
Return Improved figure
```

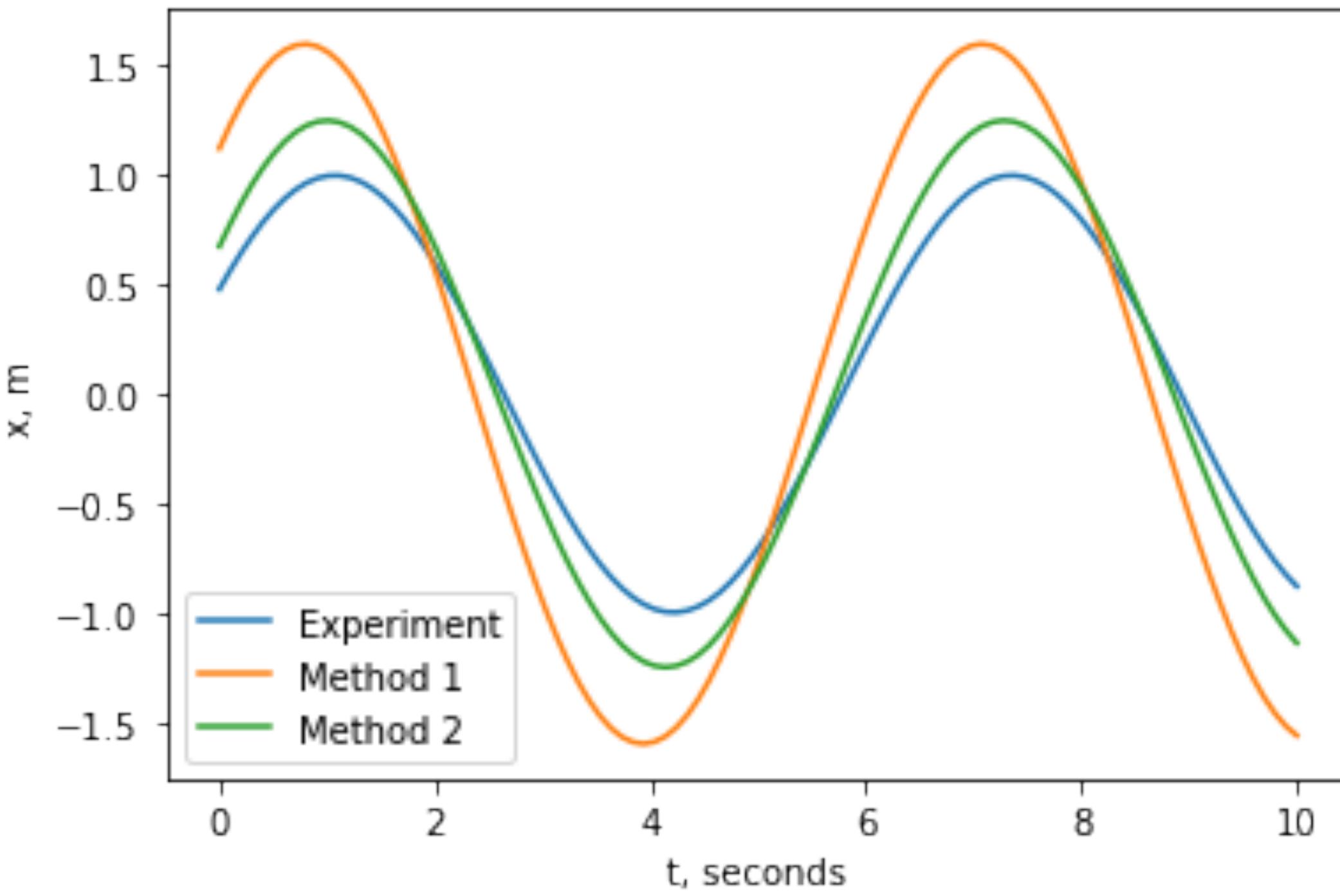
---

# Improved figure



1. OK, 1 is better than 2, but by how much?
2. Did we do experiments for the whole curve?
3. Constantly switch back and forth from the legends to the curves?
4. Shall we rotate our head to read?
5. Do we need so many ticks points? Noise!
6. Math symbol, use Latex!
7. Do we need to box it? Noise!

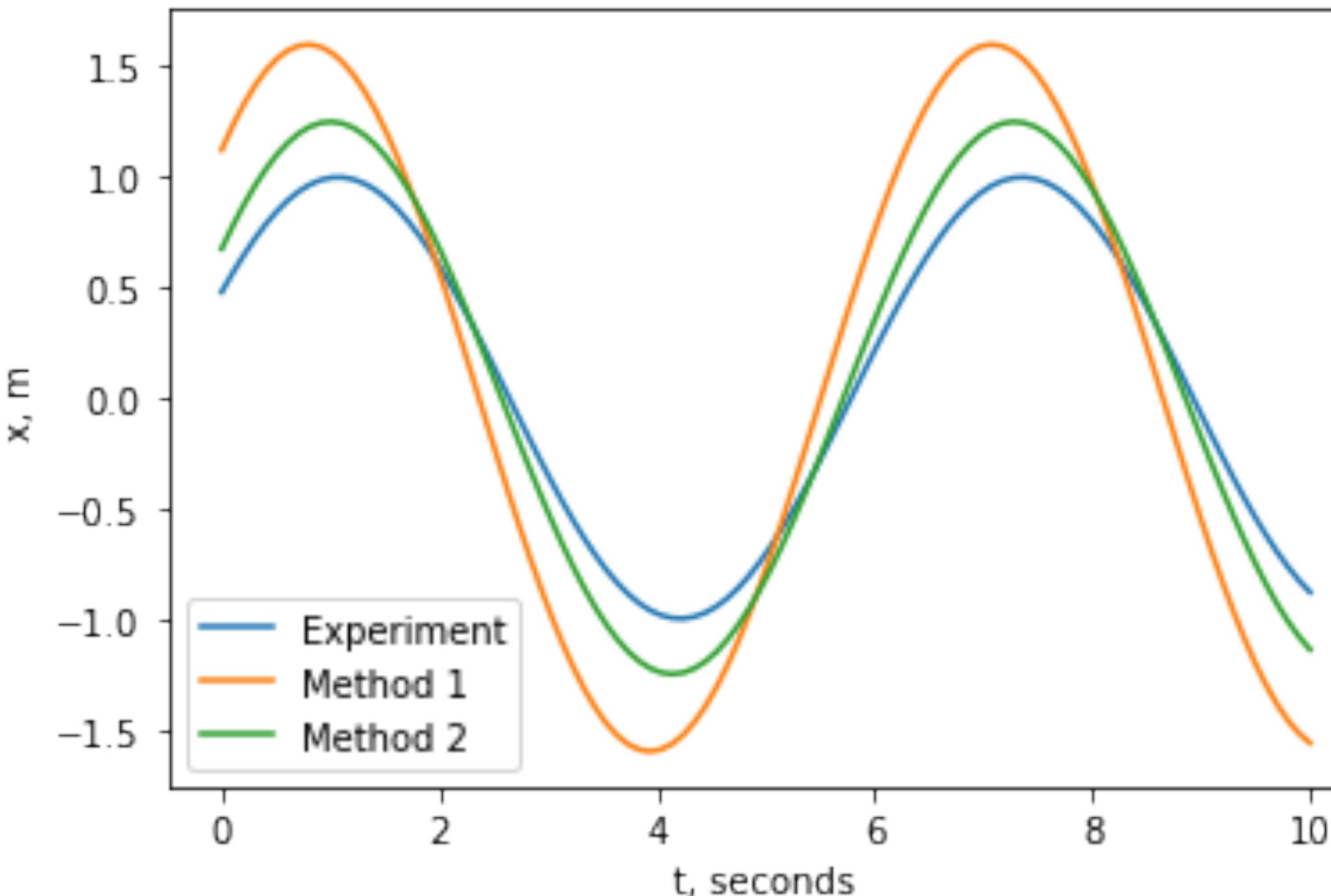
# How to do it?



1. OK, 1 is better than 2, but by how much?
2. Did we do experiments for the whole curve?
3. Constantly switch back and forth from the legends to the curves?
4. Shall we rotate our head to read?
5. Do we need so many ticks points? Noise!
6. Math symbol, use Latex!
7. Do we need to box it? Noise!

```
1 import niceplots
2
3 fig, ax = plt.subplots()
4
5 colors = plt.rcParams["axes.prop_cycle"].by_key()["color"]
6 print("colors", colors)
7
8 # Applying niceplots standard
9 # Fix (I7) and in general makes the figure looks better
10 niceplots.setRCParams()
11 niceplots.All()
12 # If not using niceplots
13 # ax.spines['top'].set_visible(False)
14 # ax.spines['right'].set_visible(False)
15
16 # Fix (I2)
17 ax.plot(x, yref, 'o')
18 ax.plot(x, y1)
19 ax.plot(x, y2)
20
21 # Fix (I4, I6)
22 ax.set_ylabel(r'$x$', (m)', rotation = 0)
23 ax.yaxis.set_label_coords(-.15, .5)
24 ax.set_xlabel(r'$t$', (sec)')
```

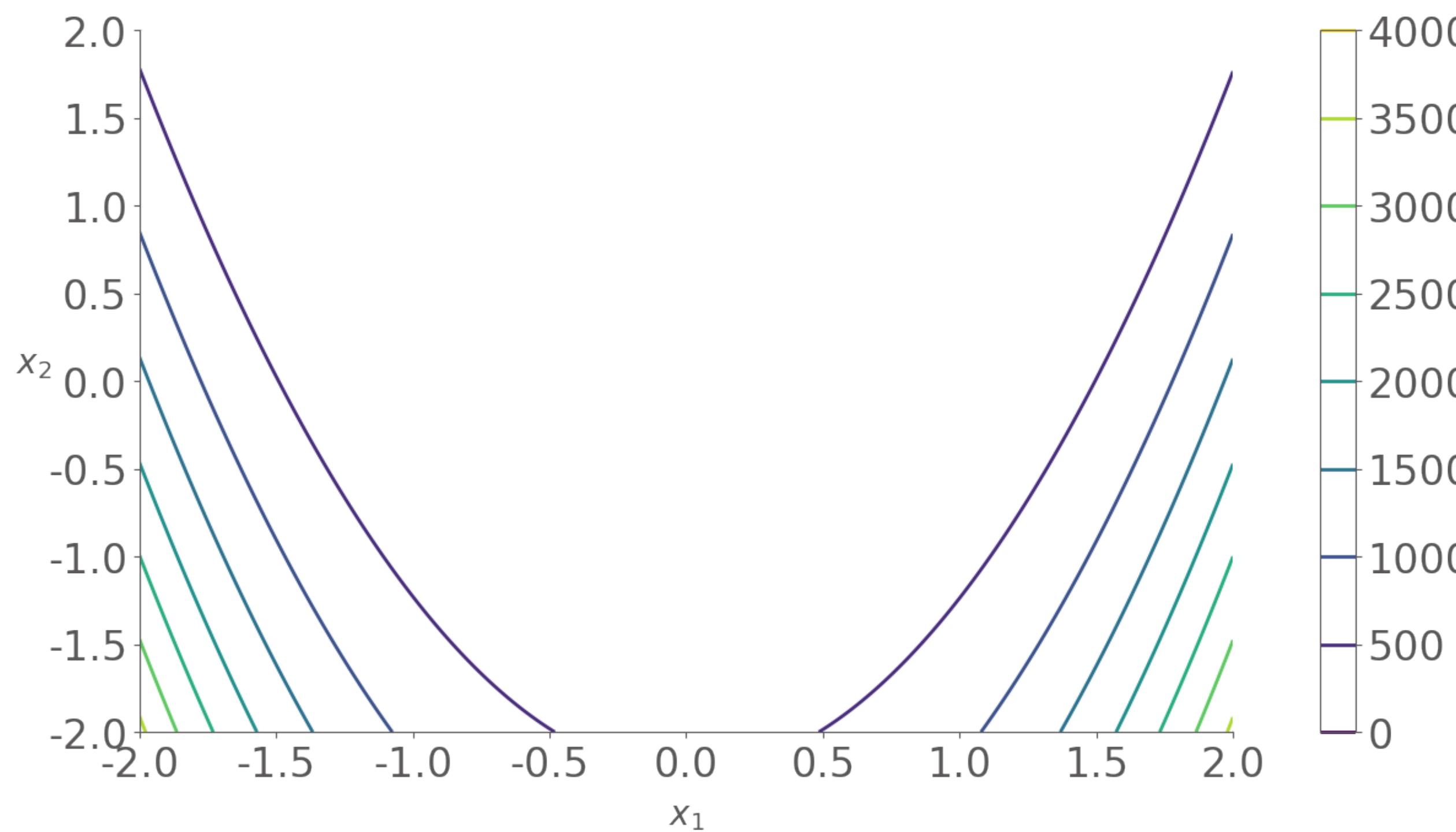
# How to do it?



1. OK, 1 is better than 2, but by how much?
2. Did we do experiments for the whole curve?
3. Constantly switch back and forth from the legends to the curves?
4. Shall we rotate our head to read?
5. Do we need so many ticks points? Noise!
6. Math symbol, use Latex!
7. Do we need to box it? Noise!

```
26 # Fix (I3)
27 ax.text(2.8, 0, "Experiment", color = colors[0])
28 ax.text(1, 1.75, "Method 1", color = colors[1])
29 ax.text(2.1, 1.3, "Method 2", color = colors[2])
30 ax.plot([x[13], 2.1], [y2[13], 1.4], '--', color = colors[2], alpha = 0.3)
31
32 # Fix (I1)
33 ind_max_ref = np.argmax(yref)
34 ind_max_1 = np.argmax(y1)
35 ind_max_2 = np.argmax(y2)
36 ax.plot([0, 10], [yref[ind_max_ref], yref[ind_max_ref]], color = 'k', alpha =
37 ax.plot([0, 10], [y1[ind_max_1], y1[ind_max_1]], color = 'k', alpha = 0.1)
38 ax.plot([0, 10], [y2[ind_max_2], y2[ind_max_2]], color = 'k', alpha = 0.1)
39
40 # Fix (I5)
41 ax.set_xticks([0, 5, 10], [0, 5, 10])
42 ax.set_yticks([-2, 0, yref[ind_max_ref], y2[ind_max_2], y1[ind_max_1], 2],
43 [-2, 0, float('%.3g' % yref[ind_max_ref]), float('%.4g' % y2[ind_max_2]),
44 float('%.4g' % y1[ind_max_1]), 2])
45
46 # Fix (I8)
47 # plt.savefig('../curve.pdf', bbox_inches='tight')
```

# Case study 2: Contour plot



Background:

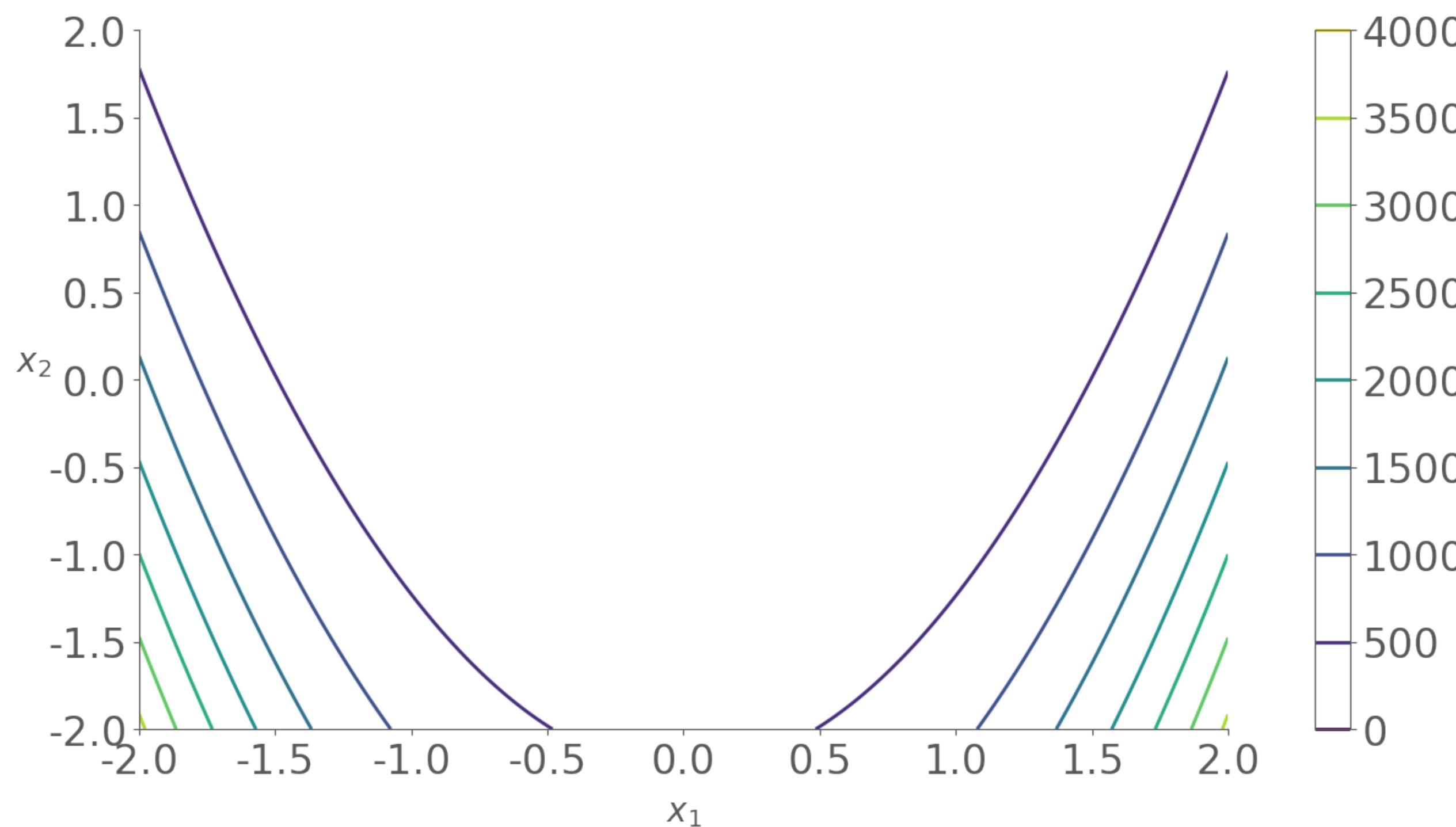
Rosenbrock plot widely used in optimization algorithm performance test. It contains multi-scaled data.

Message:

?

```
1 import matplotlib
2 import matplotlib.pyplot as plt
3 from matplotlib import ticker, cm
4
5 # Create the frame
6 fig, ax = plt.subplots()
7 X_arr, Y_arr = np.meshgrid(x_arr, y_arr)
8
9 # Plot
10 CP = ax.contour(X_arr, Y_arr, F_arr.T)
11
12 # Set axis
13 ax.set_xlabel(r'$x_1$', fontsize=20)
14 ax.set_ylabel(r'$x_2$', fontsize=20, rotation=0)
15
16 fig.colorbar(CP)
17
18 plt.show()
```

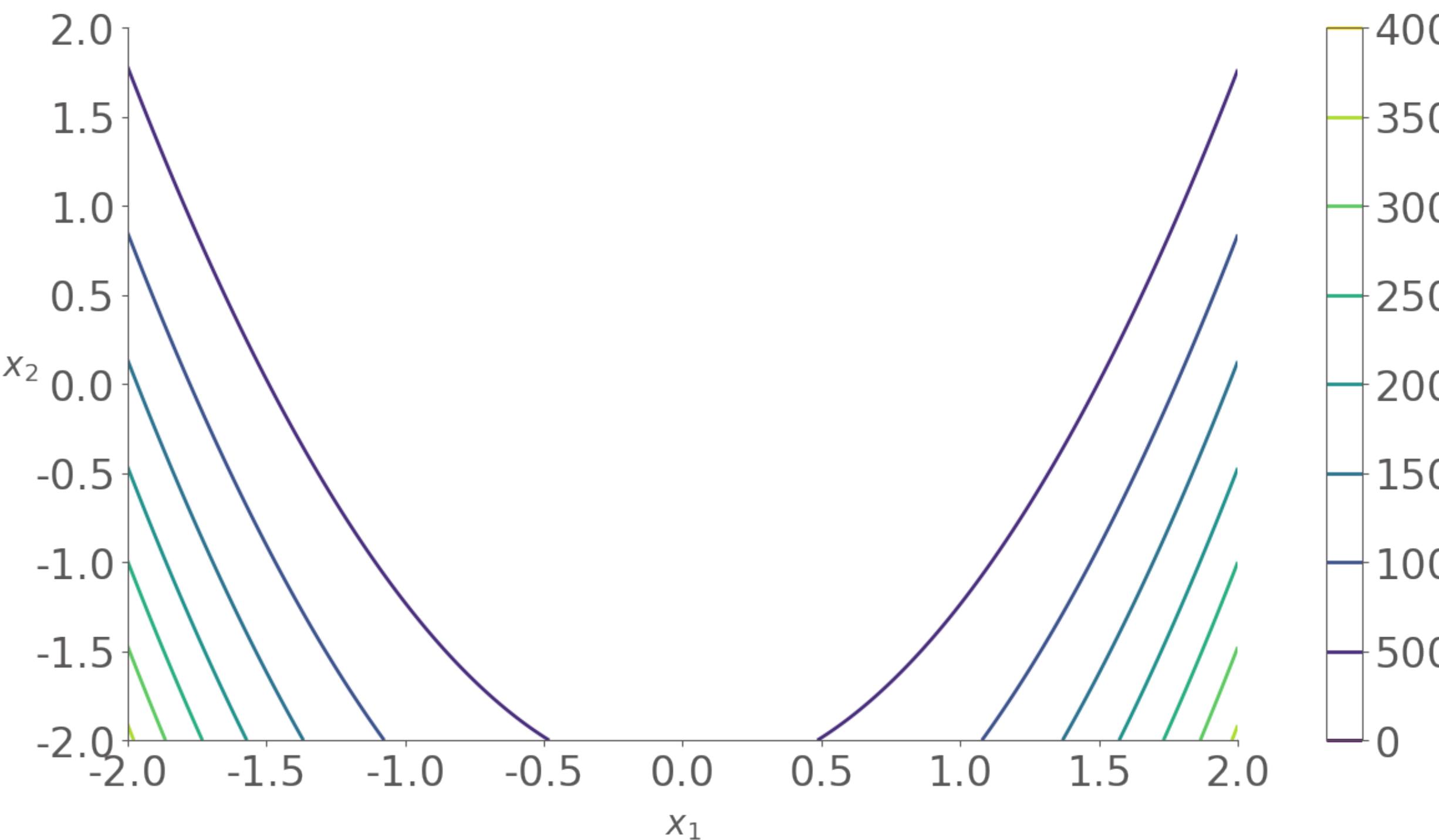
# What is the main message?



**Background:**  
Rosenbrock plot widely used in optimization algorithm performance test. It contains multi-scaled data.

**Message:**  
The optimal solution is at  $(1, 1)$ . And the trend of the contour (in log).

# What are the issues?



---

**Algorithm 1.1** Figure improving algorithm

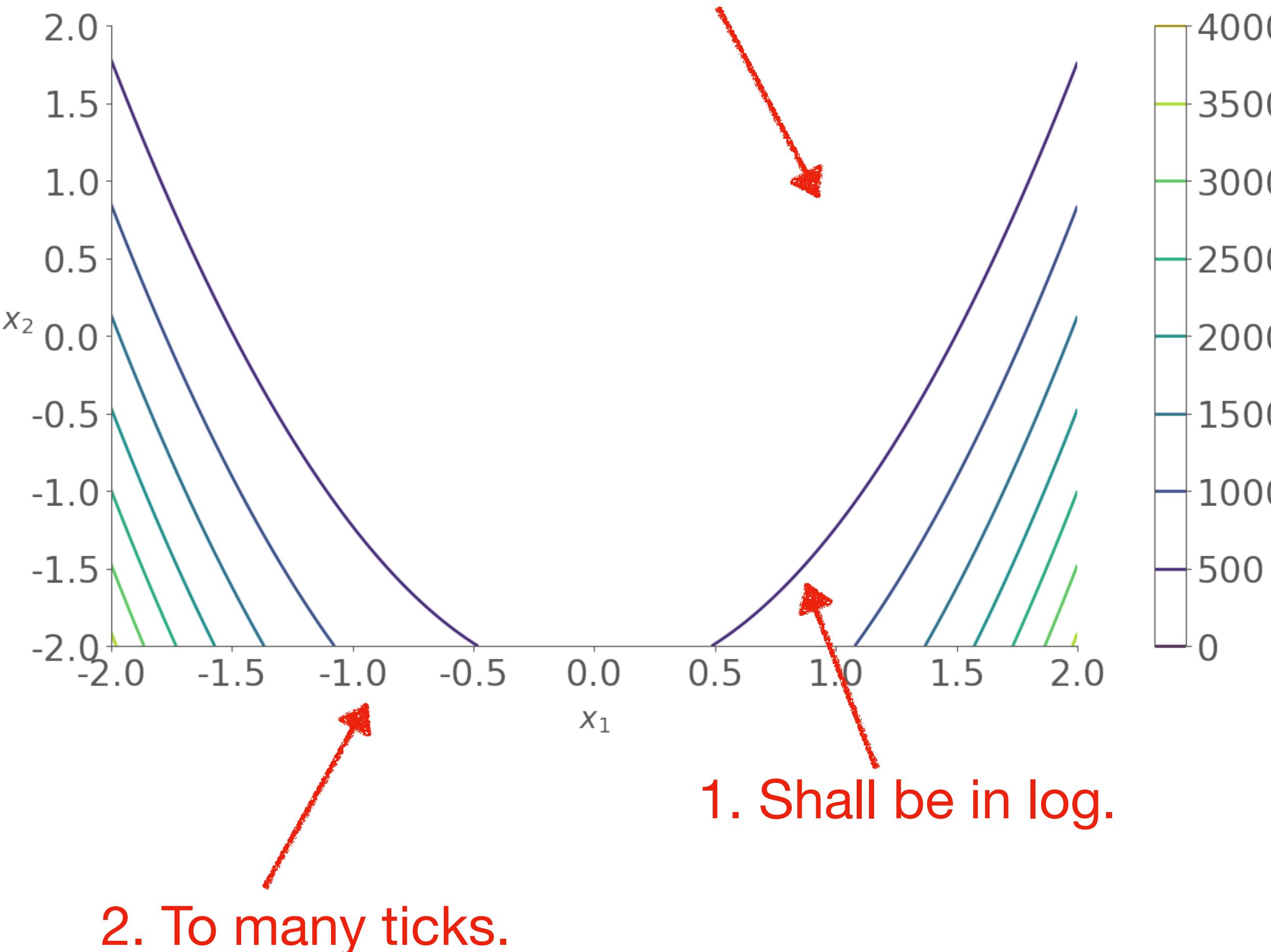
---

```
Input Message set, original figure
for message ∈ message set do
    if message unclear then
        Fix message.
    end if
end for
set = {labels, legends, boxline, words, lines, ticks, colorbar, ...}
for feature ∈ set do
    if feature unclear then
        Fix feature.
    end if
end for
Return Improved figure
```

---

# What are the issues?

## 3. Missing optimal solution.



---

**Algorithm 1.1** Figure improving algorithm

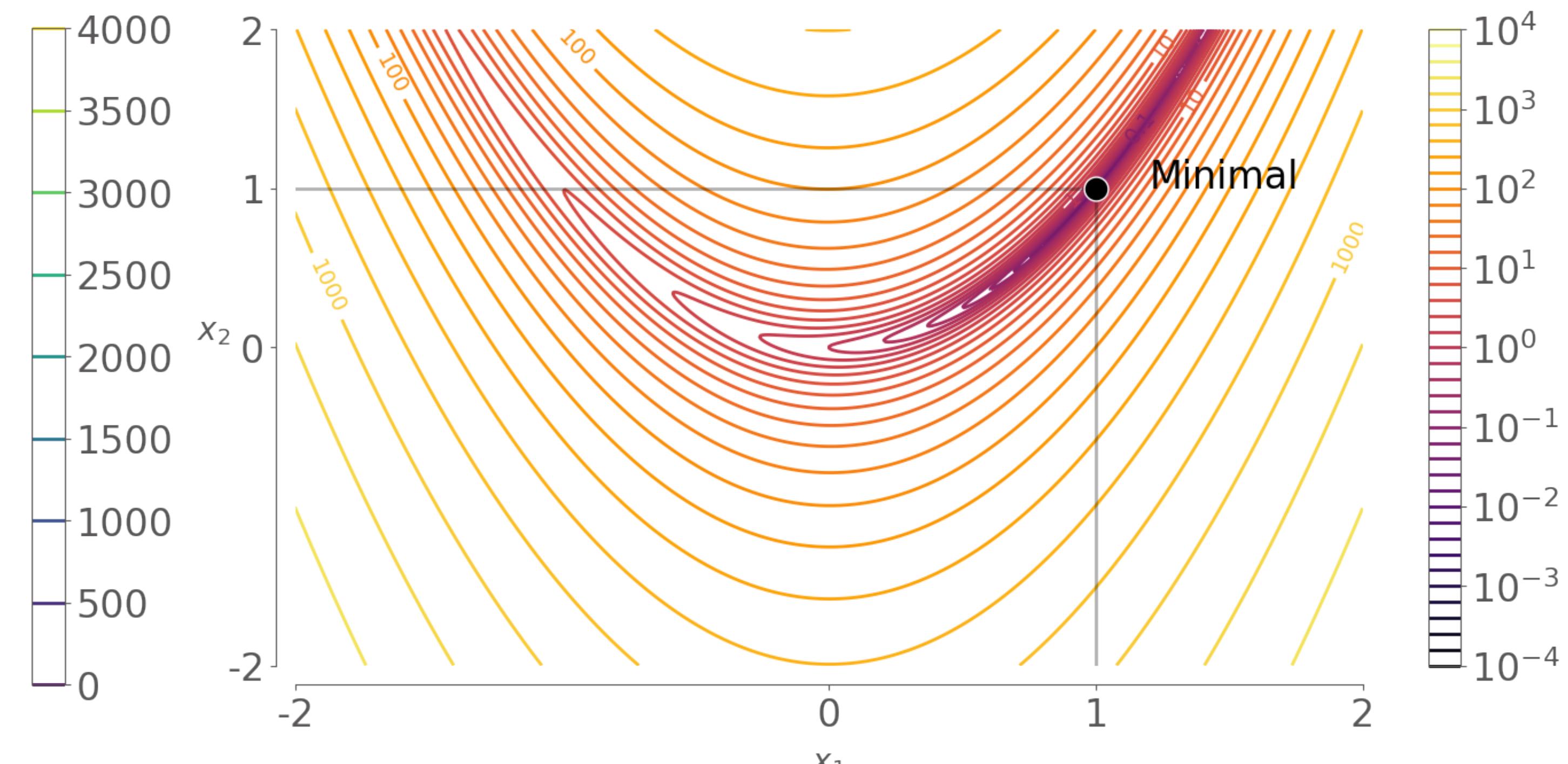
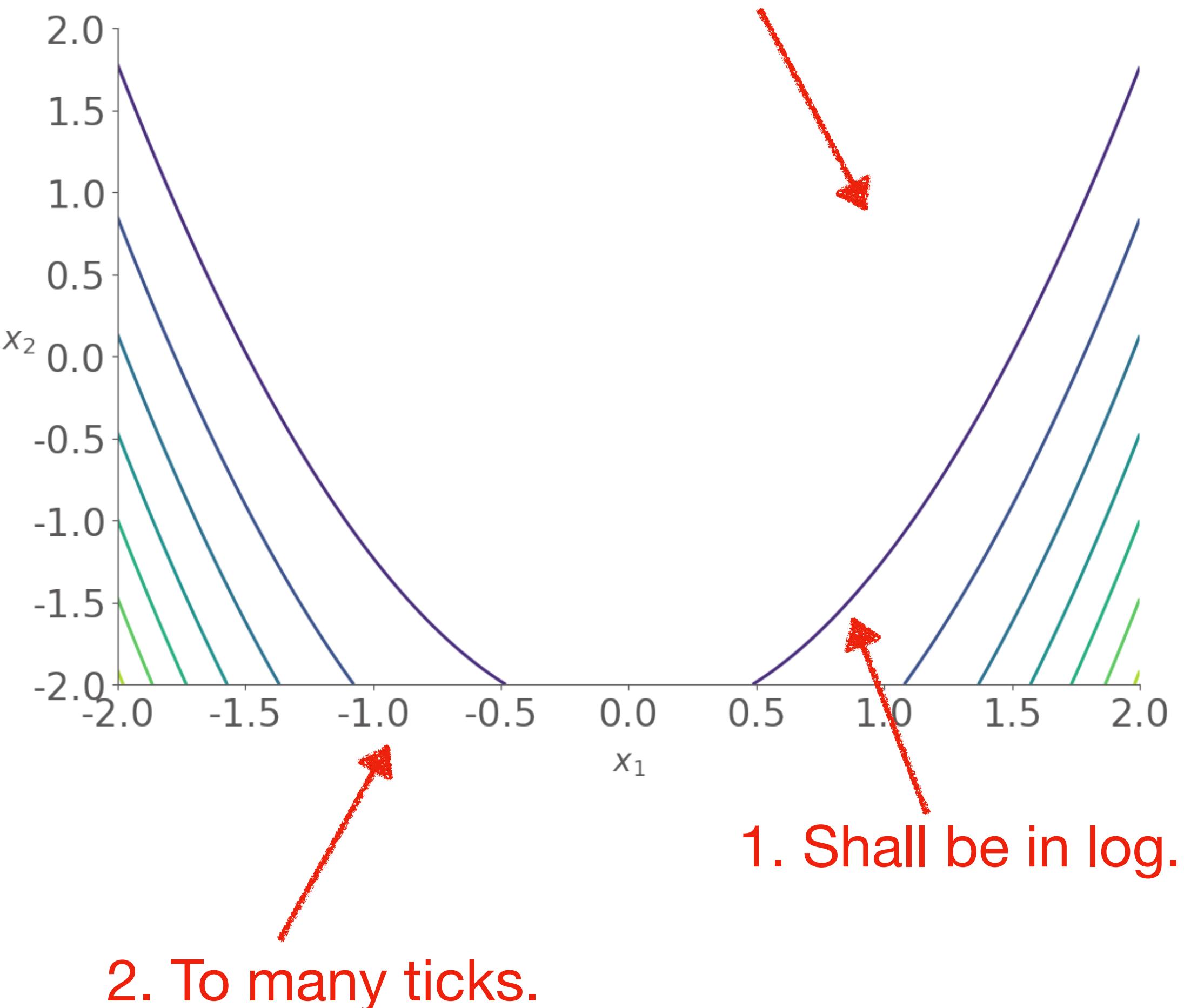
---

```
Input Message set, original figure
for message ∈ message set do
    if message unclear then
        Fix message.
    end if
end for
set = {labels, legends, boxline, words, lines, ticks, colorbar, ...}
for feature ∈ set do
    if feature unclear then
        Fix feature.
    end if
end for
Return Improved figure
```

---

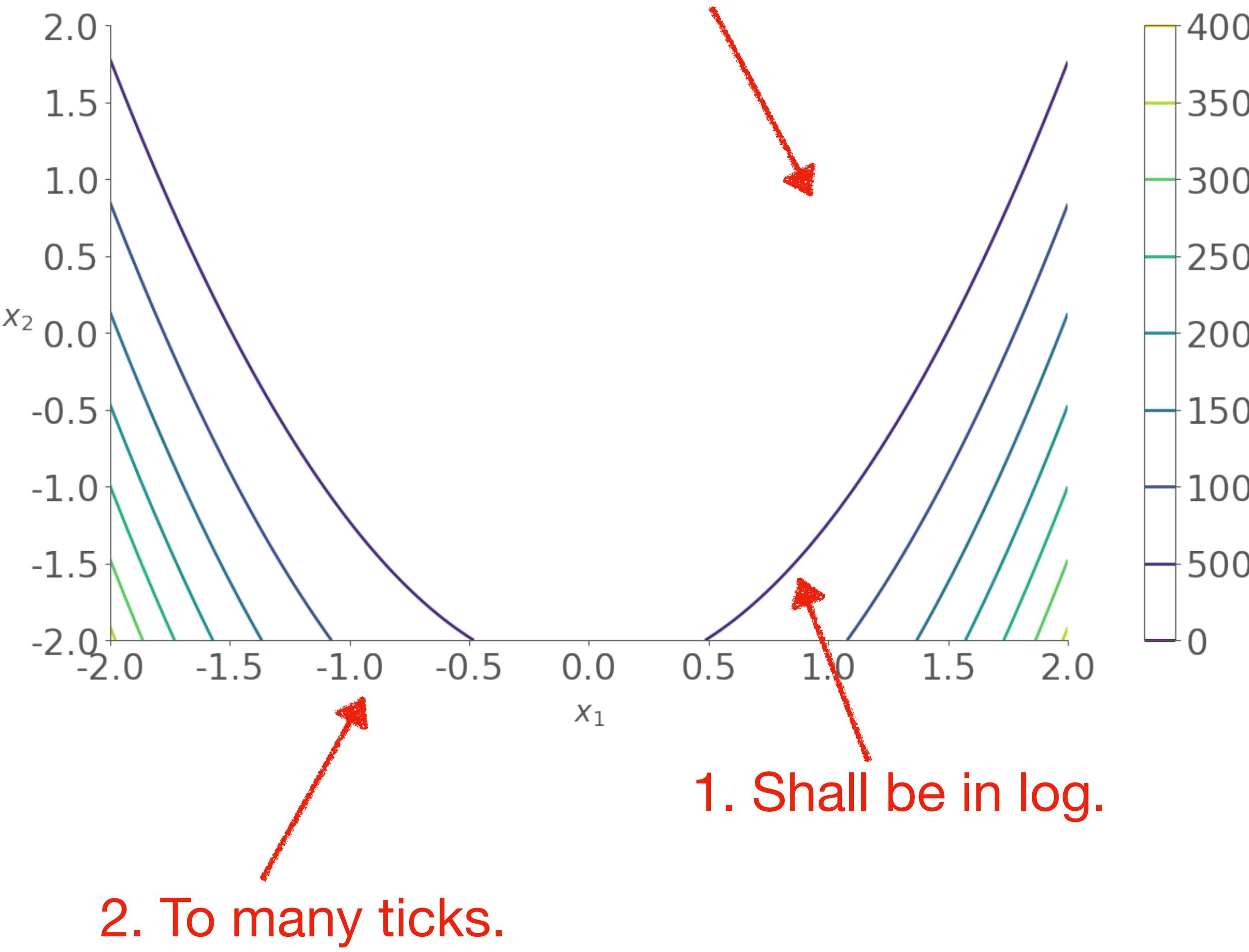
# Improved figure

3. Missing optimal solution.



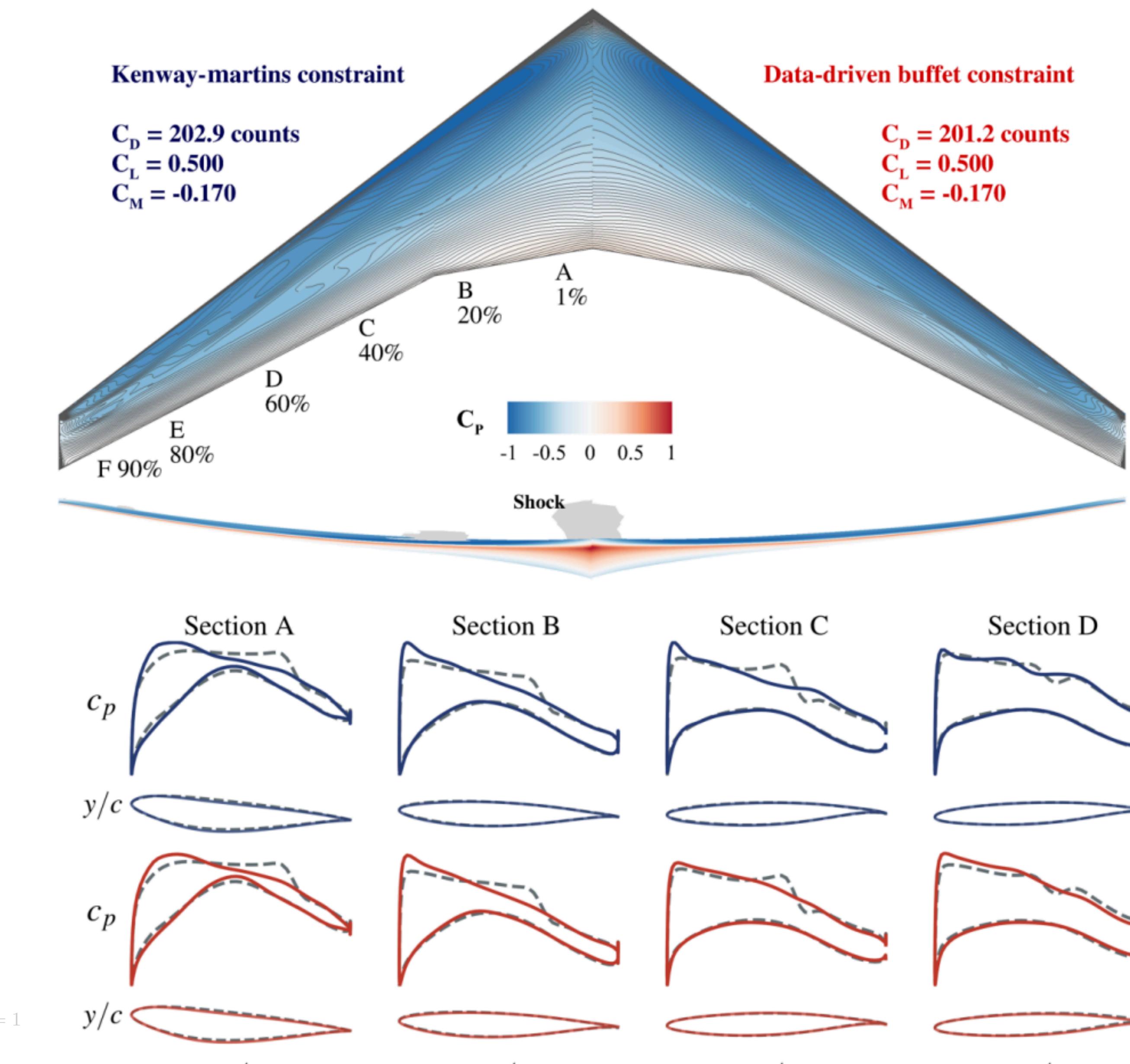
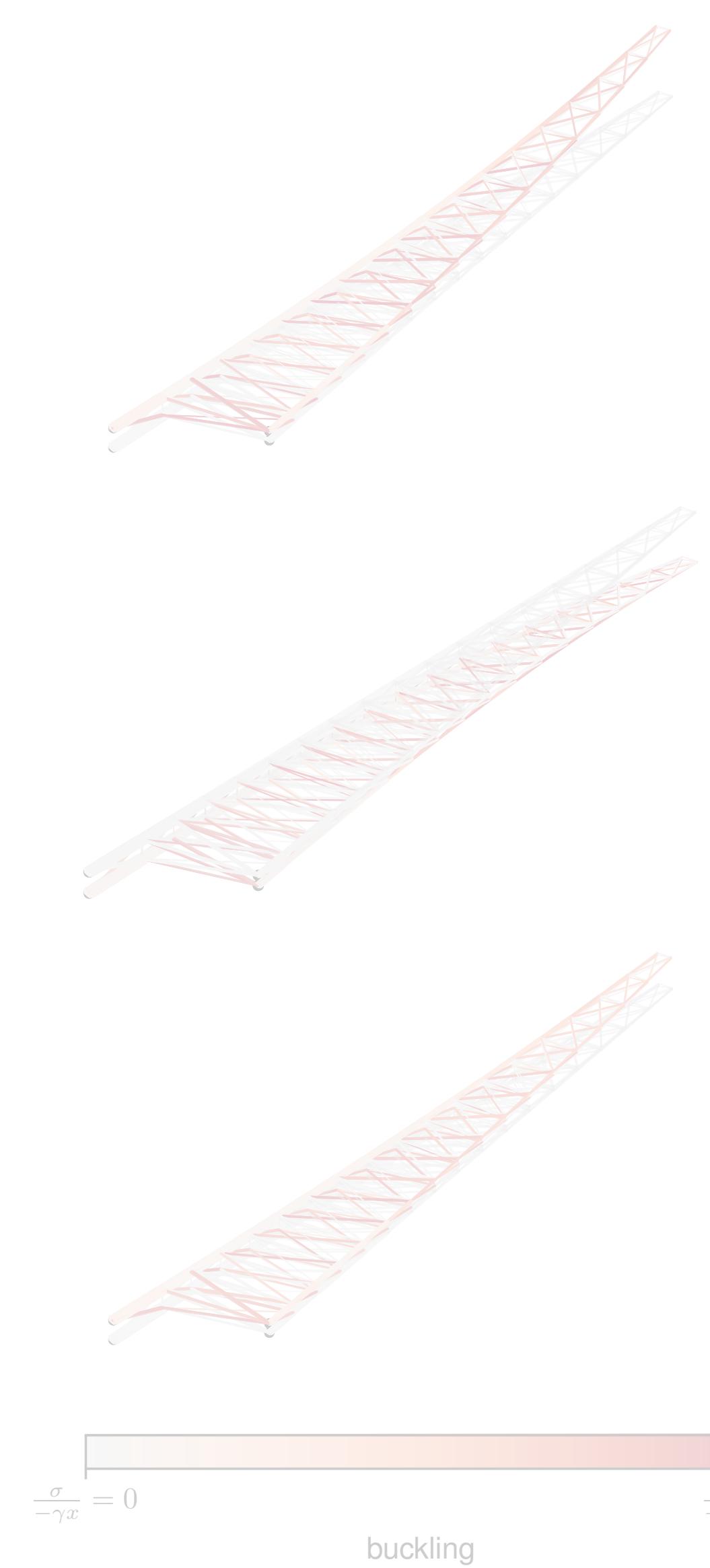
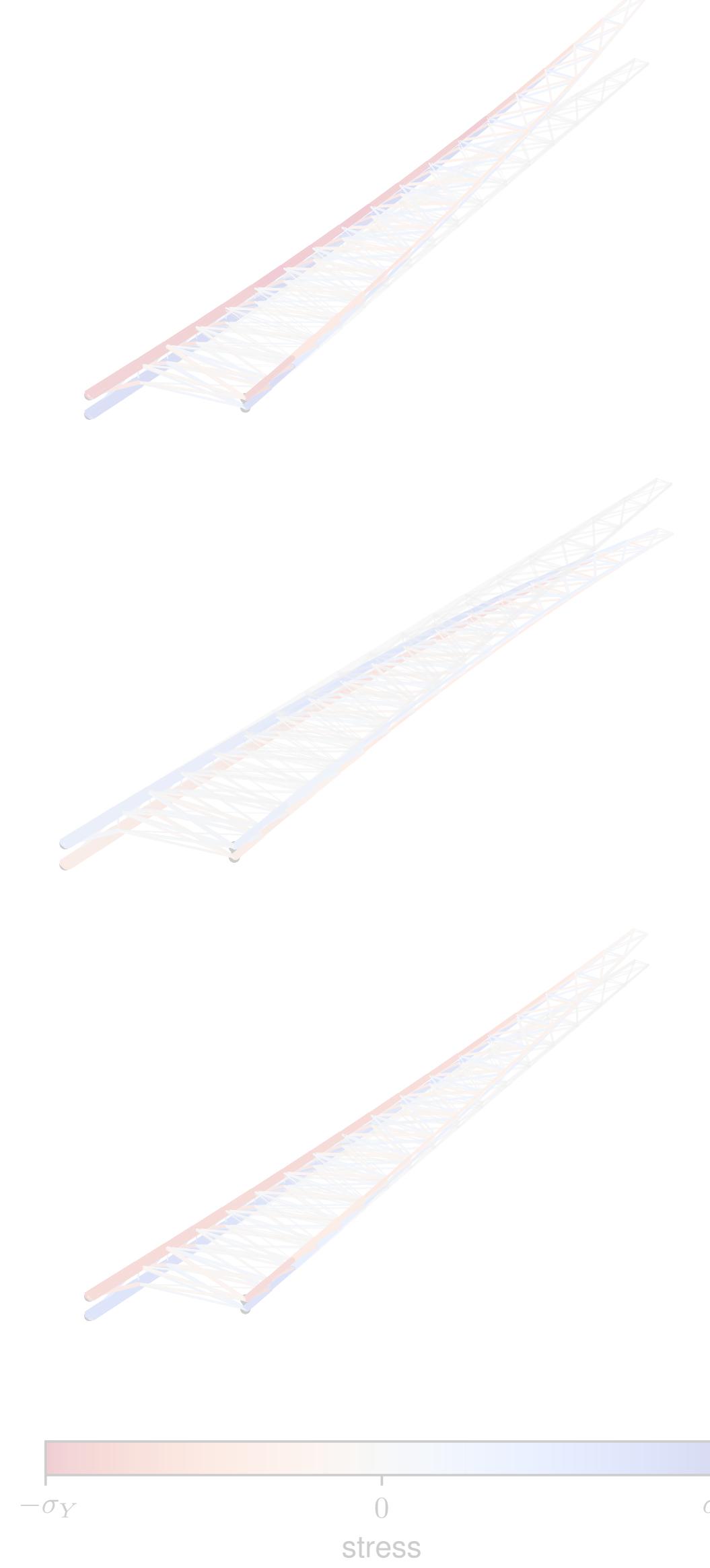
# Case study 2: Contour plot

## 3. Missing optimal solution.

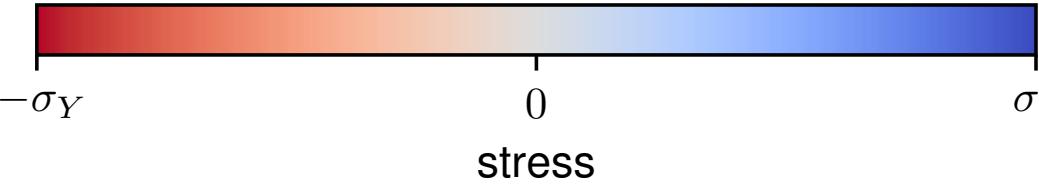
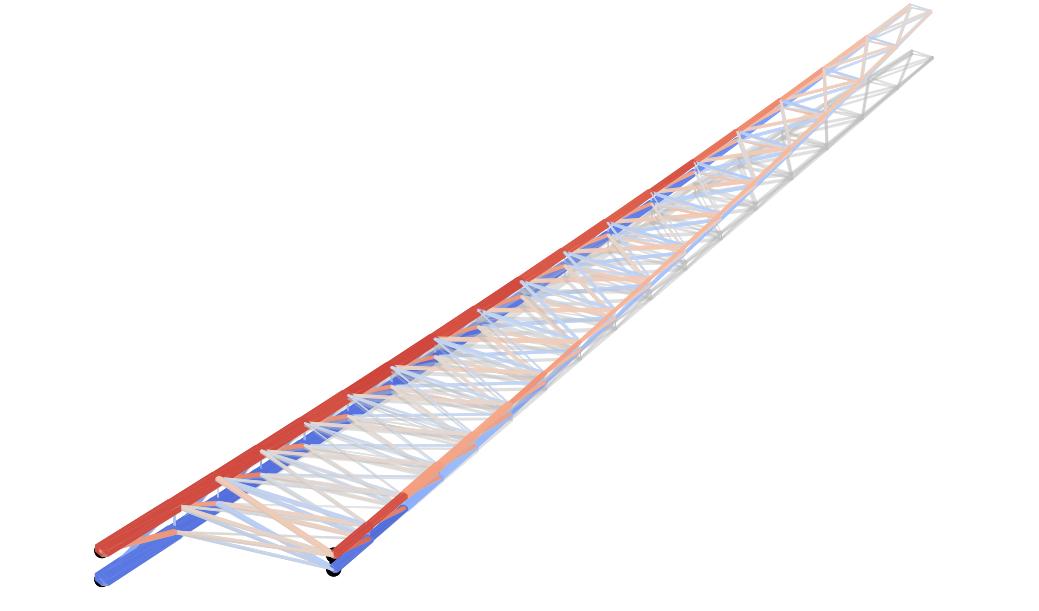
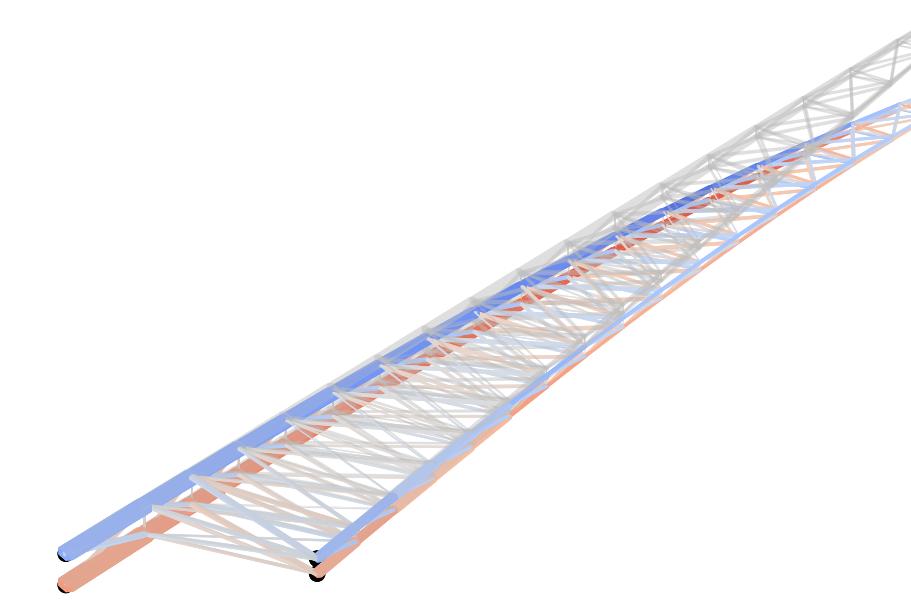
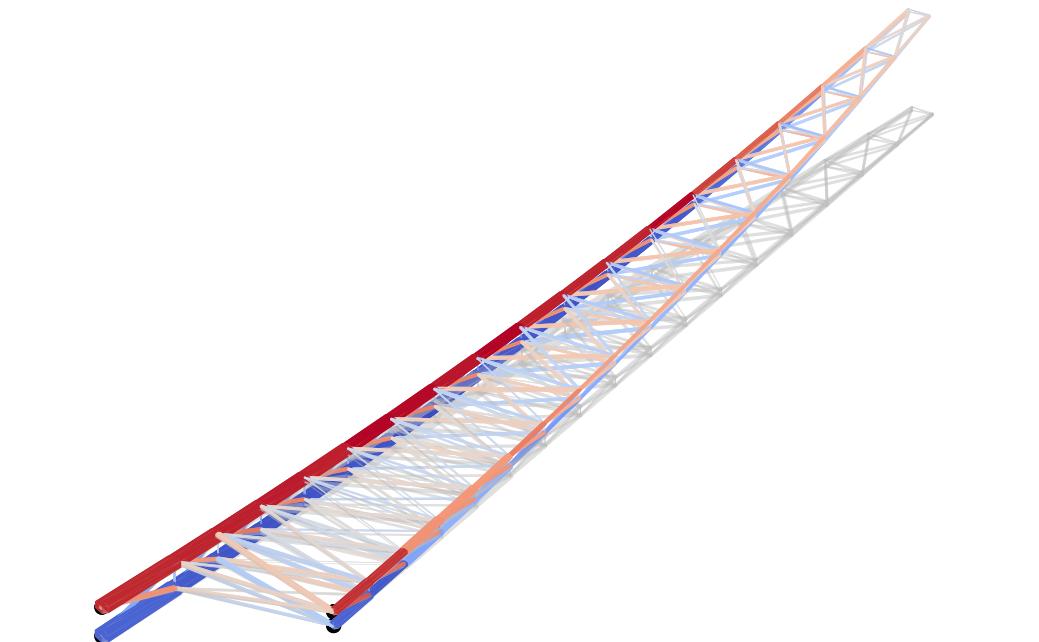


```
1 import matplotlib
2 from matplotlib import ticker, cm
3 import niceplots
4
5
6 # Create the frame
7 fig, ax = plt.subplots()
8 X_arr, Y_arr = np.meshgrid(x_arr, y_arr)
9
10 # Load niceplots
11 niceplots.setRCParams()
12 niceplots.All()
13
14 # Fix (I1)
15 # Define the contour levels
16 levels = np.logspace(-4, 4, 41)
17 # The default colormap actually is okay.
18 CP = ax.contour(X_arr, Y_arr, F_arr.T, levels, locator=plt.LogLocator(), cmap="inferno")
19
20 # Add the label
21 ax.clabel(CP, levels = [10**-3, 10**-2, 10**-1, 1.0, 10.0, 100.0, 1000.0], # label every tenth level
22 | | | | inline=True, fmt = "%1.4g", fontsize=14)
23
24 # Set axis
25 ax.set_xlabel(r'$x_1$', fontsize=20)
26 ax.set_ylabel(r'$x_2$', fontsize=20, rotation=0)
27
28 # Format the colorbar
29 def fmt(x, pos):
30     a, b = '{:.2e}'.format(x).split('e')
31     b = int(b)
32     return r'$10^{{{:{}d}}}$'.format(b)
33
34 fig.colorbar(CP, format=ticker.FuncFormatter(fmt))
35
36 # Redefine the axis ticks (I2)
37 ax.set_xticks([-2, 0, 1, 2], [-2, 0, 1, 2])
38 ax.set_yticks([-2, 0, 1, 2], [-2, 0, 1, 2])
39
40 # Auxillary lines (I3)
41 ax.plot([-2, 1], [1, 1], 'k', alpha = 0.3)
42 ax.plot([1, 1], [-2, 1], 'k', alpha = 0.3)
43 ax.plot([1], [1], "ko", markersize = 15)
44
45 ax.text(1.2, 1, "Minimal", color = 'k')
46
47 plt.show()
```

# Sequential color scheme for sequential data, divergent color scheme for signed data

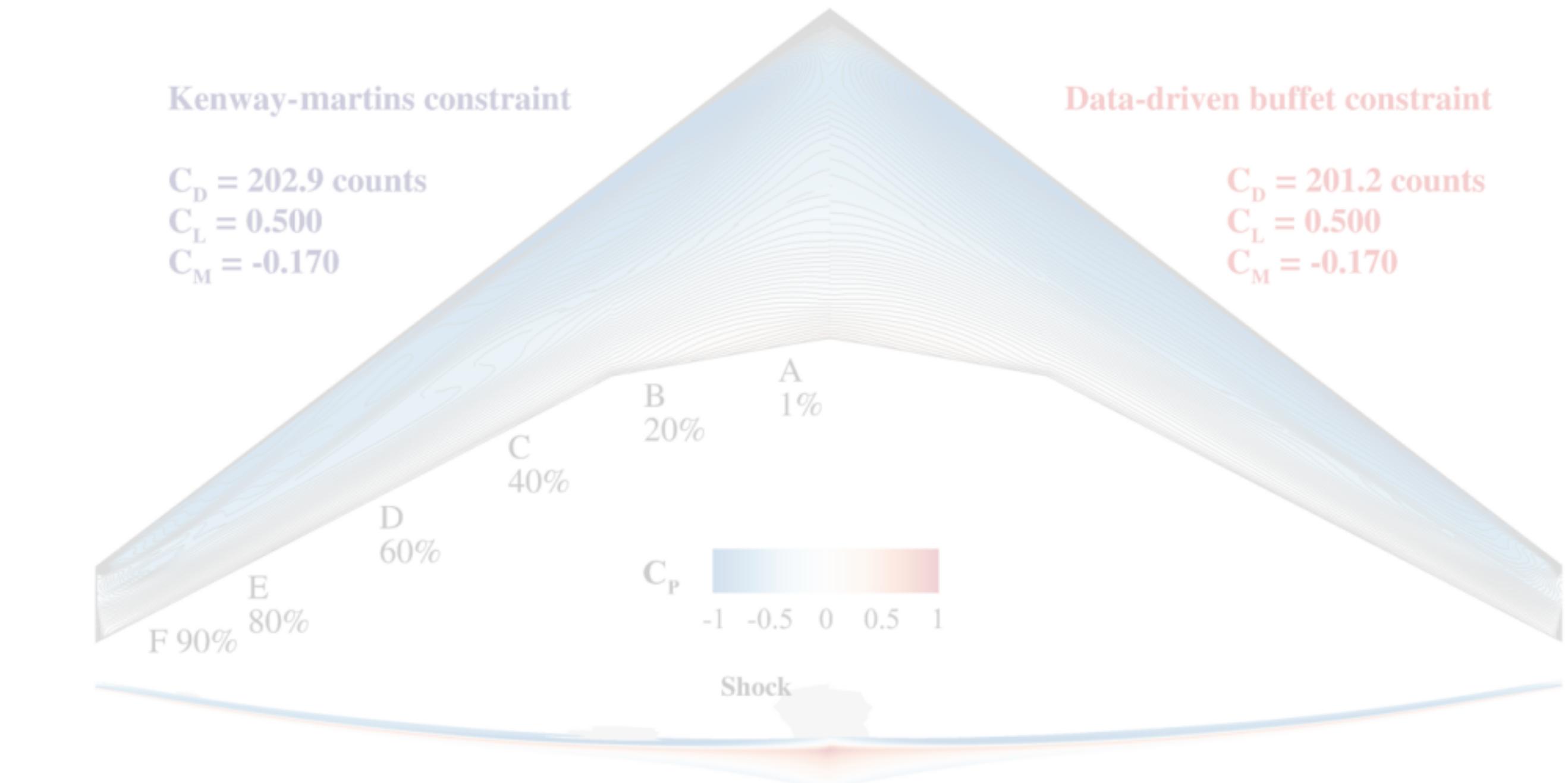


# Sequential color scheme for sequential data, divergent color scheme for signed data



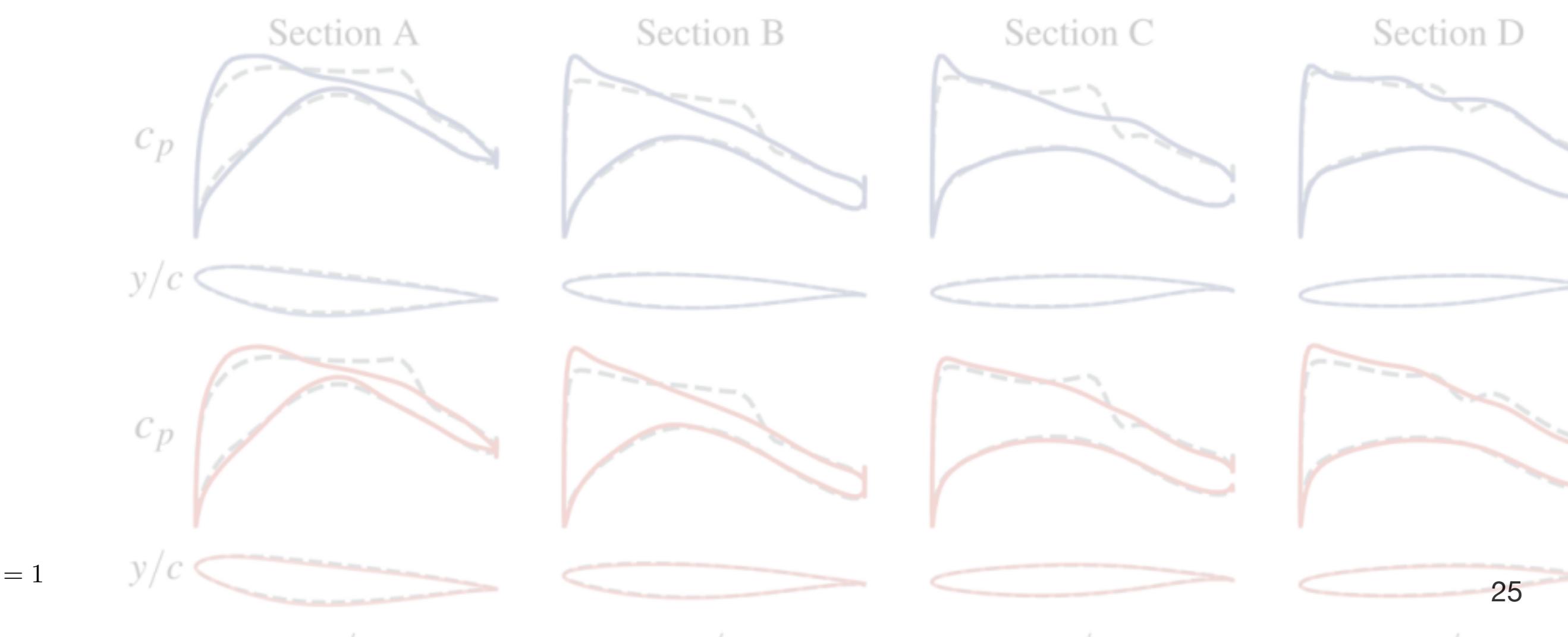
Kenway-martins constraint

$C_D = 202.9$  counts  
 $C_L = 0.500$   
 $C_M = -0.170$

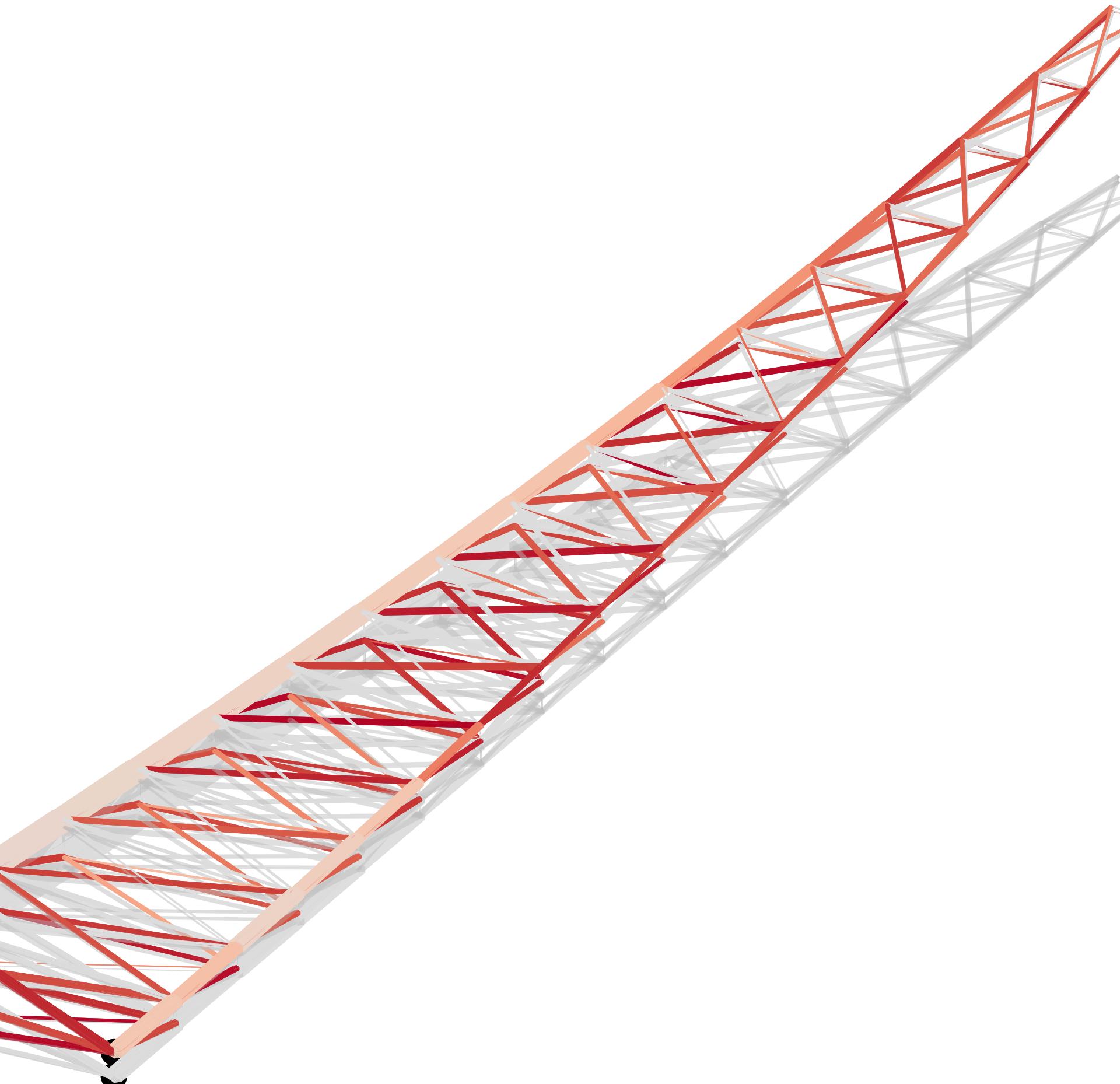


Data-driven buffet constraint

$C_D = 201.2$  counts  
 $C_L = 0.500$   
 $C_M = -0.170$



Always use a vector format (e.g., .pdf, .svg)



PDF

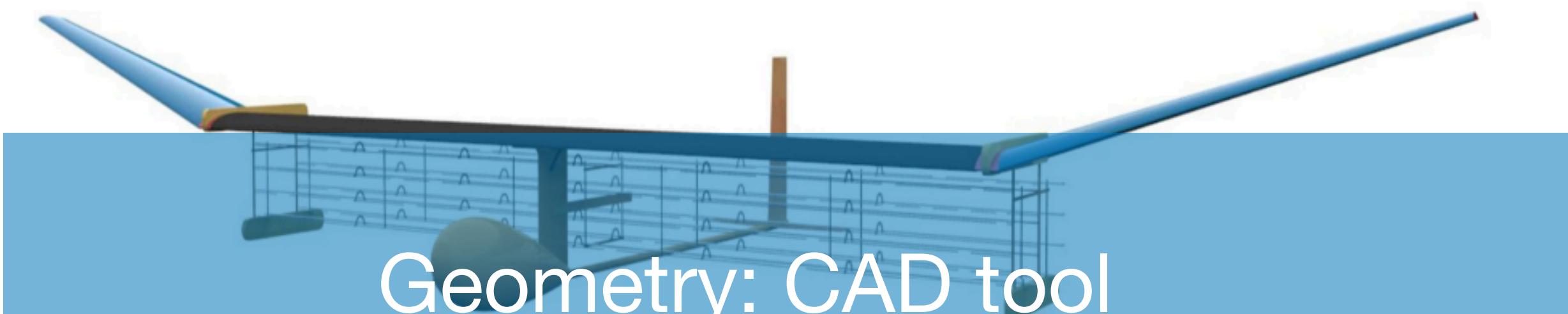


PNG

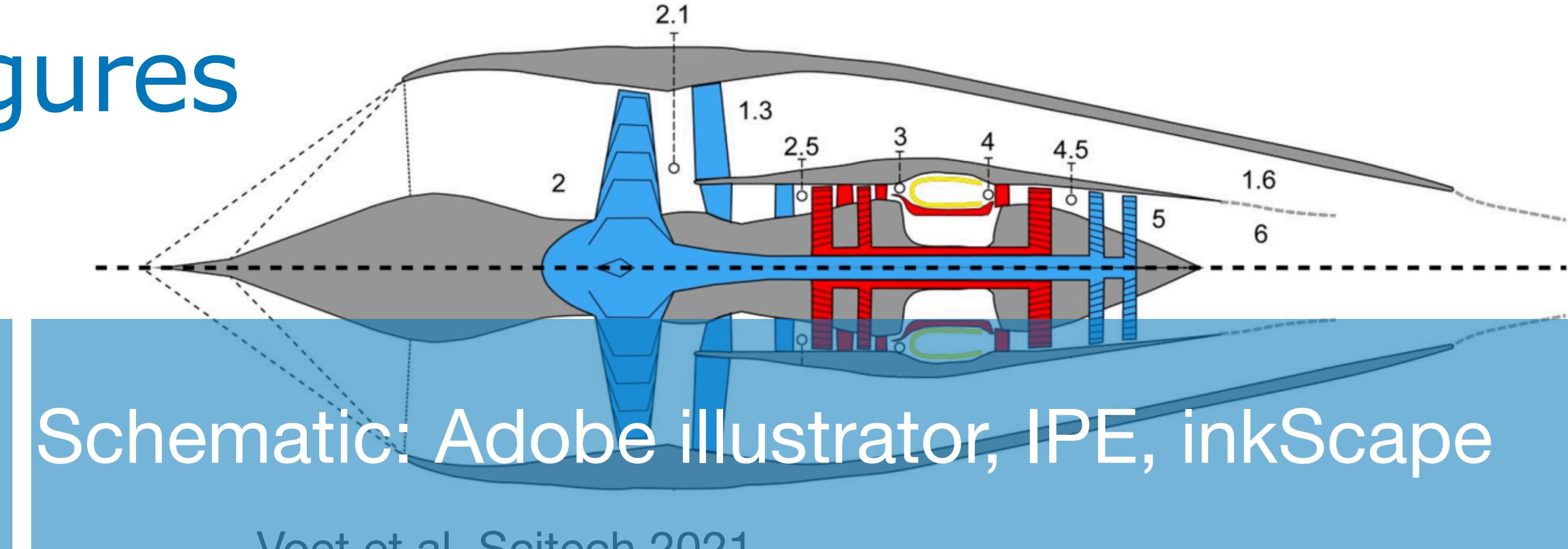
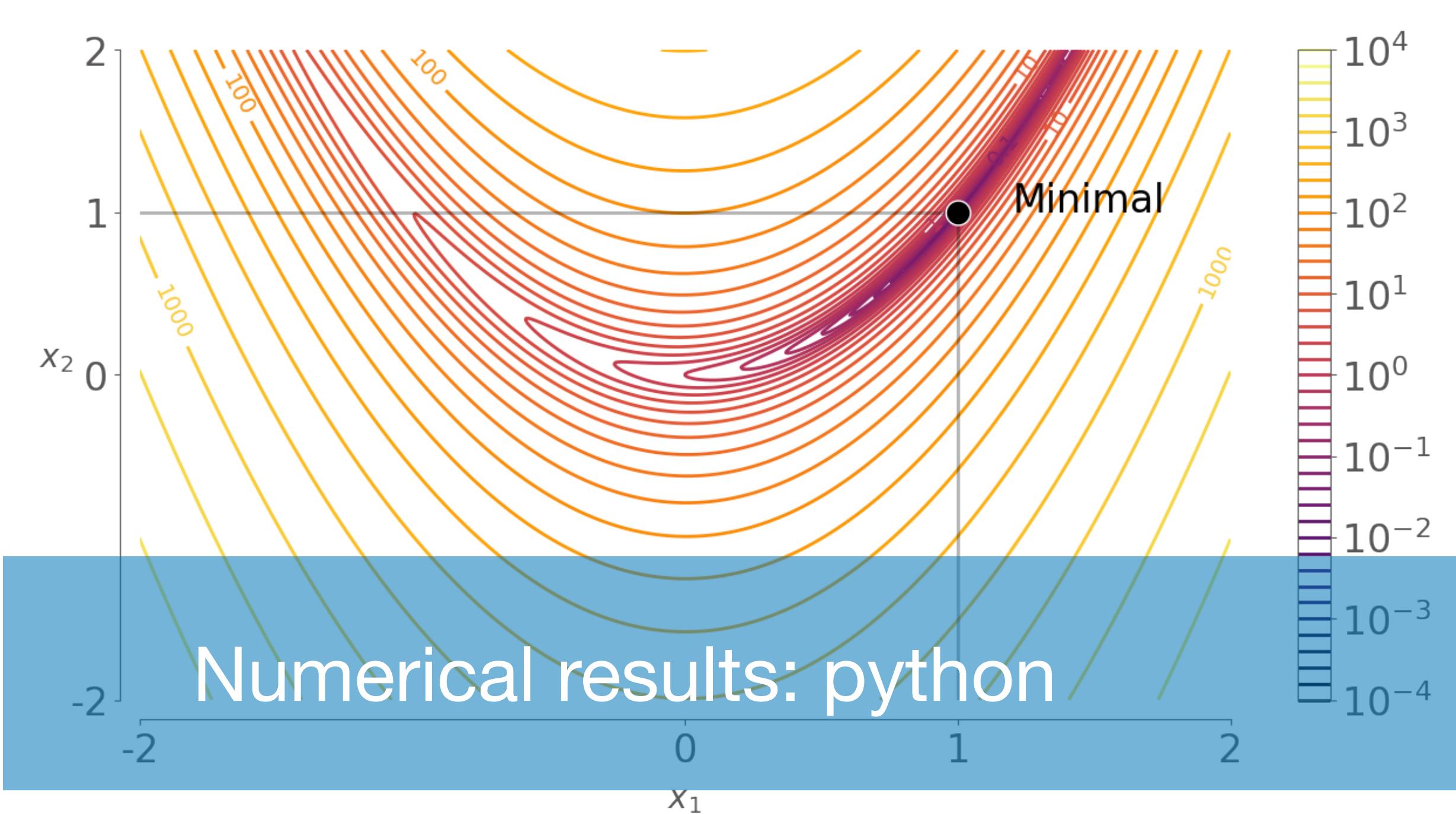
# Module 2: Try it yourself

# Module 3: Conclusion

# Different tools for different figures



Xu et al. Nature 2018



Voet et al. Scitech 2021



# Suggestions

- Do it once and do it well: Create a figure repo for various types of figures, (e.g., curve, contour, etc.)
- Do not reinvent wheels: Start from some existing figures (including your repo) and improvise.

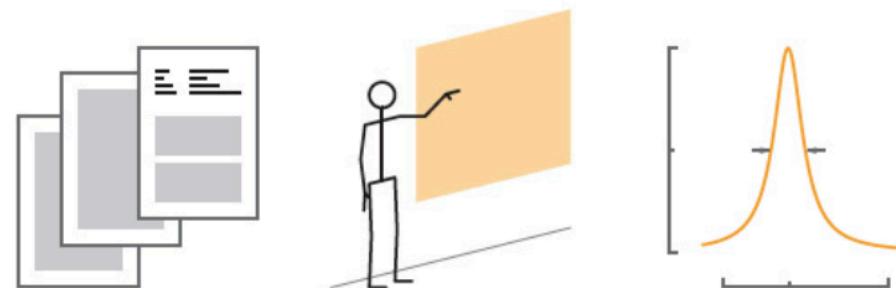
# Recommended readings

- J. Dumont, *Trees, maps, and theorems*
- M. Alley, *The Craft of Scientific Presentations: Critical Steps to Succeed and Critical Errors to Avoid*

Trees, maps, and theorems

Effective communication for rational minds

Jean-luc Doumont



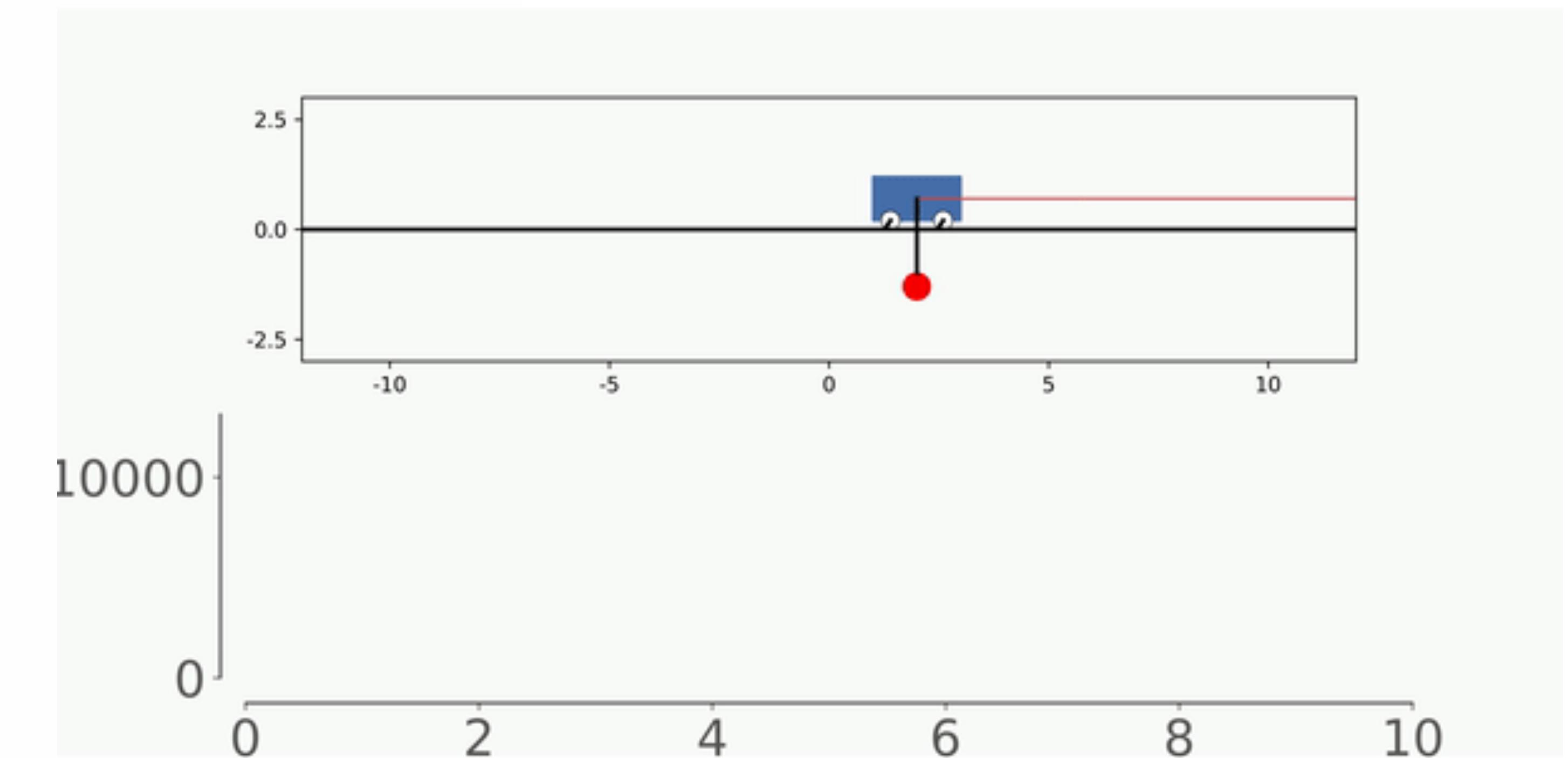
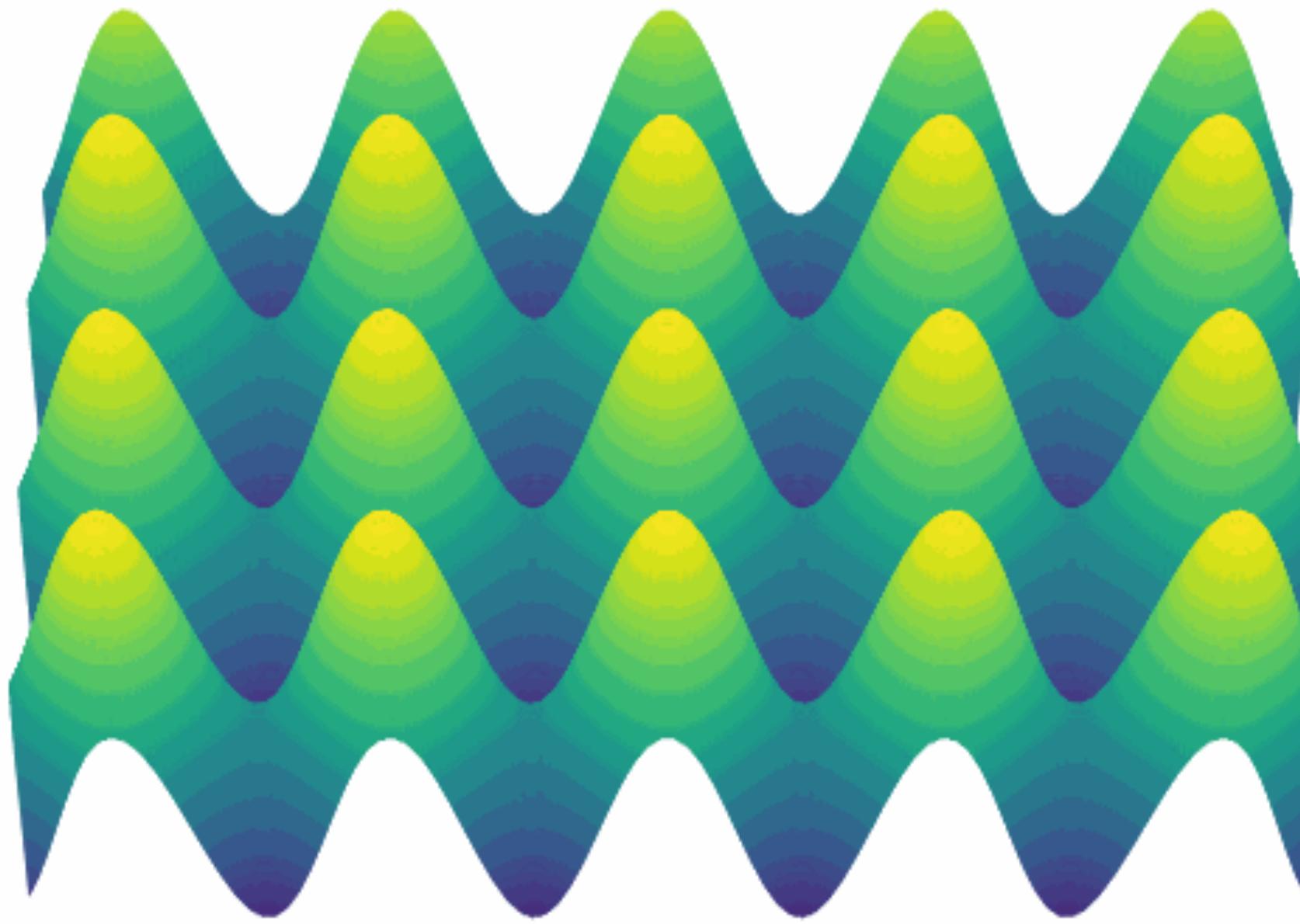
## The Craft of Scientific Presentations

Critical Steps to Succeed and  
Critical Errors to Avoid

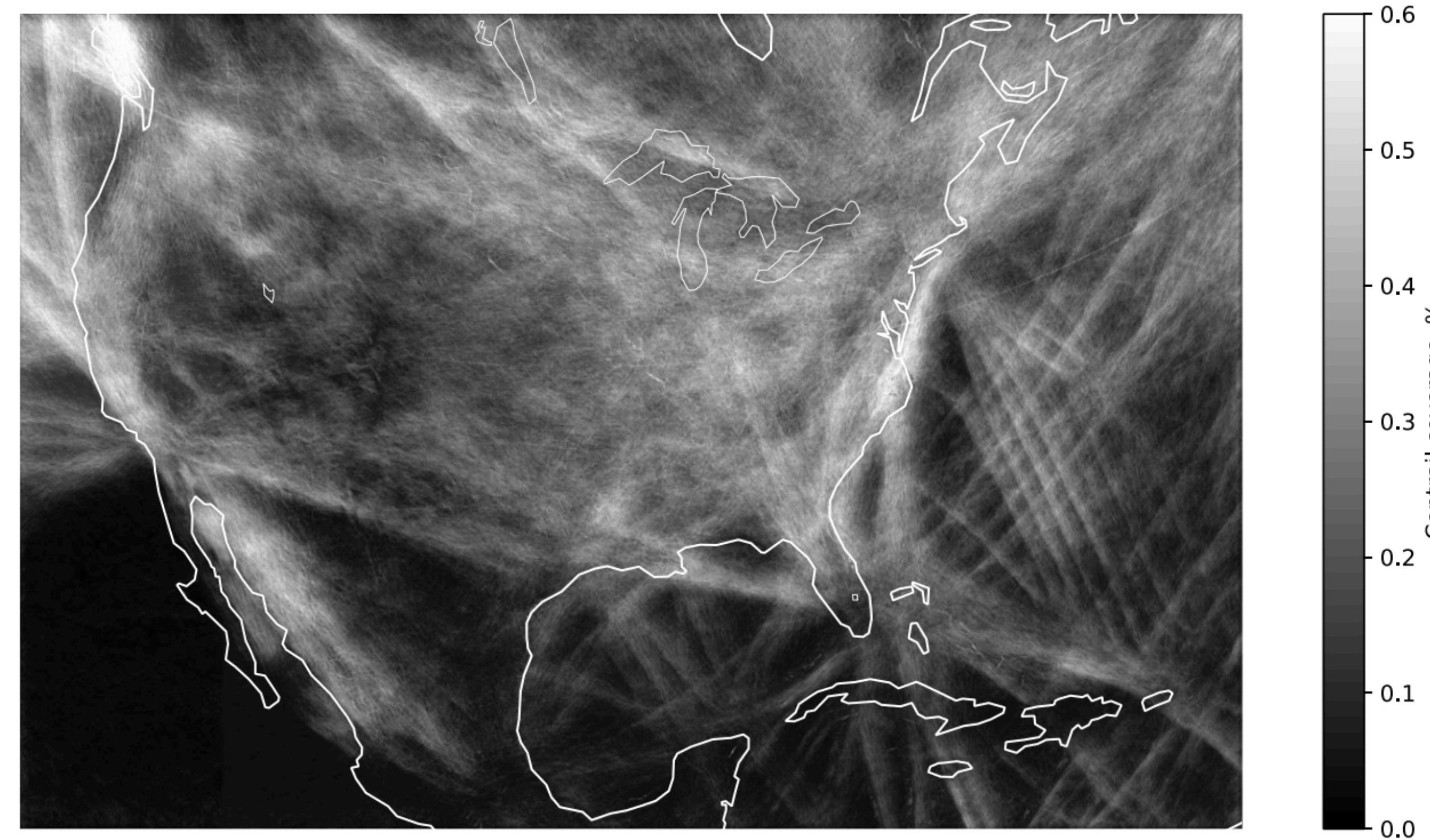
*Second Edition*

 Springer

matplotlib can be used to generate video, if possible  
put some video in your presentation



Put the main message in the title, and waste no space in the slide



Put the main message in the title, and waste no space  
in the slide

Put the main message in the title, and waste no space  
in the slide

A picture is worth a thousand words!  
Go forth and plot more!