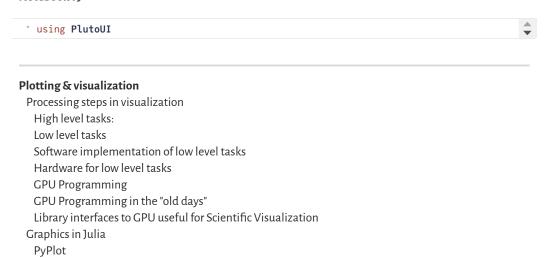
## Scientific Computing TU Berlin Winter 2021/22 © Jürgen Fuhrmann Notebook 19



# Plotting & visualization

Human perception is much better adapted to visual representation than to numbers

Purposes of plotting:

Plots PlutoVista "Bonus track"

- Visualization of research results for publications & presentations
- Debugging + developing algorithms
- "In-situ visualization" of evolving computations
- · Investigation of data
- 1D, 2D, 3D, 4D data
- Similar tasks in CAD, Gaming, Virtual Reality  $\dots$

## Processing steps in visualization

## High level tasks:

- Representation of data using elementary primitives: points, lines, triangles . . .
- Very different depending on purpose

### Low level tasks

- · Coordinate transformation from "world coordinates" of a particular model to screen coordinates
- Transformation 3D ightarrow 2D, visibility computation
- · Coloring, lighting, transparency
- · Rasterization: turn smooth data into pixels

### Software implementation of low level tasks

- Software: rendering libraries, e.g. Cairo, AGG
- · Software for vector based graphics formats, e.g. PDF, postscript, svg
- · Typically performed on CPU

### Hardware for low level tasks

- · Low level tasks are characterized by huge number of very similar operations
- Well adaped to parallelism "Single Instruction, Multiple Data" (SIMD)
- Dedicated hardware: Graphics Processing Unit (GPU) can free CPU from these taks
- · Multiple parallel pipelines, fast memory for intermediate results



(wikimedia)

## **GPU Programming**

- Typically, GPUs are processing units which are connected via bus interface to CPU
- GPU Programming:
  - Prepare low level data for GPU
  - Send data to GPU
  - Process data in rendering pipeline(s)
- Modern visualization programs have a CPU part and GPU parts a.k.a. shaders
  - Shaders allow to program details of data processing on GPU
  - Compiled on CPU, sent along with data to GPU
- Modern libraries: Vulkan, modern OpenGL/WebGL, DirectX
- Possibility to "mis-use" GPU for numerical computations

## GPU Programming in the "old days"

- "Fixed function pipeline" in OpenGL 1.1 fixed one particular set of shaders
- · Easy to program

```
glClear()
glBegin(GL_TRIANGLES)
glVertex3d(1,2,3)
glVertex3d(1,5,4)
glVertex3d(3,9,15)
glEnd()
glSwapBuffers()
```

· Not anymore: now everything works through shaders leading to highly complex programs

## Library interfaces to GPU useful for Scientific Visualization

- vtk (backend of Paraview)
- three.js (for WebGL in the browser)
- vtk.js (for WebGL in the browser)
- plotly.js (for WebGL in the browser)
- · Alternatively, work directly with OpenGL...
- very few . . .
  - Money seems to be in gaming, battlefield rendering . . .
  - Problem regadless of julia, python, C++, ...
- · Common approaches:
  - Write data into "vtk" files, use paraview for visualization
  - Visualize using matlab, Mathematica

## Graphics in Julia

- Plots.jl General purpose plotting package with different backends
  - GPU support via default gr backend (based on "old" OpenGL)
  - · Support in the browser via plotly backend
  - precompilation time significantly improved over the last 2 years
  - Problem: up to now no good support for triangulations
- Makie
  - GLMakie.jl
    - GPU based plotting using modern OpenGL in fact the only package I know (regardless of Julia) besides of vtk.
    - very good plot performance
    - Problem: still under development, long precompilation time
  - WGLMakie.jl maps Makie API to three.js, can be used from the browser
    - Problem: not very stable in the moment
- PyPlot.jl: Interface to python/matplotlib
  - realization via PyCall.jl
  - Full functionality of matplotlib
  - also as backend for Plots.jl
  - · Problem: slow
- WriteVTK.jl vtk file writer for files to be used with paraview so this is not a plotting library.

## **PyPlot**

Here I show some examples with PyPlot from the previous course

PyPlot resources:

- Julia package
- Julia examples
- Matplotlib documentation



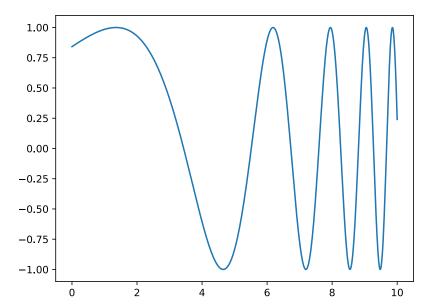


We can choose the way the plot is created: in the browser it can make sense to create it as a vector graphic in svg format. The alternatice is png, a pixel based format.

true

PyPlot.svg(true)

How to create a plot?

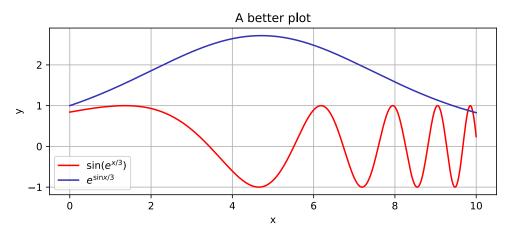


```
Let
    X=collect(0:0.01:10)
    PyPlot.clf() # Clear the figure
    PyPlot.plot(X,sin.(exp.(X/3))) # call the plot function
    figure=PyPlot.gcf() # return figure to Pluto
end
```

Instead of a begin/end block we used a let block. In a let block, all new variables are local and don't interfer with other pluto cells.

This plot is not nice. It lacks:

- orientation lines ("grid")
- title
- axis labels
- · label of the plot
- size adjustment



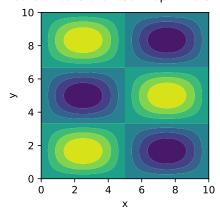
```
let
    X=collect(0:0.01:10)
    PyPlot.clf()
    PyPlot.plot(X,sin.(exp.(X/3)),
        label="\$\\sin(e^{x/3})\$", color=:red) # Plot with label
    PyPlot.plot(X,exp.(sin.(X/3)),
    label="\$e^{\\sin x/3}\$",color=(0.2,0.2,0.7)) # Plot with label
PyPlot.legend(loc="lower left") # legend placement
    PyPlot.title("A better plot") # The plot title
    PyPlot.grid() # add grid lines to the plot
    PyPlot.xlabel("x") # x axis label
    PyPlot.ylabel("y") # y axis label
    figure=PyPlot.gcf()
    figure.set_size_inches(8,3) # adjust size
    PyPlot.savefig("myplot.png") # save figure to disk
    figure # return figure
end
```

We can use  $\angle TEX$  math strings in plot labels here, we just need to escape the \$ symbols with \!!

### Plotting 2D data

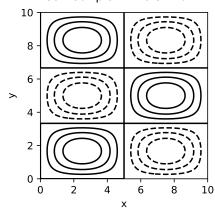


Filled contours aka heatmap: k=0.3 I=0.2



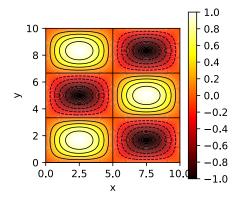
```
PyPlot.clf()
X=collect(0:0.05:10)
Y=X
PyPlot.suptitle("Filled contours aka heatmap: k=$(k) l=$(l)")
F=[sin(k*π*X[i])*sin(l*π*Y[j]) for i=1:length(X), j=1:length(Y)]
PyPlot.contourf(X,Y,F) # plot filled contours
PyPlot.xlabel("x")
PyPlot.ylabel("y")
figure=PyPlot.gcf()
figure.set_size_inches(3,3)
figure
end
```

### Contour plot: k=0.3 l=0.2



```
PyPlot.clf()
X=collect(0:0.05:10)
Y=X
PyPlot.suptitle("Contour plot: k=$(k) l=$(l)")
F=[sin(k*π*X[i])*sin(l*π*Y[j]) for i=1:length(X), j=1:length(Y)]
PyPlot.contour(X,Y,F,colors=:black)
PyPlot.xlabel("x")
PyPlot.ylabel("y")
figure=PyPlot.gcf()
   figure.set_size_inches(3,3)
   figure
end
```

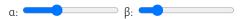
#### Contour + filled contours: k=0.3 l=0.2



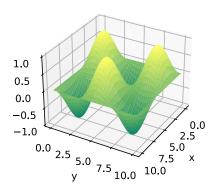
```
let
    PyPlot.clf()
    X=collect(0:0.05:10)
    PyPlot.suptitle("Contour + filled contours: k=$(k) l=$(l)")
    F=[\sin(k*\pi*X[i])*\sin(l*\pi*Y[j]) \text{ for } i=1:length(X), j=1:length(Y)]
    fmin=minimum(F)
    fmax=maximum(F)
    number_of_isolines=10
    isolines=collect(fmin:(fmax-fmin)/number_of_isolines:fmax)
    cnt=PyPlot.contourf(X,Y,F,cmap="hot",levels=100)
    if fix_moire
        for c in cnt.collections
           c.set_edgecolor("face")
        end
    end
    axes=PyPlot.gca()
    axes.set_aspect(1)
    PyPlot.colorbar(ticks=isolines)
    PyPlot.contour(X,Y,F,colors=:black,linewidths=0.75,levels=isolines)
    PyPlot.xlabel("x")
    PyPlot.ylabel("y")
    figure=PyPlot.gcf()
    figure.set_size_inches(3,3)
    figure
end
```

Remove the moire in the plot:  $\Box$ 

This occurs in contourf when we use many colors to make a smooth impression.



Surface plot: k=0.3 I=0.2



```
PyPlot.clf()
X=collect(0:0.05:10)
Y=X
PyPlot.suptitle("Surface plot: k=$(k) l=$(l)")
F=[sin(k*π*X[i])*sin(l*π*Y[j]) for i=1:length(X), j=1:length(Y)]

PyPlot.surf(X,Y,F,cmap=:summer) # 3D surface plot
ax=PyPlot.gca(projection="3d") # Obtain 3D plot axes
ax.view_init(α,β) # Adjust viewing angles

PyPlot.xlabel("x")
PyPlot.ylabel("y")
figure=PyPlot.gcf()
figure.set_size_inches(3,3)
figure
end
```

There are analogues for contour contourf and surf on triangular meshes which will be discussed once we get there in the course.

### **Plots**

```
" using Plots

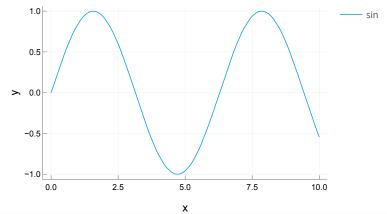
In Pluto it is best to use the plotly interface. Plotly is a Javascript library for plotting which is quite
good and all kinds of x-y plots.

PlotlyBackend()

Plots.plotly()

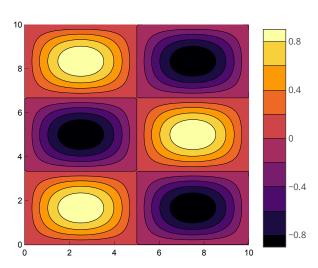
X =
[0.0, 0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9, 1.0, 1.1, 1.2, 1.3, 1.4, 1.5, 1.6, 1.7, 1]

**X=collect(0:0.1:10)
```

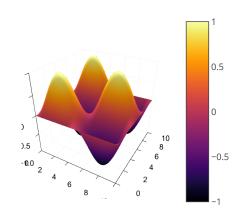


- X
   Plots.plot(X,sin.(X),size=(500,300),xlabel="x",ylabel="y",label="sin")
- \*  $F=[\sin(k1*\pi*X[i])*\sin(l1*\pi*X[j])$  for i=1:length(X), j=1:length(X)];





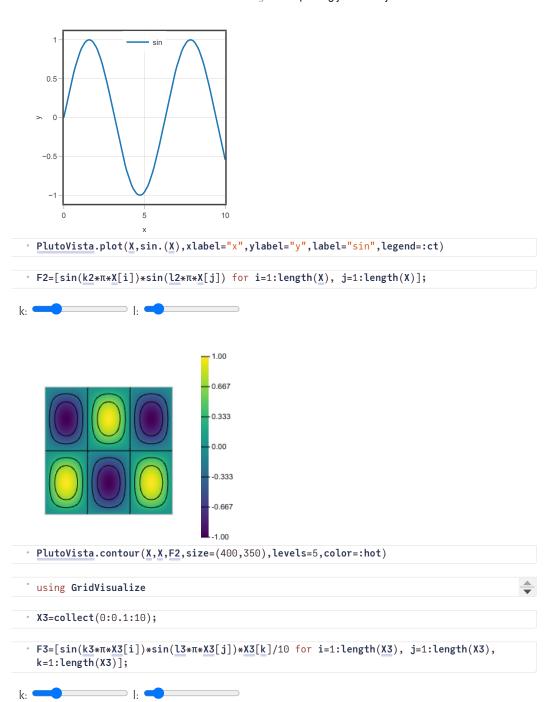
Plots.contour(X,X,F,fill=true,size=(400,350))

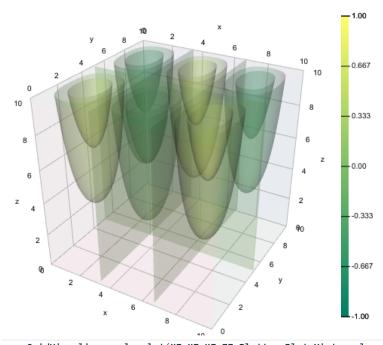


Plots.surface(X,X,F,size=(300,300))

### PlutoVista

° using PlutoVista





GridVisualize.scalarplot(X3,X3,X3,F3,Plotter=PlutoVista,colormap=:summer)

This renders 1030301 values.

### "Bonus track"

Feel free watch my <u>vizcon2 talk</u> about using vtk from Julia - just to show what could be possible. Unfortunately, these things currently work only on Linux...

Since then, I created code which can do many of these things with GLMakie or PlutoVista through **<u>GridVisualize.jl</u>** 

```
LoadError: UndefVarError: @htl_str not defined
```

in expression starting at /home/fuhrmann/Wias/teach/scicomp/pluto/nb19-plotting.jl#==#95f489e6-6ea8-4281-adf2-67c7dacc905c:3

```
1. top-level scope @ :0
2. #macroexpand#50 @ expr.jl:112 [inlined]
3. macroexpand @ expr.jl:111 [inlined]
4. try_macroexpand(::Module, ::Base.UUID, ::Expr) @ PlutoRunner.jl:248
5. var"#run_expression#25"(::Bool, ::typeof(Main.PlutoRunner.run_expression), ::Module, ::Expr, ::Base.UUID, ::Nothing, ::Nothing) @ PlutoRunner.jl:477
6. top-level scope @ none:1
```

```
begin

using HypertextLiteral
highlight(mdstring,color)= htl"""<blockquote style="padding: 10px; background-color: $(color);">$(mdstring)</blockquote>"""

macro important_str(s) :(highlight(Markdown.parse($s),"#ffcccc")) end
macro definition_str(s) :(highlight(Markdown.parse($s),"#cccff")) end
macro statement_str(s) :(highlight(Markdown.parse($s),"#ccffcc")) end

html"""
<style>
h1{background-color:#dddddd; padding: 10px;}
h2{background-color:#e7e7e7; padding: 10px;}
h3{background-color:#eeeeee; padding: 10px;}
h4{background-color:#f7f7f7; padding: 10px;}
</style>
"""
end
```