

# Python for Data Analysis and Scientific Computing

X433.3 (2 semester units in COMPSCI)

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# Python for Data Analysis and Scientific Computing

Lecture 5 part 2 ...

*... show project presentations ...*

# Class exercise

- Solution 1/2

```
hw3_alex.py  lecture5.py  lecture6.py  Midterm.py
1  # =====
2  # Part 1 – create your data:
3  # =====
4
5  # 1. Include a section line with your name:
6  ## HW 2: Alexander Iliev
7
8  # 2. Work only with these imports:
9  from numpy import matrix, array, random, min, max
10 import pylab as plb
11
12 # 3. Cerate a list A of 600 random numbers bound between (0:10):
13 A = list(random.random(600)*10)
14
15 # 4. Create an array B with 500 elements bound in the range [-3*pi:2*pi]:
16 B = plb.linspace(-plb.pi*3, plb.pi*2, 500)
17
18 # 5. Using if, for or while, create a function that overwrites every element
19 # in A that falls outside of the interval [2:9), and overwrite that element with
20 # the average between the smallest and largest element in A:
21 index = 0;
22 def my_function(x):
23     for index, k in enumerate(x, start=0):
24         if k < 2:
25             x[index] = (min(x)+max(x))/2
26         elif k >= 9:
27             x[index] = (min(x)+max(x))/2
28         # 6. Normalize each list element to be bound between [0:0.1]:
29         x[index] = x[index] / 100
30     return(x)
31
32 # 7. Return the result from the function to C:
33 C = my_function(A)
34
35 # 8. Cast C as an array:
36 C = array(C)
37
38 # 9. Add C to B (think of C as noise) and record the result in D:
39 D = C[0:len(B)] + B
```

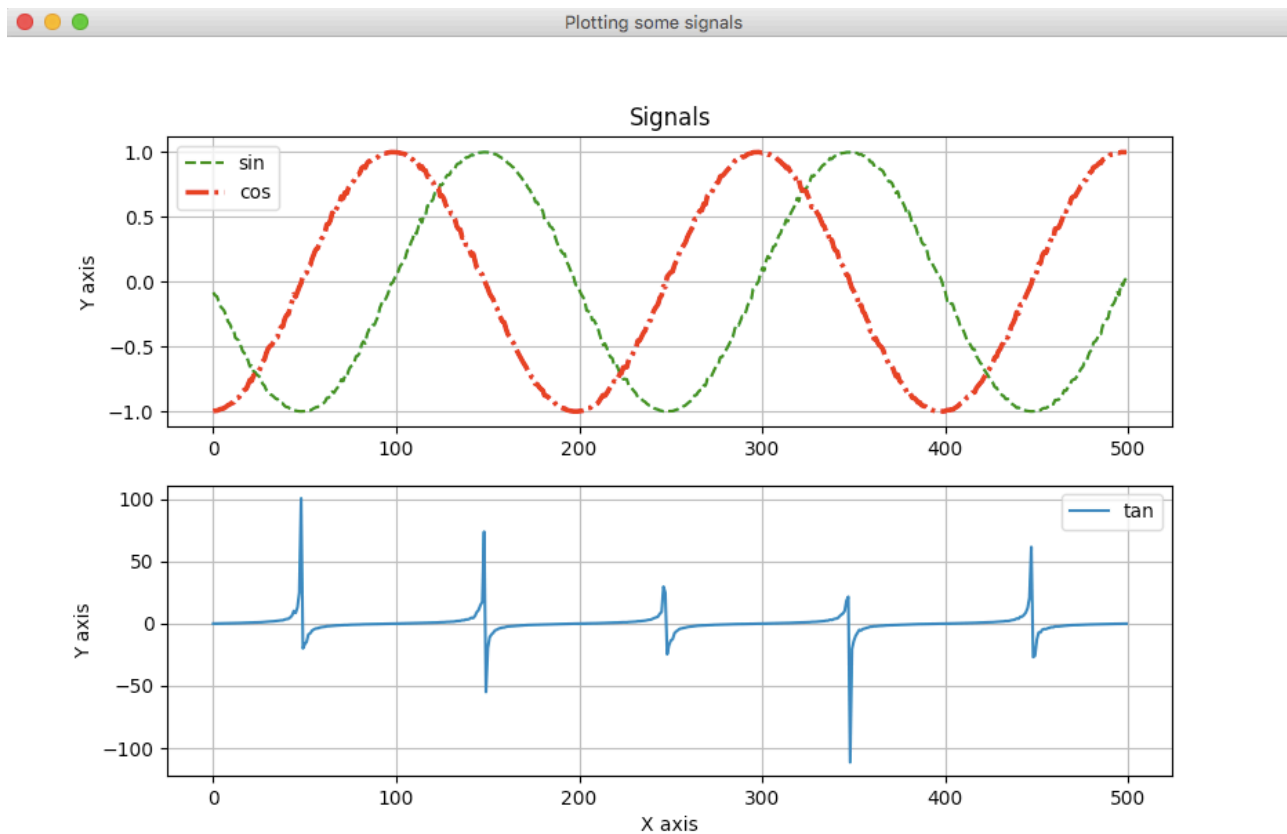
# Class exercise

- Solution 2/2

```
41 # =====
42 # Part 2 - plotting:
43 # =====
44
45 # 10. Create a figure, give it a title and specify your own size and dpi:
46 plb.figure('Plotting some signals', figsize=(6,4), dpi=100)
47
48 # 11. Plot the sin of D, in the (2,1,1) location of the figure:
49 plb.subplot(2,1,1), plb.plot(plb.sin(D), color="green", linewidth=1.5, linestyle="--", label='sin')
50
51 # 12. Overlay a plot of cos of D, with different color, thickness and type of
52 # line:
53 plb.plot(plb.cos(D), color="red", linewidth=2.5, linestyle="-. ", label='cos')
54
55 # 13. Create some space on top and bottom of the plot (on the y axis) and show
56 # the grid:
57 plb.ylim(-1.12, 1.12), plb.grid()
58
59 # 14. Specify the following: title, Y-axis label and legend to fit in the best way:
60 plb.ylabel('Y axis'), plb.title('Signals'), plb.legend(loc='best')
61
62 # 15. Plot the tan of D, in location (2,1,2) with grid showing, X-axis label,
63 # Y-axis label and legend on top right:
64 plb.subplot(2,1,2), plb.plot(plb.tan(D), label='tan'), plb.grid()
65 plb.xlabel('X axis'), plb.ylabel('Y axis'), plb.legend(loc='upper right')
```

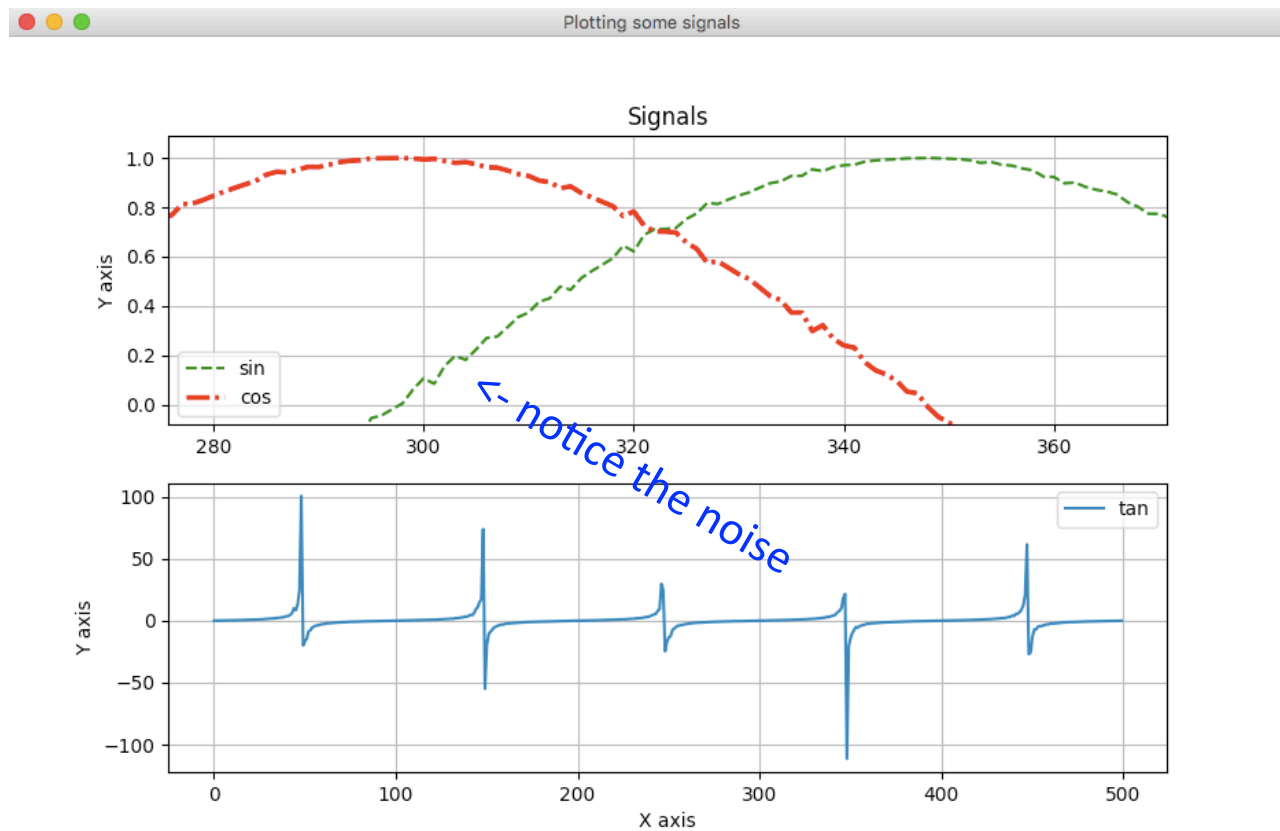
# Class exercise

- Solution plots



# Class exercise

- Solution plots



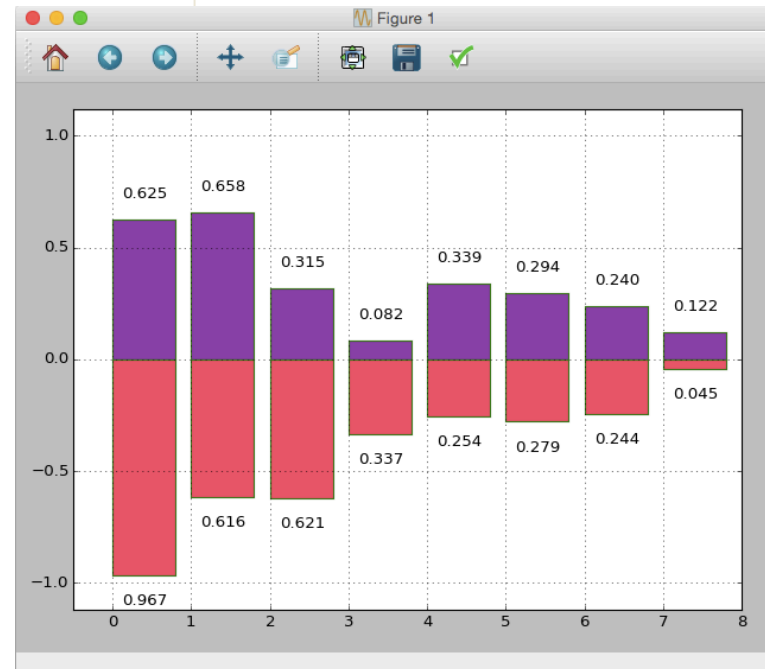
# Advanced plotting

- Advanced plotting
  - There are many different types of 2-D and 3-D plotting choices in Matplotlib
  - Matplotlib provides a huge variety of plots aiming at specific visualization of results
  - Some of these types are:
    - Bar plots
    - Scatter plots
    - Imshow
    - Histogram
    - Pie charts
    - Contour plots
    - Polar axis
    - 3-D plots
    - *Text plots*
    - *Grids*
    - *Quiver plots*

# Advanced plotting

- Advanced plotting: *bar plot*

```
244 ## Advanced plotting:
245 # Bar plot:
246 import pylab as plb
247
248 k = 8
249 x = plb.arange(k)
250 y1 = plb.rand(k) * (1 - x / k)
251 y2 = plb.rand(k) * (1 - x / k)
252 plb.axes([0.075, 0.075, .88, .88])
253
254 plb.bar(x, +y1, facecolor='#9922aa', edgecolor='green')
255 plb.bar(x, -y2, facecolor='#ff3366', edgecolor='green')
256
257 for a, b in zip(x, y1):
258     plb.text(a+0.41, b+0.08, '%.3f' % b, ha='center', va='bottom')
259 for a, b in zip(x, y2):
260     plb.text(a+0.41, -b-0.08, '%.3f' % b, ha='center', va='top')
261
262 plb.xlim(-.5, k), plb.ylim(-1.12, +1.12)
263 plb.grid(True)
264 plb.show()
```

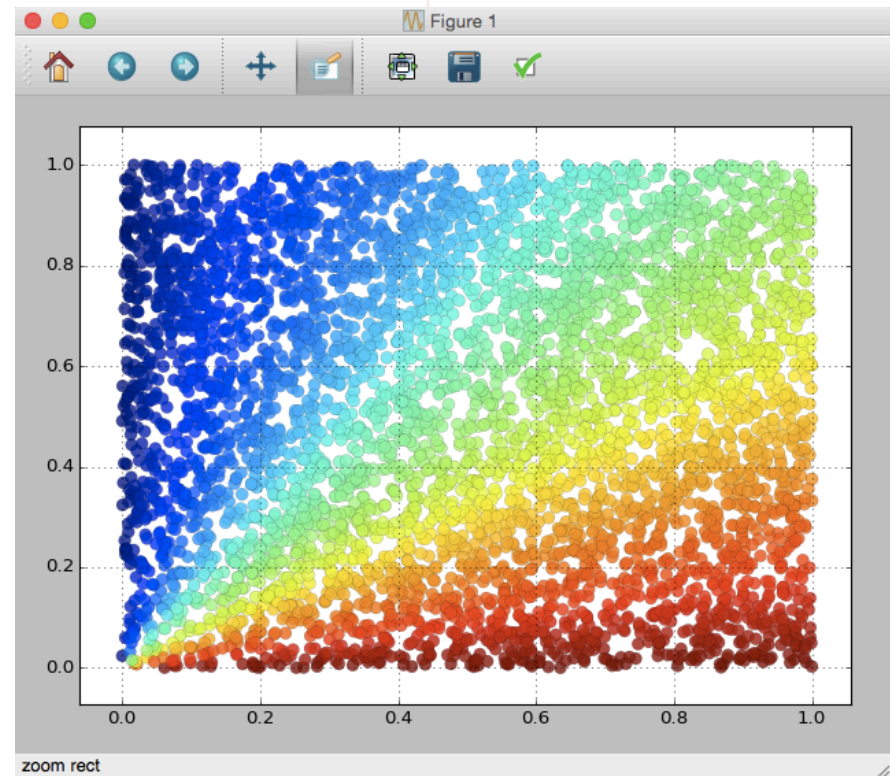




# Advanced plotting

- Advanced plotting: *scatter plot*

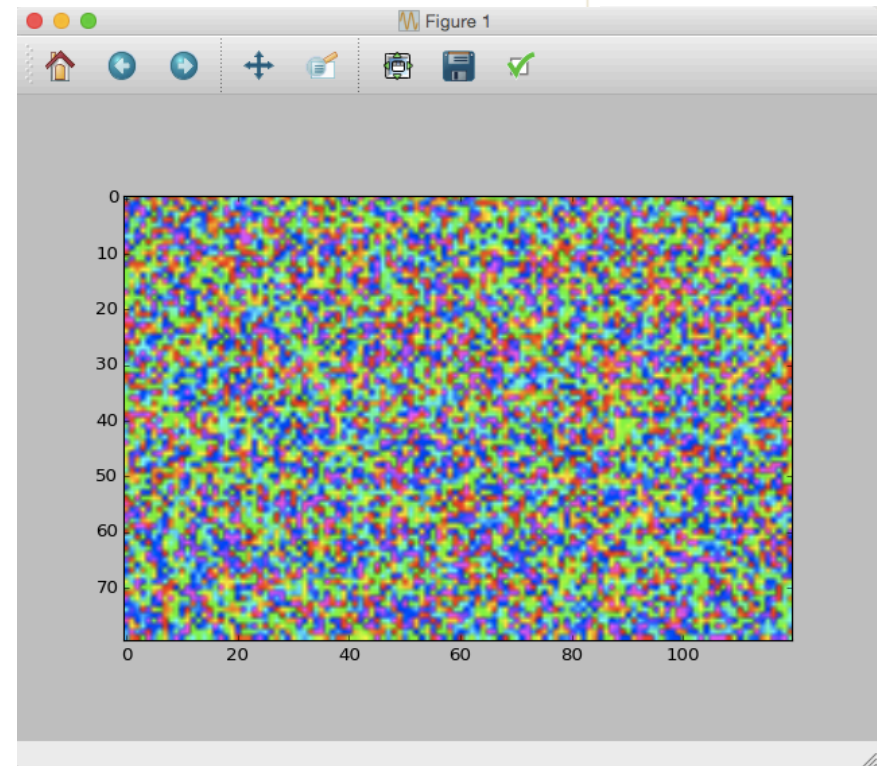
```
312 ## Scatter plot:
313 import pylab as plb
314
315 x = plb.rand(1,2,1500)
316 y = plb.rand(1,2,1500)
317 plb.axes([0.075, 0.075, .88, .88])
318
319 plb.cla() # clear the current axis
320 plb.scatter(x, y, s=65, alpha=.75, linewidth=.125,
321            c=plb.arctan2(x, y))
322
323 plb.grid(True)
324 plb.xlim(-0.085,1.085), plb.ylim(-0.085,1.085)
325 plb.pause(1)
```



# Advanced plotting

- Advanced plotting: *image plot*

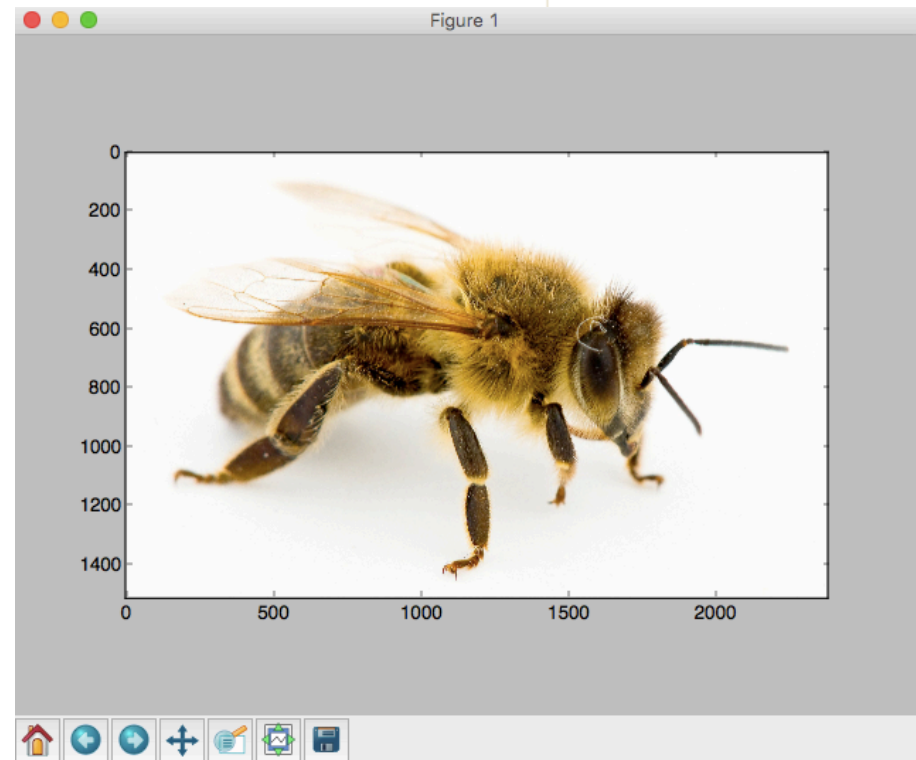
```
358 ## Image 1/2:  
359 plb.cla()  
360 array = plb.random((80, 120))  
361 plb.imshow(array, cmap=plb.cm.gist_rainbow) # with a specific colormap  
362 plb.pause(1)
```



# Advanced plotting

- Advanced plotting: *image plot*

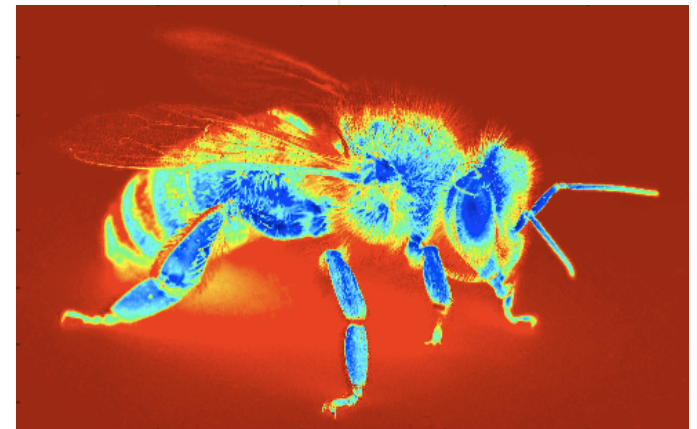
```
364 ## Image 2/2:  
365 import matplotlib.image as img  
366 import matplotlib.pyplot as plt  
367  
368 image = img.imread('files/lecture5/bee.jpg')  
369 plt.imshow(image)  
370 plt.pause(1)
```



# Advanced plotting

- Advanced plotting: *image plot 1/3*

```
372 # luminosity display using 1-channel only (no RGB color).
373 # A default colormap (lookup tabel = LUT) is applied called 'jet':
374 luminosity = image[:, :, 0]
375 plt.imshow(luminosity)
376 plt.pause(5)
377
378 # Other colormaps can be:
379 plt.imshow(luminosity, cmap="hot")
380 plt.pause(5)
381 plt.imshow(luminosity, cmap="spectral")
382 plt.pause(5)
```



# Advanced plotting

- Advanced plotting: *image plot 2/3*

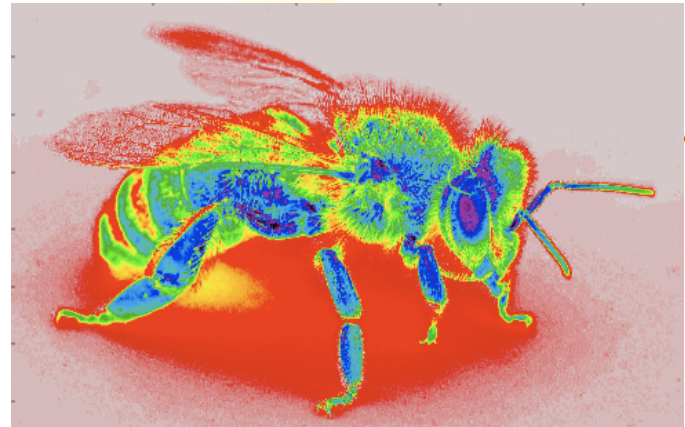
```
372 # luminosity display using 1-channel only (no RGB color).
373 # A default colormap (lookup tabel = LUT) is applied called 'jet':
374 luminosity = image[:, :, 0]
375 plt.imshow(luminosity)
376 plt.pause(5)
377
378 # Other colormaps can be:
379 plt.imshow(luminosity, cmap="hot")
380 plt.pause(5)
381 plt.imshow(luminosity, cmap="spectral")
382 plt.pause(5)
```



# Advanced plotting

- Advanced plotting: *image plot 3/3*

```
372 # luminosity display using 1-channel only (no RGB color).
373 # A default colormap (lookup tabel = LUT) is applied called 'jet':
374 luminosity = image[:, :, 0]
375 plt.imshow(luminosity)
376 plt.pause(5)
377
378 # Other colormaps can be:
379 plt.imshow(luminosity, cmap="hot")
380 plt.pause(5)
381 plt.imshow(luminosity, cmap="spectral")
382 plt.pause(5)
```

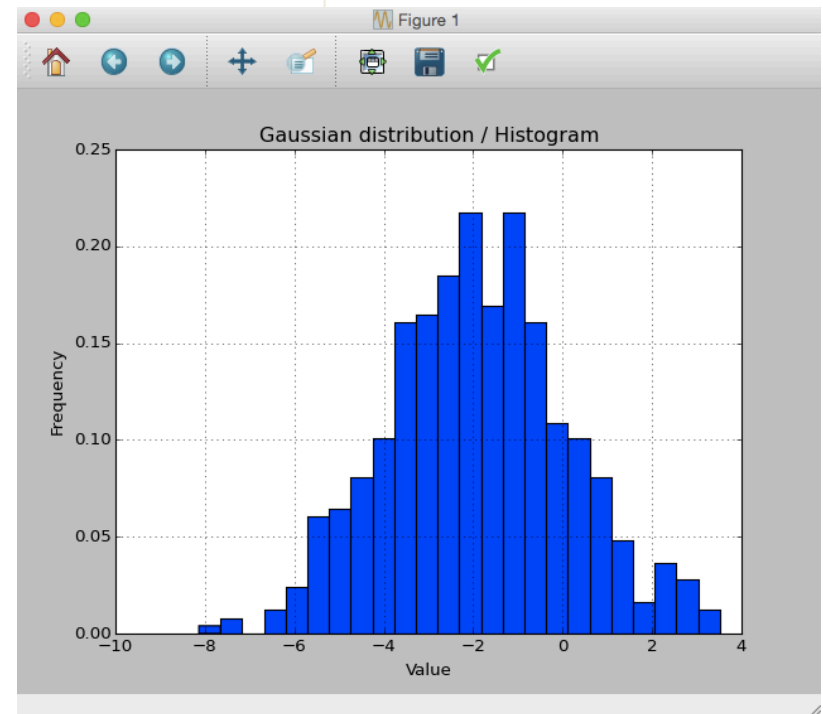


# Advanced plotting

- Advanced plotting: *histogram plot 1/2*

```
285 # Histogram 1/2:
286 import pylab as plb
287
288 plb.figure(1)
289 gaus_dist = plb.normal(-2,2,size=512) # create a random floating point vector
290
291 # plot the histogram with specific: bin number
292 plb.hist(gaus_dist, normed=True, bins=24) # default: bins=10, color='blue'
293
294 plb.title("Gaussian distribution / Histogram")
295 plb.xlabel("Value")
296 plb.ylabel("Frequency")
297 plb.grid(True)
298 plb.show()
```

Histogram is **great** for visualizing **statistical distribution** of a set of variables in a given pool of samples, divided into classes called bins



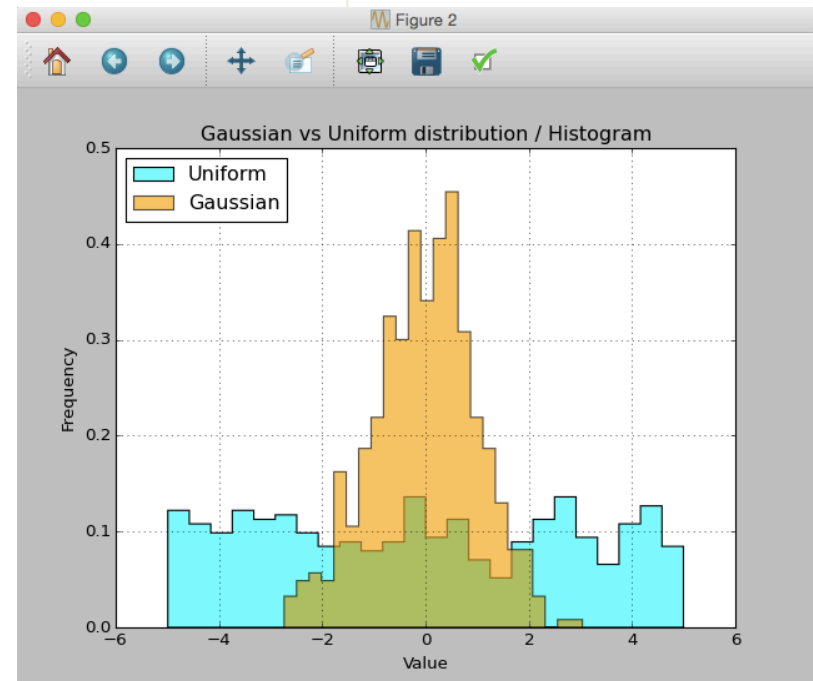


# Advanced plotting

- Advanced plotting: *histogram plot 2/2*

```
300 # Histogram 2/2:
301 plb.figure(2)
302 gaus_dist = plb.normal(size=512)
303 unif_dist = plb.uniform(-5,5,size=512) # create uniform distribution vector
304
305 # plot the histogram with specific: bin number, color, transparency, label
306 plb.hist(unif_dist, bins=24, histtype='stepfilled',
307         normed=True, color='cyan', label='Uniform')
308 plb.hist(gaus_dist, bins=24, histtype='stepfilled',
309         normed=True, color='orange', label='Gaussian', alpha=0.65)
310
311 plb.legend(loc='upper left')
312 plb.title("Gaussian vs Uniform distribution / Histogram")
313 plb.xlabel("Value")
314 plb.ylabel("Frequency")
315 plb.grid(True)
316 plb.show()
```

Histogram is **great** for visualizing **statistical distribution** of a set of variables in a given pool of samples, divided into classes called bins

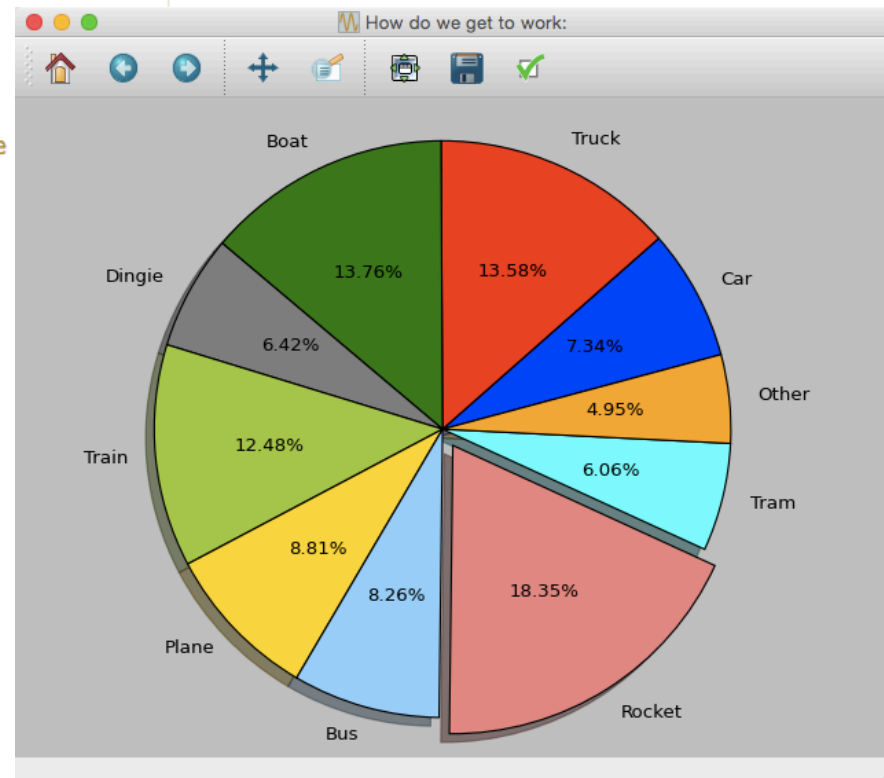




# Advanced plotting

- Advanced plotting: *pie chart*

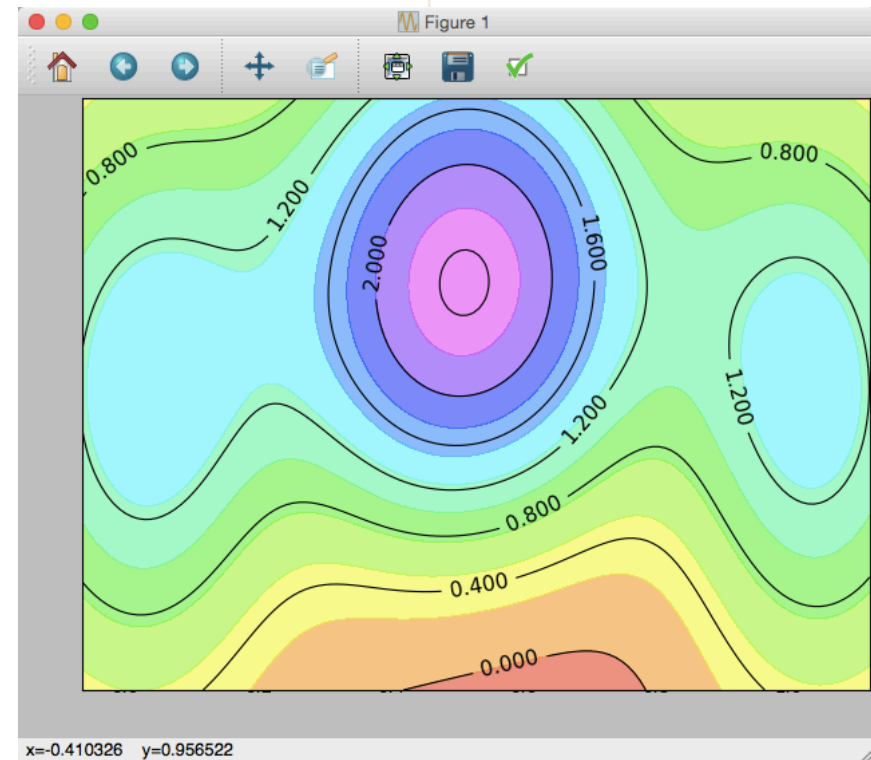
```
318 # Pie chart:
319 plb.figure('How do we get to work:')
320 plb.axes([0.035, 0.035, 0.9, 0.9])
321
322 l = 'Car', 'Truck', 'Boat', 'Dingie', 'Train', 'Plane', 'Bus', 'Rocket', 'Tram', 'Other'
323 b = plb.round_(plb.random(10), decimals=2)
324 c = ['blue', 'red', 'green', 'gray', 'yellowgreen',
325      'gold', 'lightskyblue', 'lightcoral', 'cyan', 'orange']
326 e = (0, 0, 0, 0, 0, 0, 0, 0.05, 0, 0) # 'explode' the 8th slice only - 'Rocket'
327
328 plb.cla()
329 plb.pie(b, explode = e, labels=l, colors=c, radius=.75,
330        autopct='%1.2f%%', shadow=True, startangle=15)
331
332 # we set the aspect ratio to 'equal' so the pie is drawn in a circle
333 plb.axis('equal')
334 plb.xticks(()); plb.yticks(())
335 plb.show()
```



# Advanced plotting

- Advanced plotting: *contour plot*

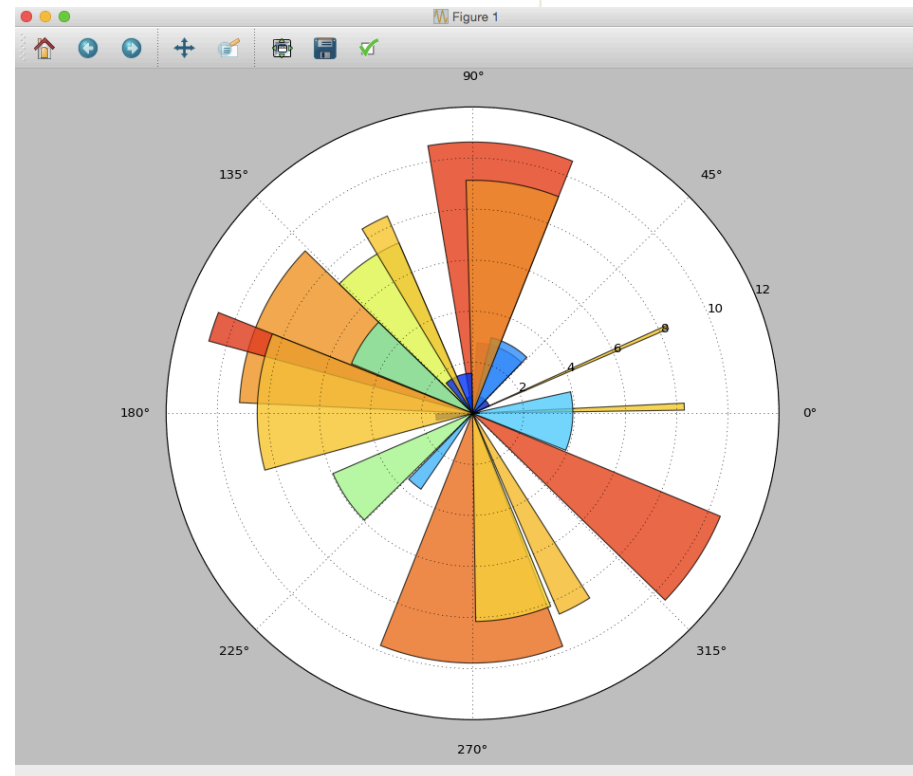
```
337 # Contour plot:
338 import pylab as plb
339
340 def f(x,y):
341     return (2 - x/3 + x**6 + 2.125*y) * plb.exp(-x**2 -y**2)
342
343 n = 128
344 x = plb.linspace(-2, 2, n)
345 y = plb.linspace(-1, 1, n)
346 A,B = plb.meshgrid(x, y)
347
348 plb.cla()
349 plb.axes([0.075, 0.075, 0.92, 0.92])
350
351 plb.contourf(A, B, f(A, B), 12, alpha=.50,
352             cmap=plb.cm.gist_rainbow)
353 C = plb.contour(A, B, f(A, B), 8, colors='black',
354               linewidth=.65)
355
356 plb.clabel(C, inline=1, fontsize=14)
357 plb.xticks(); plb.yticks()
358 plb.show()
```



# Advanced plotting

- Advanced plotting: *polar plot*

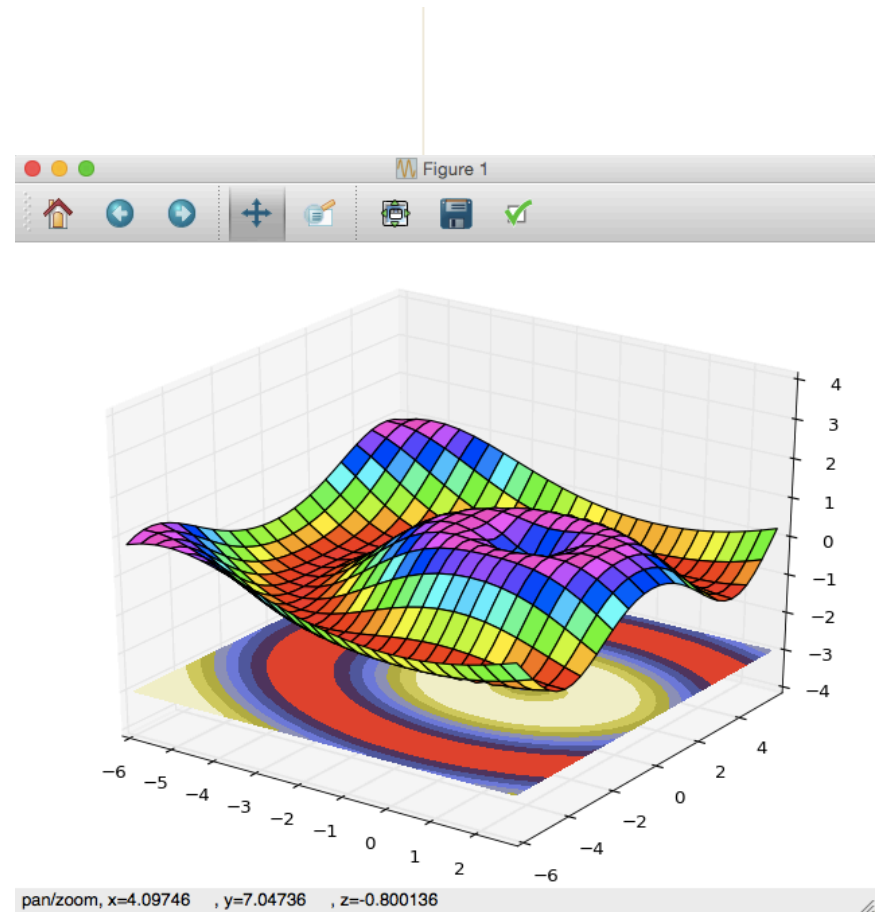
```
360 # Polar plot:
361 import pylab as plb
362
363 plb.axes([0.065, 0.065, 0.88, 0.88], polar=True)
364
365 q = 24
366 t = plb.arange(0.015, 3*plb.pi, 3 * plb.pi / q)
367 rad = 12 * plb.rand(q)
368 w = plb.pi / 4 * plb.rand(q)
369 ba = plb.bar(t, rad, width = w)
370
371 for r,bar in zip(rad, ba):
372     bar.set_facecolor(plb.cm.jet(r/12.))
373     bar.set_alpha(0.75)
374
375 plb.show()
```



# Advanced plotting

- Advanced plotting: *3-D plot*

```
377 # 3-D plot:
378 import pylab as plb
379 from mpl_toolkits.mplot3d import Axes3D
380
381 ax = Axes3D(plb.figure())
382 x = plb.arange(-6, 3, 0.35)
383 y = plb.arange(-6, 6, 0.35)
384 x, y = plb.meshgrid(x, y)
385 k = plb.sqrt(x**2 + y**2)
386 z = plb.sin(k)
387
388 ax.plot_surface(x, y, z, rstride=2, cstride=1,
389               cmap=plb.cm.gist_rainbow)
390 ax.contourf(x, y, z, zdir='z', offset=-3,
391            cmap=plb.cm.gist_stern)
392 ax.set_zlim(-4, 4)
393
394 plb.show()
```



# Discussion

*Important:*

- - **dictionary keys are not sorted in Python**: Standard Python dictionaries are **NOT** ordered. Even if you sorted the (key,value) pairs, preserving the order in **dict** is not possible. However, to get the order numerically or alphabetically (1<sup>st</sup> letter) based on the keys use the following:

```
In [1]: import collections
```

```
In [2]: a = {2:3, 1:89, 4:5, 3:0}
```

```
In [3]: ord = collections.OrderedDict(sorted(a.items()))
```

```
In [4]: ord
```

```
Out[4]: OrderedDict([(1, 89), (2, 3), (3, 0), (4, 5)])
```

- - **to access dictionary keys if they are numbers and were assigned to a dict\_key variable**:  
the **dict\_values** object does not support indexing, The method **dict().values()** is quite different between python 3 and python 2. In python 2 it returns a list, and in python 3 it doesn't. To fix this we cast to list:

```
In [41]: d = {1: 89, 2: 3, 3: 0, 4: 5}
```

```
In [42]: h = d.values()
```

```
In [43]: h
```

```
Out[43]: dict_values([89, 3, 0, 5])
```

```
In [44]: list(h)[0]
```

```
Out[44]: 89
```

# Discussion

- What is a moving average?
  - In general it is a **series of averages** of **different interval subsets** of data points **out of a full data set**
  - Simple Moving Average (**SMA**) – is defined as **unweighted mean** of the previous data. Used in stock market to calculate the average closing price of a stock over specific time interval
  - Exponential Moving Average (**EMA**) – a.k.a. exponentially weighted moving average (EWMA), applies **weighting factors which decrease exponentially** and is a type of an IIR filter where the weighting for each older data decreases exponentially, but never reaching zero or:  $(0 < \alpha \leq 1]$
  - Cumulative Moving Average (**CMA**) – is an **average of all presented data up to the current point**
  - Weighted Moving Average (**WMA**) – it has multiplying factors serving as **different weights to the data** at different positions of a given sample frame
  - Modified Moving Average (**MMA**) – or running moving average (RMA), or smoothed moving average (SMMA) is like an **exponential moving average**, but with different degree of weighting decrease  $\alpha$

# Class exercise

1. Import `paylab`
2. Create a dictionary with 5 keys and empty values in A
3. Using a function, assign random values to each key between [0:10], using a `for` loop and return the result in B
4. Using a function, change the value of any member in B that is less than 5 with the result from (4.1) (consider using an `if` statement in a loop):
  - 4.1 Using normal distribution with mean = 2 and std = 3 create an array of size 256 points
  - 4.2 Using a histogram with 12 bins, plot the result from 4.1 with a pause of 1 sec. Use proper labeling (figure, title, labels, legend, grid, etc.)
5. Assign the result from 4. in C
6. Update one of the keys in C with another using the `pop` feature
7. Update another key in C manually (add the new one and delete the old one)
8. Compare A, B and C using a short conditional expression