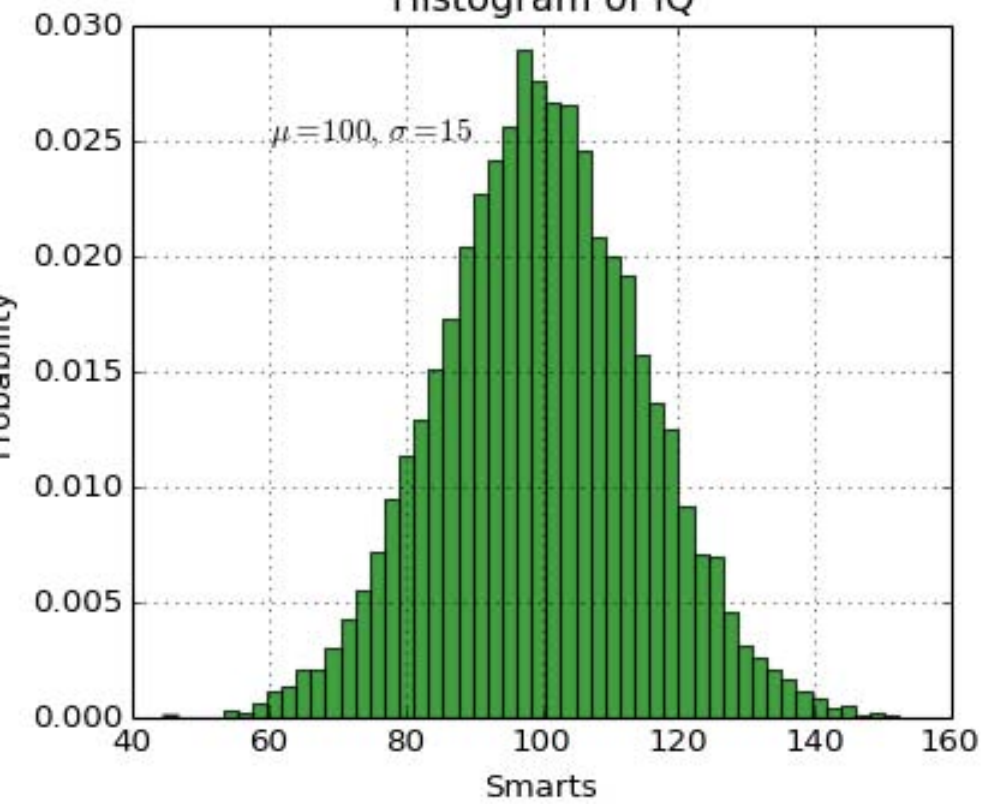
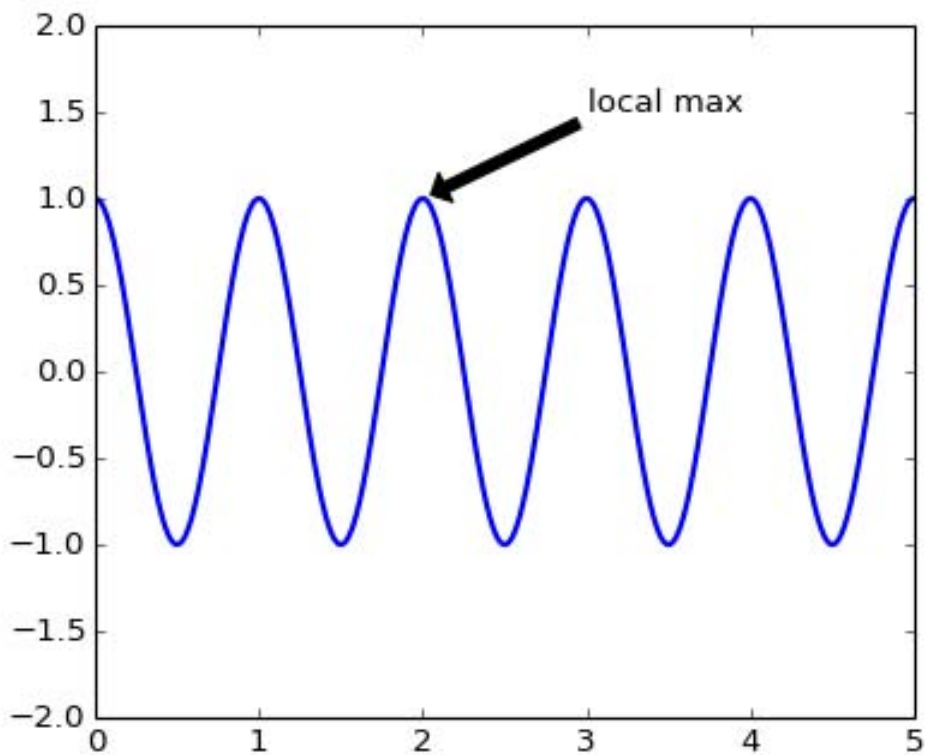
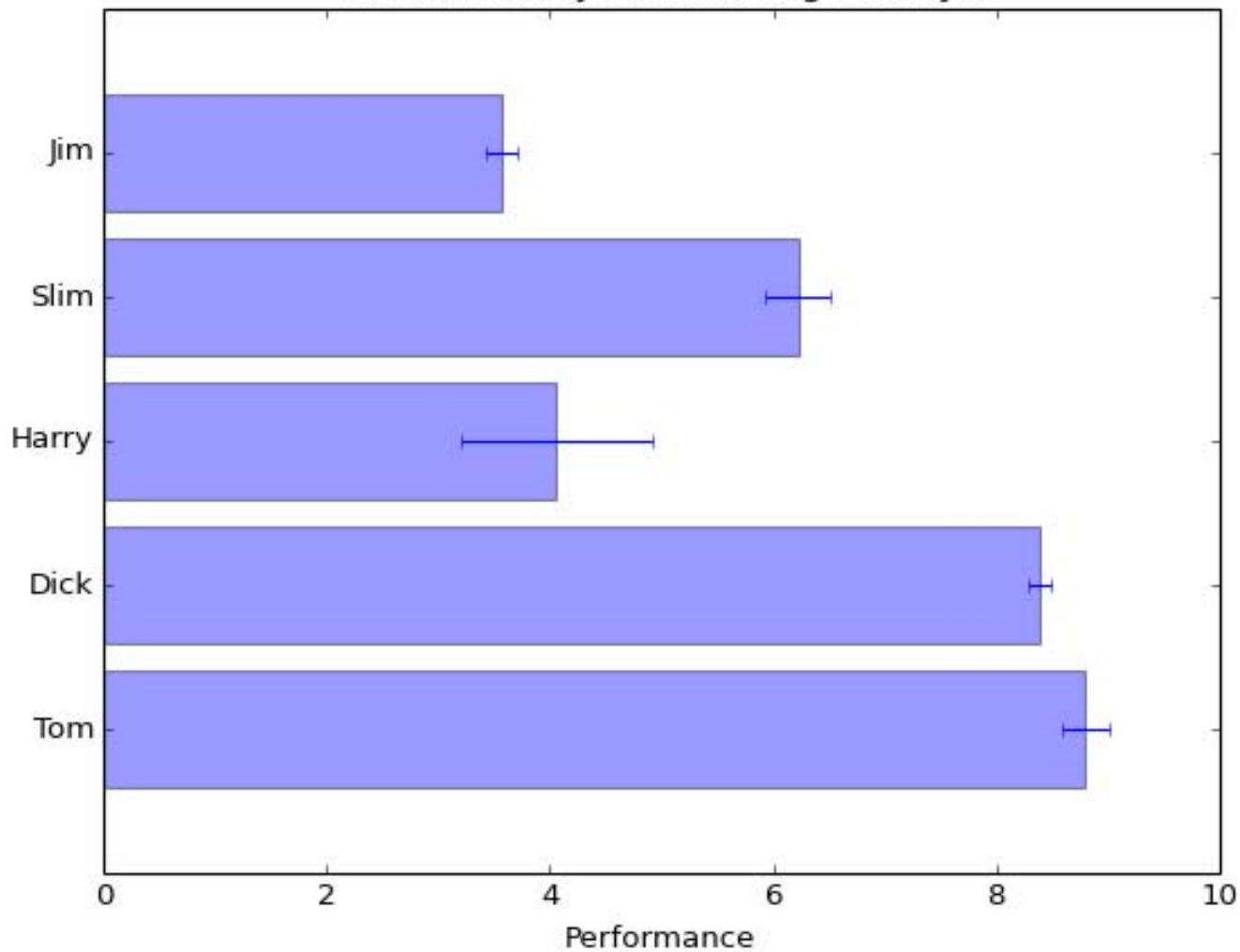


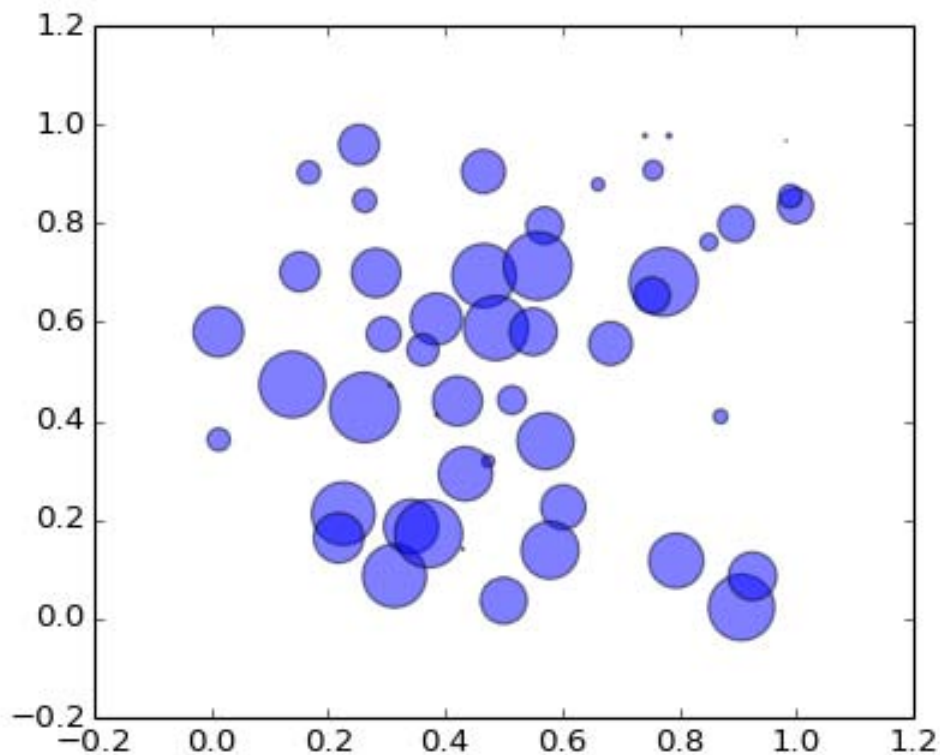
Histogram of IQ

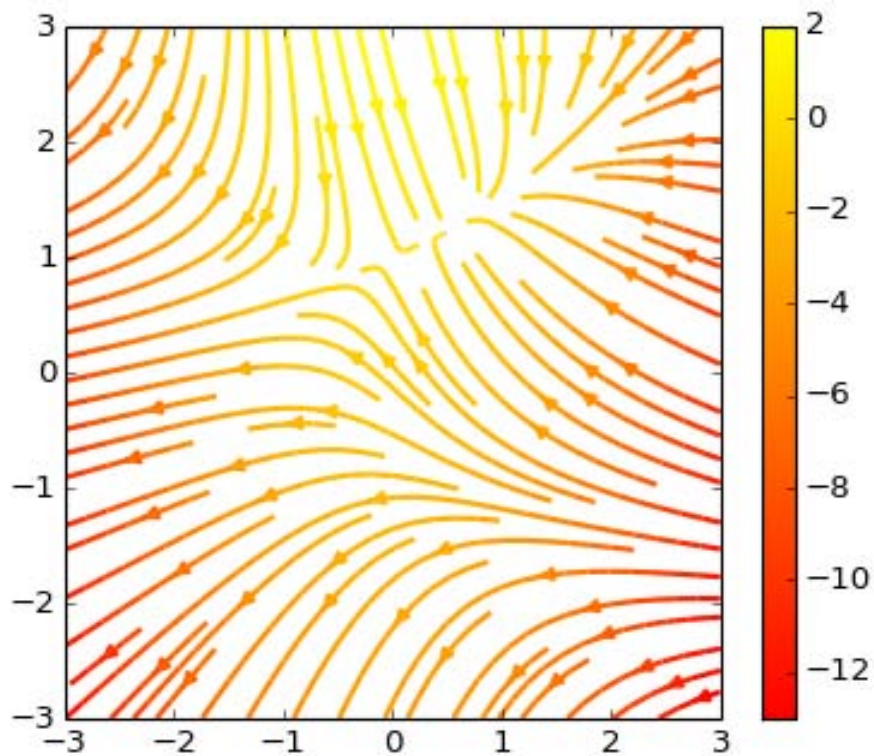


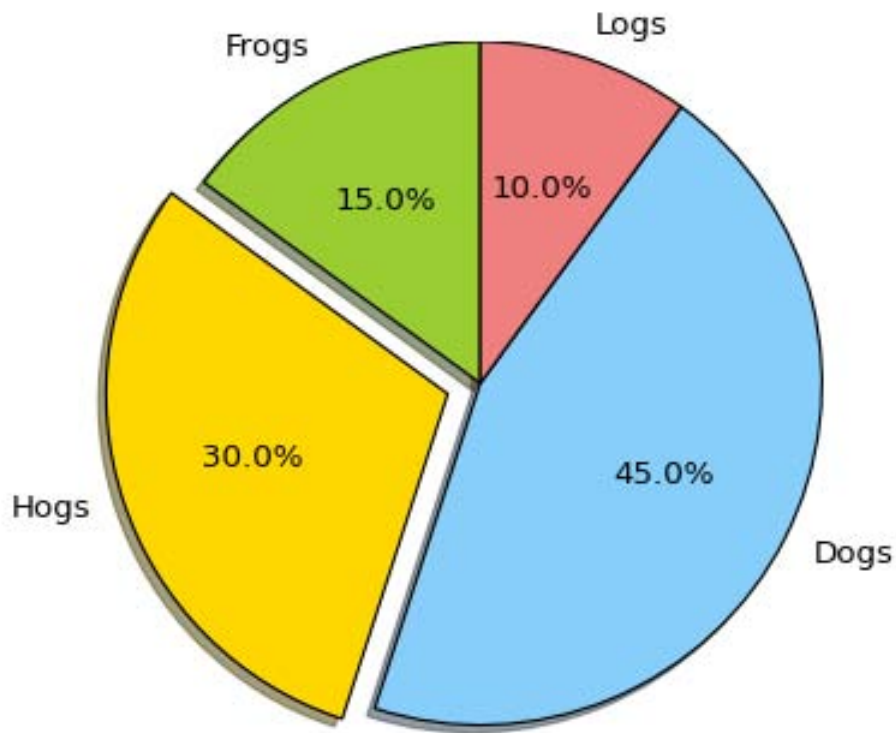


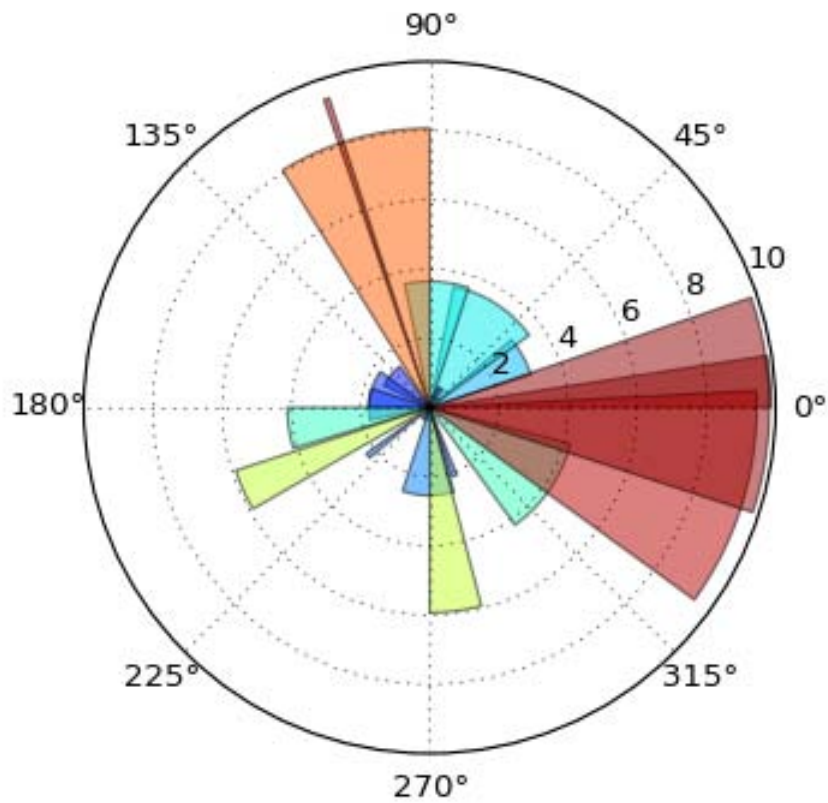
How fast do you want to go today?



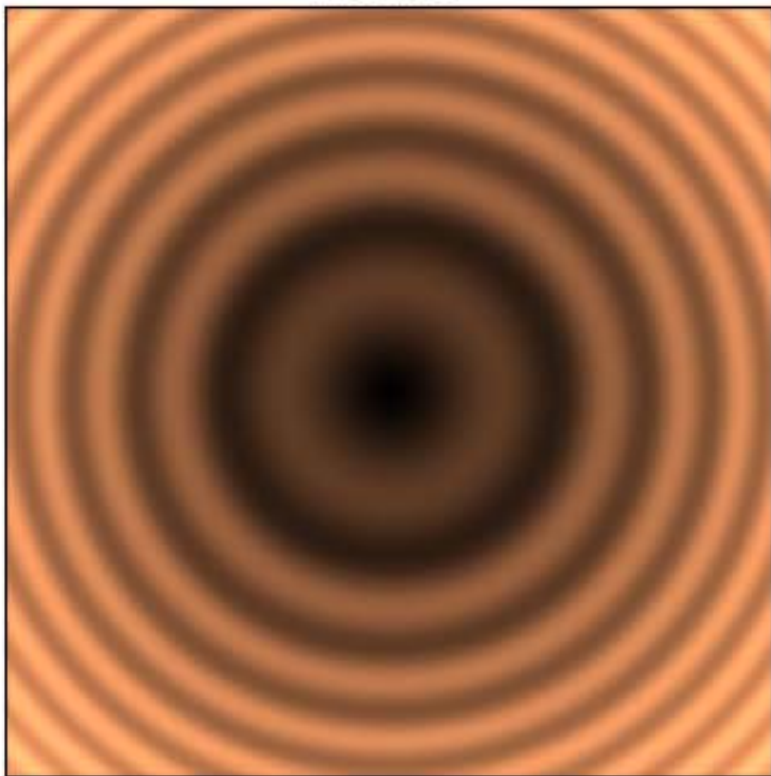




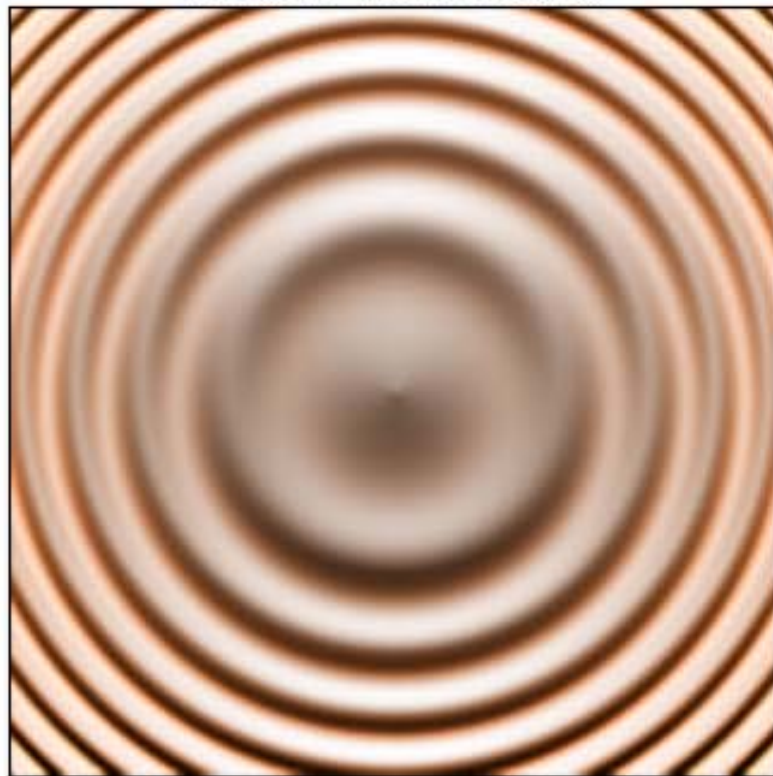


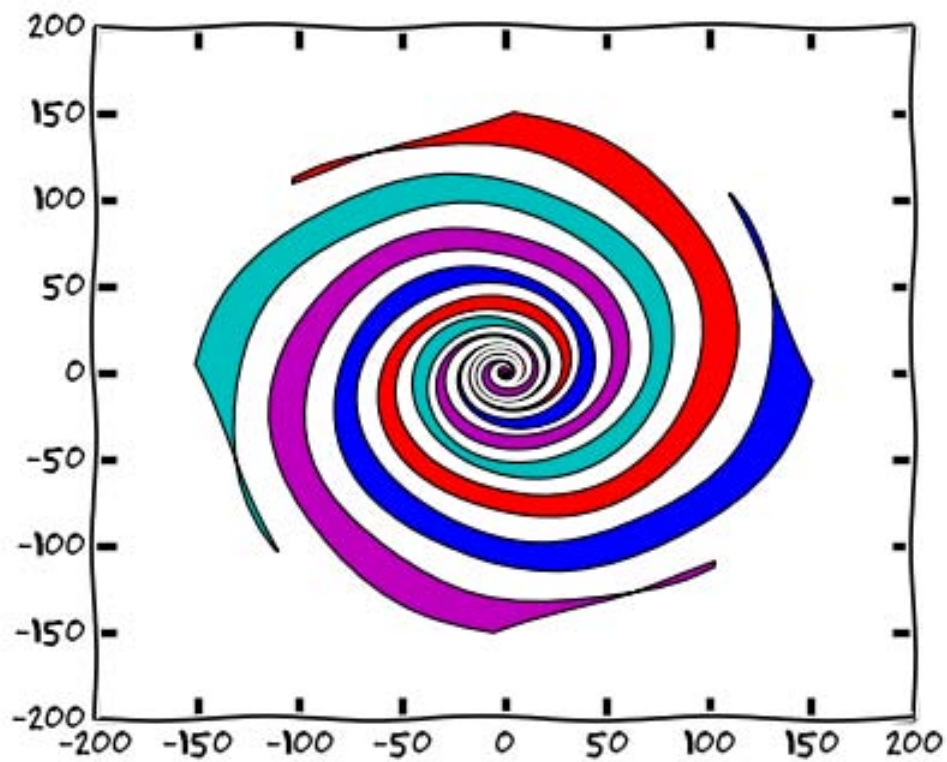


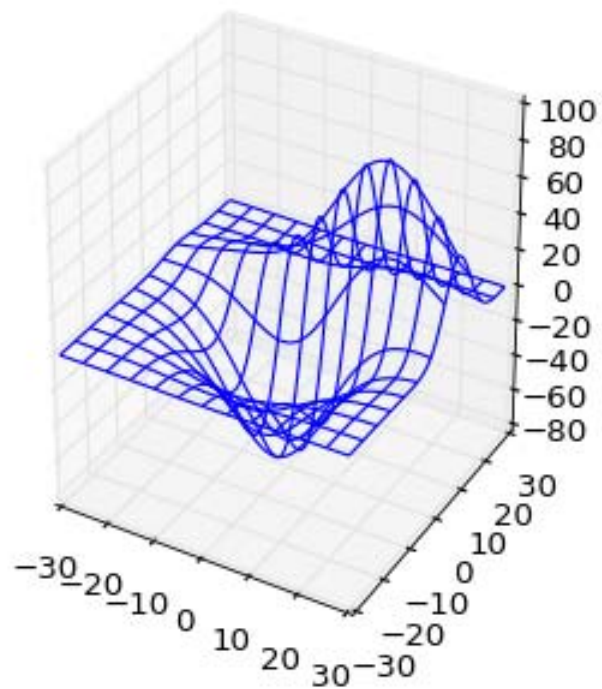
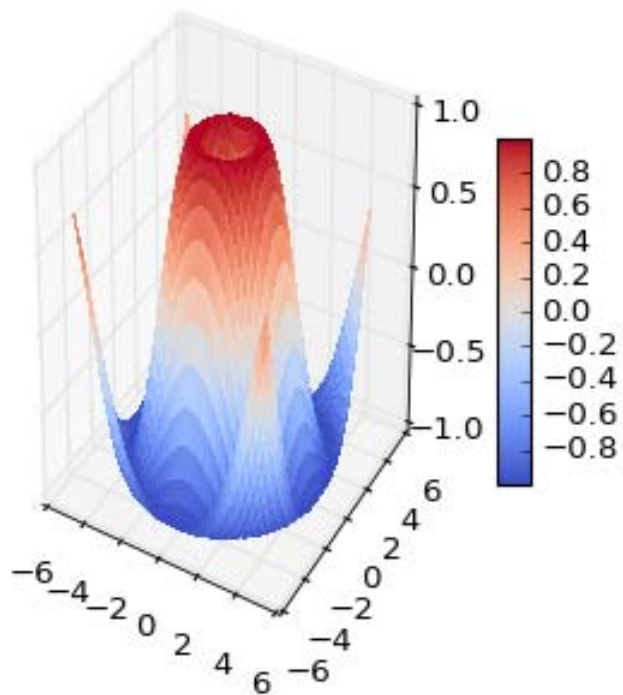
imshow

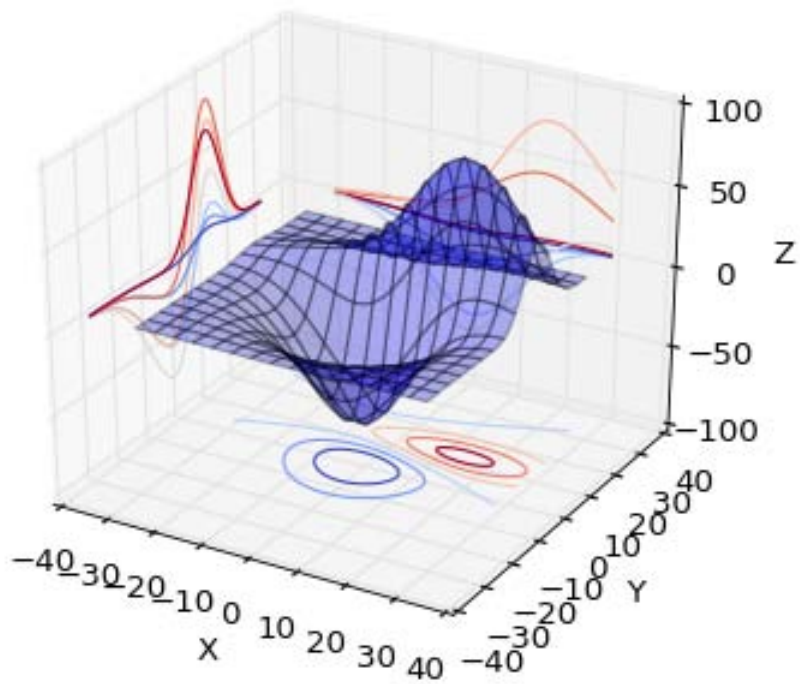


imshow with shading











IP[y]: Notebook FirstNotebook (unsaved changes)

File Edit View Insert Cell Kernel Help

Code Cell Toolbar: None

In [1]: 2+3

Out[1]: 5

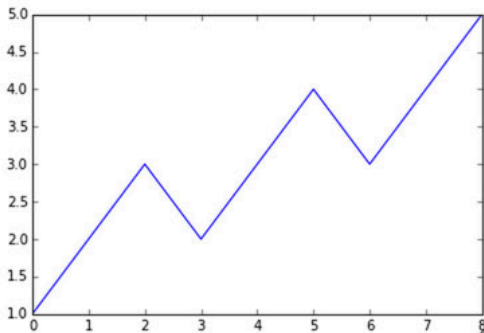
In [2]: import numpy as np
import matplotlib.pyplot as plt

In [3]: np.sin(np.pi/6)

Out[3]: 0.49999999999999994

In [4]: plt.plot([1,2,3,2,3,4,3,4,5])

Out[4]: [<matplotlib.lines.Line2D at 0x10cfc8b90>]



In []:

IP[y]: Notebook

spectrogram

Last saved: Mar 07 11:14 PM

File Edit View Insert Cell Kernel Help

Markdown

Simple spectral analysis

An illustration of the [Discrete Fourier Transform](#)

$$X_k = \sum_{n=0}^{N-1} x_n e^{-\frac{2\pi i}{N} kn} \quad k = 0, \dots, N-1$$

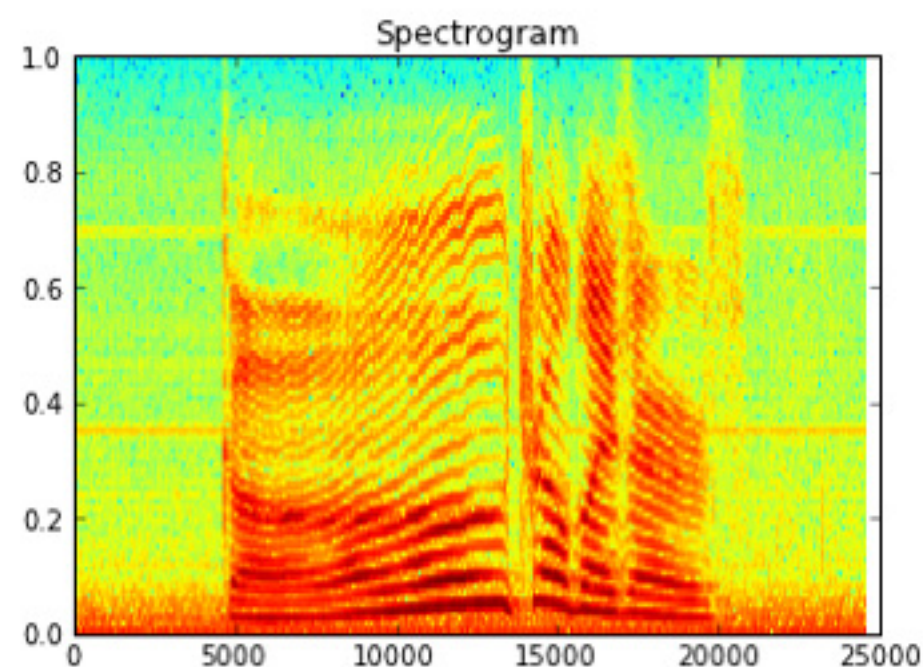
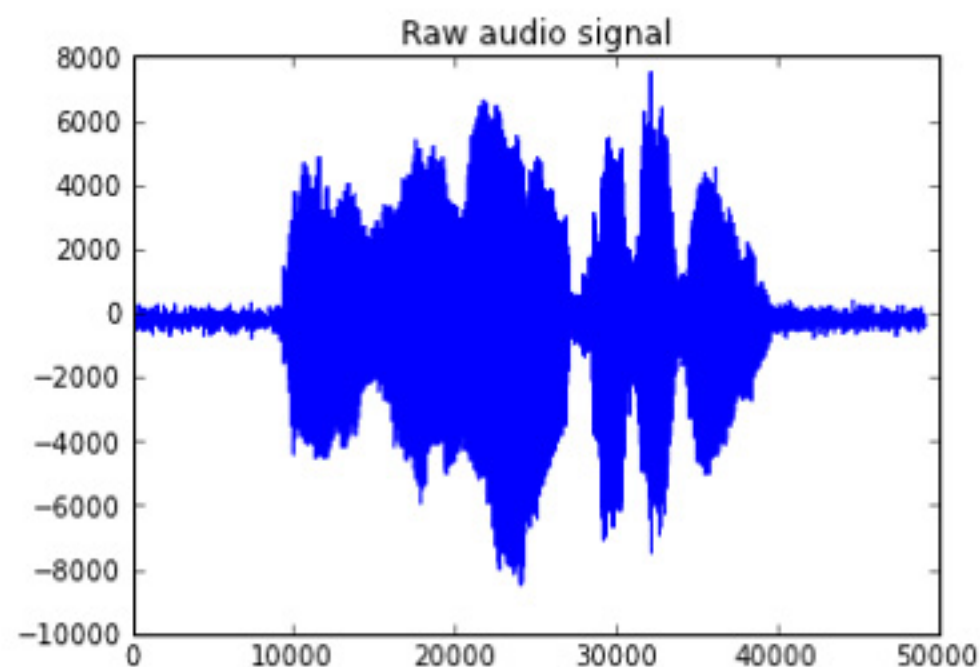
using windowing, to reveal the frequency content of a sound signal.

We begin by loading a datafile using SciPy's audio file support:

```
In [1]: from scipy.io import wavfile
rate, x = wavfile.read('test_mono.wav')
```

And we can easily view its spectral structure using matplotlib's builtin specgram routine:

```
In [2]: fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(12, 4))
ax1.plot(x); ax1.set_title('Raw audio signal')
ax2.specgram(x); ax2.set_title('Spectrogram');
```



```
In [5]: interact(plot_random_graph, n=(2,30), m=(1,10), k=(1,10), p=(0.0, 1.0, 0.001),
generator={'lobster': random_lobster,
'power law': powerlaw_cluster,
'Newman-Watts-Strogatz': newman_watts_strogatz,
u'Erdős-Rényi': erdos_renyi,
});
```

x

n m k p

generator

Erdős-Rényi

