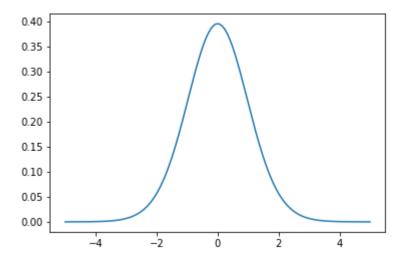
1. Redo the problem 4 in HW1 in python (use only the distribution objects, like norm, binom), make sure you print conclusions from two methods(using clssic statistics method and p-value)	

```
In [25]: import numpy as np
         import scipy as sp
         #from scipy.stats import ttest ind
         #from scipy.stats import norm
         #from scipy.stats import binom
         from scipy import stats
         N = 100.0
         second = 28.0
         first = 0.37
         alpha = 0.01
         z = ((second/N) - first) * np.sqrt(N / (first*(1-first)))
         z = round(z, 2)
         \#pcomp = sp.stats.t.sf(np.abs(z), (first*(1-first)-1)*2
         print 'z = ', z
         #Studnt, Single tail
         #Should be -2.33
         cv = -stats.t.ppf(1-alpha, 999)
         cv = round(cv, 2)
         if z < cv:
              print z, ' < ', cv</pre>
              print 'Reject the null hypothesis'
         else: #should print this
              print z, ' > ', cv
              print 'Fail to reject the null hypothesis'
              print 'There is not enough evidence to conclude that the cars have improve
         d'
         print
         p = sp.stats.norm.cdf(z)
         p = round(p, 3)
         print 'p-value = ', p
         if p < alpha:</pre>
              print p, ' < ', alpha</pre>
              print 'We reject the null hypothesis'
         else: #should print this
              print p, ' > ', alpha
              print 'We fail to reject the null hypothesis'
              print 'There is not enough evidence to conclude that the cars have improve
         d'
         z = -1.86
         -1.86 > -2.33
         Fail to reject the null hypothesis
         There is not enough evidence to conclude that the cars have improved
         p-value = 0.031
         0.031 > 0.01
         We fail to reject the null hypothesis
         There is not enough evidence to conclude that the cars have improved
```

2. Plot the pdf of a t distribution with degree of freedom 30

```
In [26]: #sample size = 31
# plot the t-distribution pdf
from numpy import arange
from matplotlib import pyplot
from scipy.stats import t
# define the distribution parameters
sample_space = arange(-5, 5, 0.001)
#dof = len(sample_space) - 1
dof = 30
# calculate the pdf
pdf = t.pdf(sample_space, dof)
# plot
pyplot.plot(sample_space, pdf)
pyplot.show()
```



3.Anomaly detection Assume we have a datasets, it represents the daily number of clicks of a small website for 3 months. We assume it follows a poisson distribution. We also assume the the data that exceeds the 95% confidence interval of the sample mean is not likely to happen. We consider them as anomalies. Write a function to extract those anomalies:

```
In [89]: def anomaly_detection_poisson(values):
             anomalies = []
             margin = 1.96 #margin of error associated with 95% confidence interval
             counter = 0
             num = len(values)
             mean = np.mean(values)
             std = np.std(values)
             se = std / (np.sqrt(num))
             upper = mean + (margin * se)
             lower = mean - (margin * se)
             for i in values:
                  if i/1. > upper or i/1. < lower:
                      anomalies.append(i)
                      counter += 1
             #print counter
             return anomalies
         #748, 863
```

The following is a test function to test your detection function. If your function is correct, you should get the following result: [731, 745, 868, 713, 863, 877, 867, 875, 867, 748, 900, 747, 717]

```
In [90]: value=[796, 756, 827, 779, 787, 731, 847, 829, 794, 790, 831, 820, 822, 792, 7
67, 757, 777, 775, 799, 756, 849, 808, 800, 823, 798, 809, 783, 800, 848, 772,
853, 759, 745, 868, 713, 863, 819, 829, 799, 777, 785, 812, 877, 839, 783, 867
, 875, 829, 818, 813, 763, 829, 812, 867, 790, 748, 822, 774, 900, 830, 781, 7
95, 825, 799, 795, 839, 840, 765, 781, 781, 797, 821, 852, 836, 811, 771, 800,
752, 776, 755, 798, 839, 821, 794, 846, 834, 825, 825, 830, 814, 839, 760, 786
, 747, 803, 717, 801, 819, 789, 824, 835, 806, 858, 774, 848, 793, 826, 796, 7
98, 773, 779, 775, 779, 806, 768, 787, 788, 822, 843, 780]
print anomaly_detection_poisson(value)
```

[796, 756, 827, 779, 787, 731, 847, 829, 794, 790, 831, 820, 822, 792, 767, 757, 777, 775, 756, 849, 823, 783, 848, 772, 853, 759, 745, 868, 713, 863, 819, 829, 777, 785, 812, 877, 839, 783, 867, 875, 829, 818, 813, 763, 829, 812, 867, 790, 748, 822, 774, 900, 830, 781, 795, 825, 795, 839, 840, 765, 781, 781, 797, 821, 852, 836, 811, 771, 752, 776, 755, 839, 821, 794, 846, 834, 825, 825, 830, 814, 839, 760, 786, 747, 717, 819, 789, 824, 835, 858, 774, 848, 793, 826, 796, 773, 779, 775, 779, 768, 787, 788, 822, 843, 780]