

In [1]:

```
%pylab
%matplotlib inline
```

Using matplotlib backend: Qt4Agg

Populating the interactive namespace from numpy and matplotlib

In [2]:

```
import pandas as pd

#from IPython.display import display
#from IPython.core.pylabtools import figsize, getfigs
import scipy.stats
```

Se añade la carpeta 'funciones' al PYTHONPATH

In [3]:

```
import sys, os
#sys.path.append('../funciones')

#import inspect
# realpath() will make your script run, even if you symlink it :)
#cmd_folder = os.path.realpath(os.path.abspath(os.path.split(inspect.getfile( inspect.currentframe() ))
[0]))
cmd_folder = os.path.realpath('funciones')
if cmd_folder not in sys.path:
    sys.path.insert(0, cmd_folder)

from fun_bootstrap import *
from fun_plothist import *
import aeronettools as at
import aeronettools2 as at2
#import aeronettools_pynb as at3
from aeronettools_pynb import extract_aeronet_data
from fun_check import *
```

In [4]:

```
# Autocarga del modulo aeronettools_pynb
%load_ext autoreload
%aimport aeronettools_pynb
%aimport aeronettools
%aimport aeronettools2
#%aimport fun_bootstrap
%aimport fun_bootstrap
%aimport fun_plothist
%aimport fun_check
%autoreload 1
```

MURCIA 2013

Se comprueba si existe el archivo con los datos, y si no se descarga y descomprime

In [5]:

```
file_default = '/spread/pn32/Renovables/Red/Aeronet/lev15/130101_131231_Murcia.lev15'
FILE = '130101_131231_Murcia.lev15'

if os.path.isfile(file_default):
    FILE = file_default
elif not os.path.isfile(FILE):
    url = 'http://aeronet.gsfc.nasa.gov/zip_files/130101_131231_Murcia.zip'
    zFILE = FILE.replace('lev15', 'zip')

    import urllib
    urllib.URLopener().retrieve(url, zFILE)
```

```

import urllib2
#file_name = url.split('/')[ -1]
#u = urllib2.urlopen(url)
#f = open(file_name, 'wb')
#meta = u.info()
#file_size = int(meta.getheaders("Content-Length") [0])
#print "Downloading: %s Bytes: %s" % (file_name, file_size)
#
#file_size_dl = 0
#block_sz = 8192
#while True:
#    buffer = u.read(block_sz)
#    if not buffer:
#        break
#
#    file_size_dl += len(buffer)
#    f.write(buffer)
#    status = r"%10d  [%3.2f%%]" % (file_size_dl, file_size_dl * 100. / file_size)
#    status = status + chr(8)*(len(status)+1)
#    #print status,
#
#f.close()

import zipfile
zfile = zipfile.ZipFile(zFILE)
zfolder = os.path.dirname(os.path.realpath(zFILE))
zfile.extractall(zfolder)
# os.getcwd()
zfile.close()

os.remove(zFILE)

```

Filtrado de los datos con el algoritmo de AEMet llamado aeronettools.py. Se ha modificado algunos parámetros para tener datos en todos los meses del año.

In [6]:

```

file_aeronet_Murcia = FILE
df_out_Murcia = at2.extract_aeronet_data(file_aeronet_Murcia)
df_out_Murcia['month'] = df_out_Murcia.index.month

aodMen_Murcia = [df_out_Murcia[df_out_Murcia['month']==i]['AOT_500'].mean() for i in range(1,13)]
dataAODmen_Murcia = [len(df_out_Murcia[df_out_Murcia['month']==i]['AOT_500']) for i in range(1,13)]
df_Murcia = pd.DataFrame({'AODmean_Murcia':aodMen_Murcia, 'number of AOD data':dataAODmen_Murcia })
df_Murcia

```

Out[6]:

	AODmean_Murcia	number of AOD data
0	0.037084	80
1	0.056487	121
2	0.058808	100
3	0.104211	98
4	0.135081	157
5	0.160551	259
6	0.187269	338
7	0.202397	213
8	0.168832	136
9	0.135315	133
10	0.048316	148
11	0.095019	80

Representación de los datos

In [7]:

```

pylab.figure(figsize=(17,5))
ax = plt.subplot(121)
#ax = df_Murcia.plot(y = 'AODmean_Murcia', marker='.')
ax.plot(df_Murcia.index.values, df_Murcia['AODmean_Murcia'], marker='.')
ax2 = df_Murcia['number of AOD data'].plot(kind = 'bar', secondary_y=True, color = 'white',
align='center')
#ax2 =df_Murcia.plot(y = 'number of AOD data', secondary_y=True)

ax.set_ylabel("AOD monthly mean", fontsize=12)
ax2.set_ylabel("number of AOD data", fontsize=12)

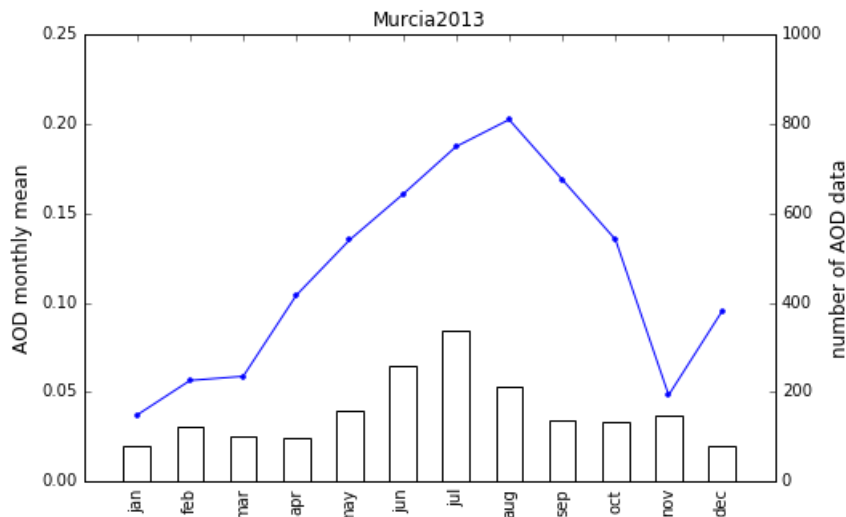
#ax.set_ylim(-150,200)
width = 0.3
ind = np.arange(len(dataAODmen_Murcia))
#rects1 = ax2.bar(ind, dataAODmen_Murcia, width, color='white', align='center')
xTickMarks = ['jan', 'feb', 'mar', 'apr', 'may', 'jun', 'jul', 'aug', 'sep',
'oct', 'nov', 'dec']
ax.set_xticks(ind)
ax.set_ylim(0,0.25)
ax2.set_ylim(0,1000)
xtickNames = ax.set_xticklabels(xTickMarks)
ax.set_xlim(-1,12)
ax2.set_xlim(-1,12)

#bar_label(rects1, dataAODmen_Murcia)
ax.set_title('Murcia2013')

```

Out[7]:

<matplotlib.text.Text at 0x1d9a0ba8>



In [8]:

```

x = df_out_Murcia['AOT_500'].values

# find mean 95% CI and 10,000 bootstrap samples
stat_mean, ci_mean = bootstrap(x, np.mean)
obs_mean = np.mean(x)

print "Mean of sample data: \n", obs_mean
pylab.figure(figsize=(8,4))
#pylab.figure(figsize=(18,4))
#ab = pylab.subplot(121)
plot_bootstrap(stat_mean, ci_mean)
pylab.axvline(obs_mean, c='black')
#ab.set_title('Murcia2013')

```

Mean of sample data:

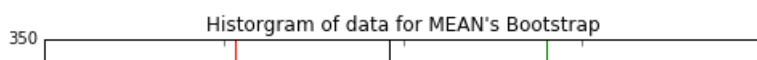
0.134623338969

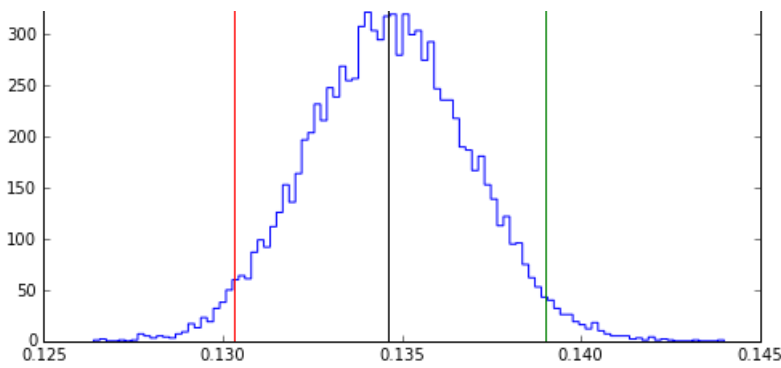
Bootstrapped 95% confidence interval of Mean:

[0.13030805179817498, 0.13902493585614598]

Out[8]:

<matplotlib.lines.Line2D at 0x1dad198>



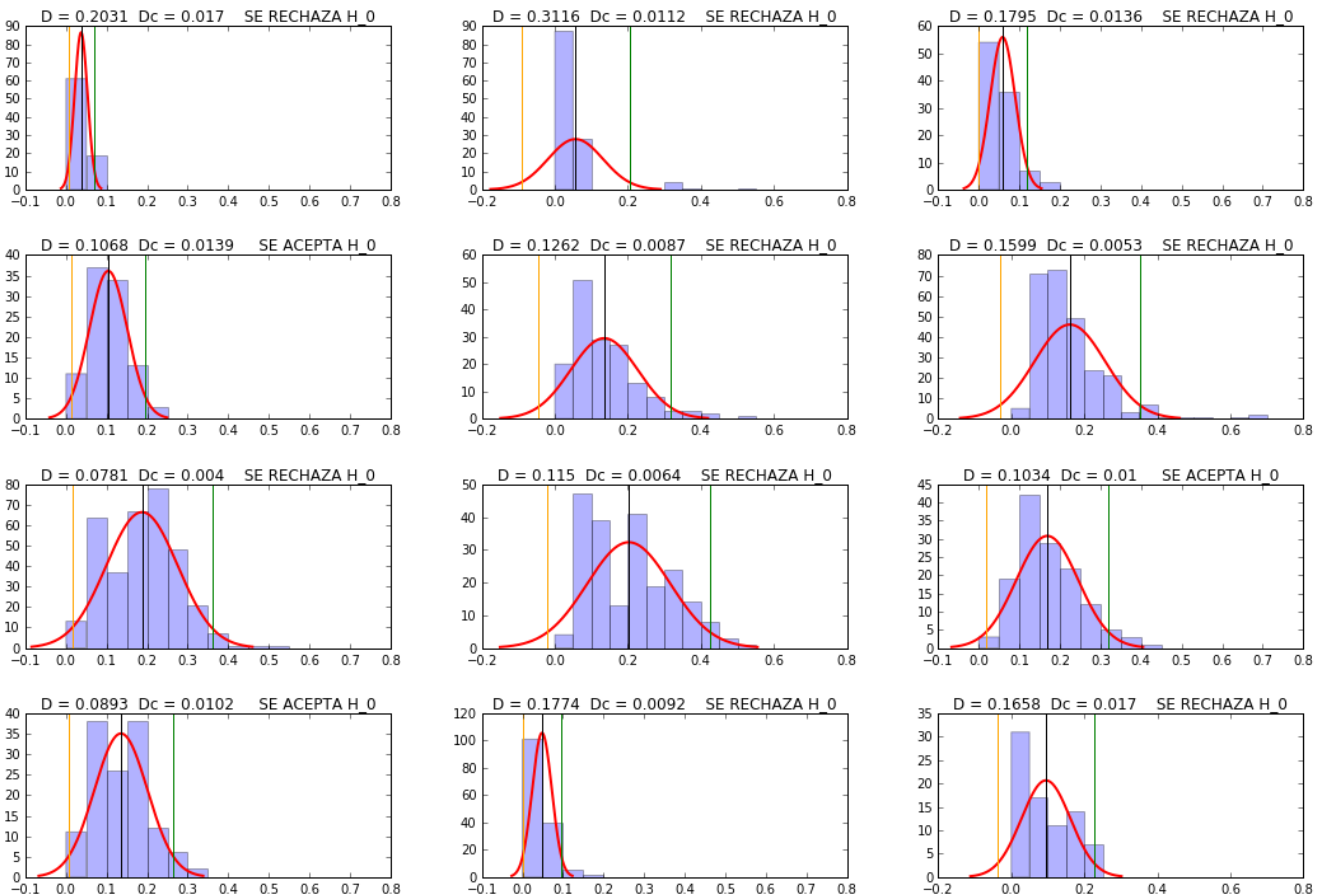


In [9]:

```
fig, axs = plt.subplots(4,3, figsize=(18, 12))
fig.subplots_adjust(hspace = .4, wspace = .25)

axs = axs.ravel()
for i in range(12):
    pylab.subplot(4,3,i+1)
    PlotHist2(df_out_Murcia[df_out_Murcia['month']==i+1]['AOT_500'], bin=np.arange(0,0.75,0.05), cdf="norm")
```

```
mu y sigma: (0.037083956250000001, 0.015996074733195045)
mu y sigma: (0.056486512396694208, 0.075342203359141116)
mu y sigma: (0.058807980000000003, 0.0309174694448721)
mu y sigma: (0.10421070918367346, 0.046904125169139041)
mu y sigma: (0.13508058598726111, 0.091927005104025183)
mu y sigma: (0.16055120270270271, 0.097247064422738158)
mu y sigma: (0.18726878846153847, 0.088059704667316732)
mu y sigma: (0.20239741079812204, 0.11410469340921443)
mu y sigma: (0.16883170588235294, 0.076371755267323968)
mu y sigma: (0.13531514661654134, 0.065546256859277161)
mu y sigma: (0.048316212837837831, 0.024207079076358884)
mu y sigma: (0.095018875000000003, 0.066918059357951906)
```



In [10]:

```
fig, axs = plt.subplots(4,3, figsize=(17, 11))
fig.subplots_adjust(hspace = .3, wspace = .15)
```

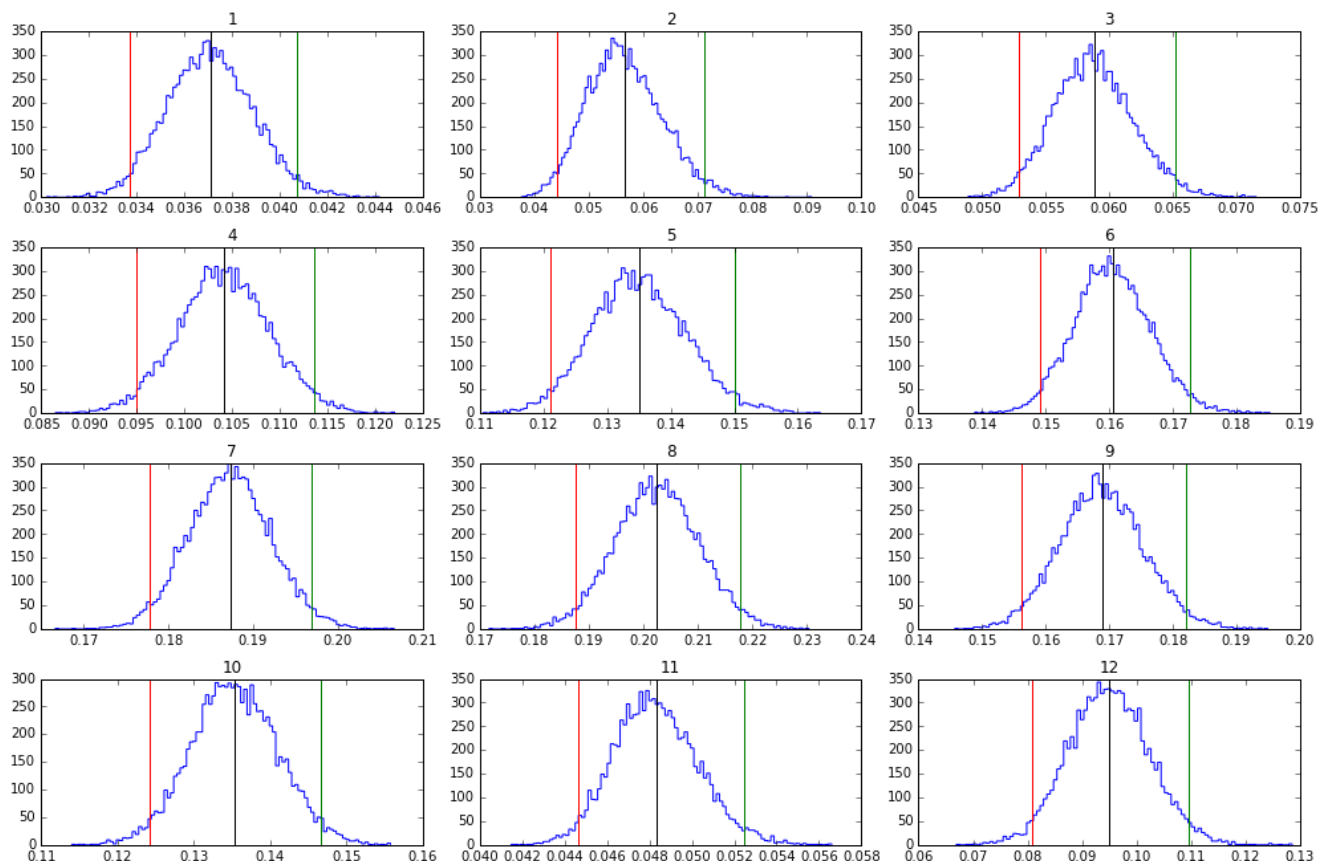
```

axs = axs.ravel()

for i in range(12):
    x = df_out_Murcia[df_out_Murcia['month']==i+1]['AOT_500'].values
    stat_mean, ci_mean = bootstrap(x, np.mean)
    obs_mean = np.mean(x)

    axs[i].hist(stat_mean, 100, histtype='step')
    axs[i].set_title(str(1+i))
    #pylab.hist(stat, 100, histtype='step')
    color = ['red', 'green']
    for j in xrange(2):
        axs[i].axvline(ci_mean[j], c=color[j])
        #pylab.title("Histogram of data for " + statistic.upper() + "'s Bootstrap")
    axs[i].axvline(obs_mean, c='black')

```



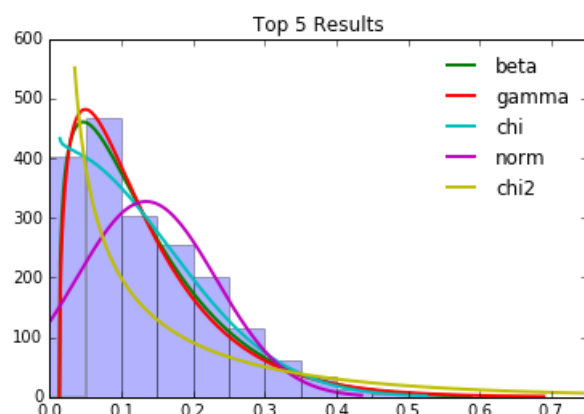
In [11]:

```
distribution_check(df_out_Murcia['AOT_500'].values, verbose=False)
```

reading data in file None ...

Top 5

1	beta	p: 0.0107518556587	D: 0.0373574103218
2	gamma	p: 0.00562898864882	D: 0.0396076248531
3	chi	p: 0.00023131791096	D: 0.0492243540352
4	norm	p: 0.0	D: 0.112474053354
5	chi2	p: 0.0	D: 0.256897289801



In [12]:

```
fig, axs = plt.subplots(4,3, figsize=(18, 12))
fig.subplots_adjust(hspace = .4, wspace = .25)

axs = axs.ravel()
for i in range(12):
    pylab.subplot(4,3,i+1)
    distribution_check(df_out_Murcia[df_out_Murcia['month']==i+1]['AOT_500'], verbose=False, name='Month
h '+str(i+1))
```

reading data in file Month 1 ...

Top 5

1	gamma	p:	0.272231692541	D:	0.109579889671
2	chi	p:	0.0730036750878	D:	0.141643367657
3	t	p:	0.015154828929	D:	0.172271925056
4	norm	p:	0.00227277931935	D:	0.203098464067
5	beta	p:	0.000106193186467	D:	0.244610388892

reading data in file Month 2 ...

Top 5

1	t	p:	0.0254558190956	D:	0.132772070227
2	chi2	p:	0.0238265609644	D:	0.133780680485
3	beta	p:	0.00363374336439	D:	0.159794782042
4	chi	p:	2.76214160433e-07	D:	0.252481754221
5	norm	p:	6.13351591738e-11	D:	0.311647753361

reading data in file Month 3 ...

Top 5

1	gamma	p:	0.515223205536	D:	0.0807294837433
2	chi	p:	0.269430095439	D:	0.0984622427301
3	chi2	p:	0.0555880231572	D:	0.132073512968
4	beta	p:	0.0366111225361	D:	0.139619121751
5	t	p:	0.0335879148627	D:	0.141125881834

reading data in file Month 4 ...

Top 5

1	beta	p:	0.395800762001	D:	0.0892484688439
2	t	p:	0.199296959502	D:	0.106765153901
3	norm	p:	0.199257361619	D:	0.106769813603
4	gamma	p:	0.0668775383557	D:	0.129877482327
5	chi2	p:	0.0668483593419	D:	0.129885897384

reading data in file Month 5 ...

Top 5

1	chi2	p:	0.898445143275	D:	0.0456921691207
2	gamma	p:	0.898400858446	D:	0.0456952211998
3	beta	p:	0.770732389804	D:	0.0529571518271
4	chi	p:	0.410337550417	D:	0.0699803746432
5	t	p:	0.0251938210466	D:	0.11687425961

reading data in file Month 6 ...

Top 5

1	chi2	p:	0.0451258658014	D:	0.0848798168867
2	gamma	p:	0.0451160315725	D:	0.0848822708974
3	beta	p:	0.0444612021815	D:	0.0850467293736
4	chi	p:	0.00116454974629	D:	0.119133124768
5	t	p:	0.000331041060676	D:	0.128816936059

reading data in file Month 7 ...

Top 5

1	t	p:	0.0307490453122	D:	0.0780687298974
2	norm	p:	0.0307354706399	D:	0.0780728783944
3	chi2	p:	0.0234268888808	D:	0.0805835220352
4	gamma	p:	0.0234268326551	D:	0.080583543881
5	beta	p:	0.00281374505282	D:	0.0979920443948

reading data in file Month 8 ...

Top 5

1	beta	p: 0.0235070413002	D: 0.101288209299
2	norm	p: 0.0065651522336	D: 0.114975568261
3	t	p: 0.00656069078185	D: 0.114982426942
4	chi2	p: 0.00086436452024	D: 0.133869017967
5	gamma	p: 0.000863857664648	D: 0.133874094832

reading data in file Month 9 ...

Top 5

1	gamma	p: 0.947394058757	D: 0.0448372175392
2	chi2	p: 0.947388776444	D: 0.0448377785827
3	beta	p: 0.940141892166	D: 0.0455782971959
4	chi	p: 0.762513422346	D: 0.0573401508204
5	t	p: 0.300303189837	D: 0.082279739563

reading data in file Month 10 ...

Top 5

1	beta	p: 0.634425970506	D: 0.0644818926875
2	chi	p: 0.328114732572	D: 0.0811965735992
3	t	p: 0.226561860692	D: 0.0892342050755
4	norm	p: 0.226248680732	D: 0.0892628795701
5	gamma	p: 0.160369107209	D: 0.0961331518101

reading data in file Month 11 ...

Top 5

1	chi2	p: 0.336832779539	D: 0.0764640607181
2	beta	p: 0.309861081601	D: 0.0782581469122
3	t	p: 0.0725405301562	D: 0.104683193818
4	chi	p: 0.0428965330161	D: 0.1127318554
5	norm	p: 0.000152073964935	D: 0.177377334252

reading data in file Month 12 ...

Top 5

1	beta	p: 0.162068987143	D: 0.12321245601
2	chi	p: 0.143689547055	D: 0.126166779292
3	chi2	p: 0.0280208642301	D: 0.161000047848
4	gamma	p: 0.0261010682017	D: 0.162342131829
5	norm	p: 0.0216504304214	D: 0.165824742248

