

This is a test file, to get started.

```
## Set up
import sys
sys.path.append('.') # shouldn't need this once installed from PyPI

from debacl import geom_tree as gtree
from debacl import utils as utl

import numpy as np
import matplotlib.pyplot as plt

## Output parameters
utl.setPlotParams(axes_labelsize=24, xtick_labelsize=18, ytick_labelsize=18, figsize=(6,6))

## Data parameters
n = 1500
p_k = 0.02
p_gamma = 0.05
mix = (0.5, 0.3, 0.2)

k = int(p_k * n)
gamma = int(p_gamma * n)

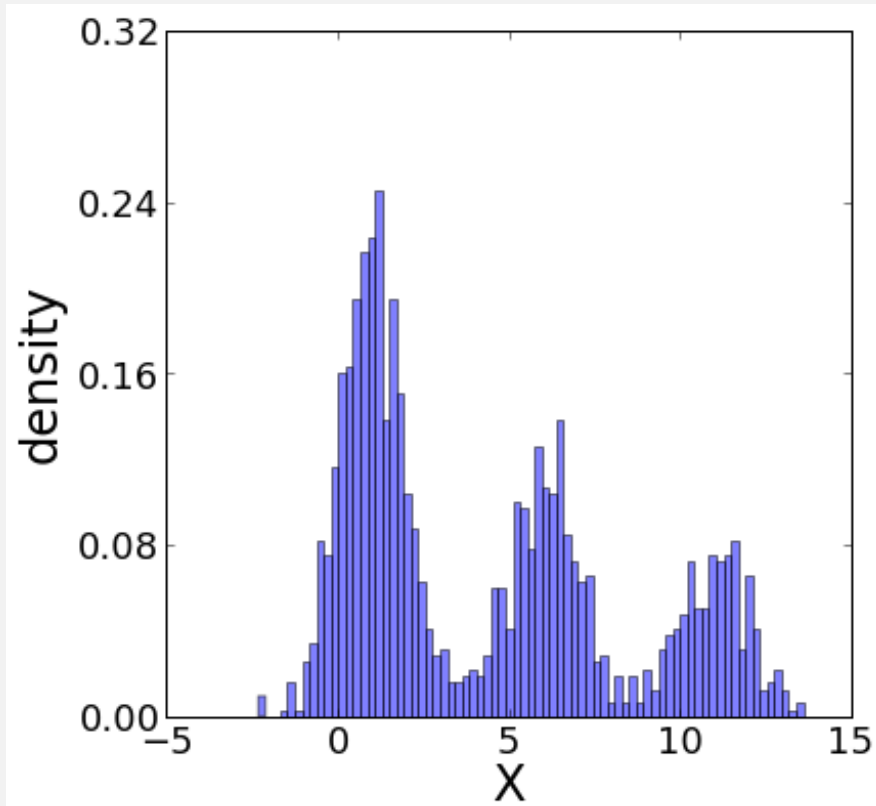
ctr = ((1,), (6,), (11,))
sdev = (np.eye(1),) * 3

## Generate data
membership = np.random.multinomial(n, pvals=mix)
p = len(ctr[0])
X = np.zeros((n, p), dtype=np.float)
g = np.zeros((n, ), dtype=np.int)
b = np.cumsum((0,) + tuple(membership))

for i, (size, mu, sigma) in enumerate(zip(membership, ctr, sdev)):
    ix = range(b[i], b[i+1])
    X[ix, :] = np.random.multivariate_normal(mu, sigma, size)
    g[ix] = i

X = np.sort(X, axis=0) # sort the points for prettier downstream plotting

## Plot a histogram of the data to show the simulation worked
fig, ax = plt.subplots()
ax.hist(X, bins=n/20, normed=1, alpha=0.5)
ax.set_xlabel('X')
ax.set_ylabel('density')
fig.show()
```



```
## Estimate the level set tree - the easy way
tree = gtree.geomTree(X, k, gamma, n_grid=None, verbose=True)
print tree

## Retrieve cluster assignments from the tree
uc, leaves = tree.getClusterLabels(method='all-mode')
print "cluster counts:", np.bincount(uc[:, 1])
print "leaf indices:", leaves

## Plot the level set tree with colored leaves
## note the plot function returns a tuple with 5 objects. The first member of
## tuple is the figure, which for most users is the only interesting part.
fig = tree.plot(form='lambda', width='uniform', color_nodes=leaves)[0]
fig.show()
```

```
iteration 0
iteration 100
iteration 200
iteration 300
iteration 400
iteration 500
iteration 600
```

```

iteration 700
iteration 800
iteration 900
iteration 1000
iteration 1100
iteration 1200
iteration 1300
iteration 1400

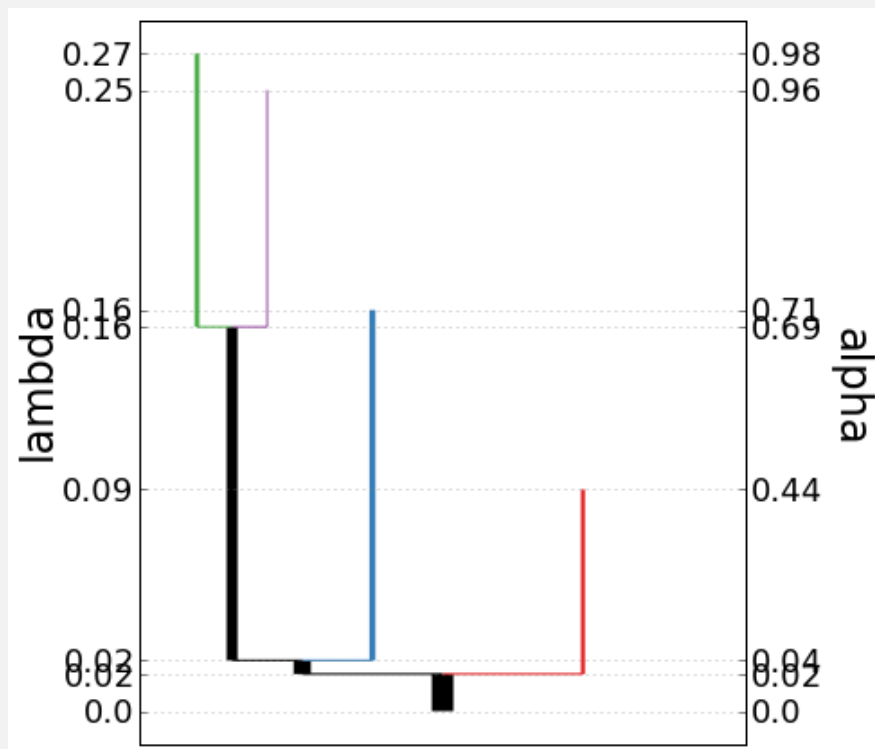
```

	alpha1	alpha2	children	lambda1	lambda2	parent	size
key							
0	0.000000	0.019333	[1, 2]	0.000000	0.015125	None	1500
1	0.019333	0.044667	[3, 4]	0.015125	0.020686	0	1198
2	0.019333	0.442000	[]	0.015125	0.090257	0	273
3	0.044667	0.688667	[37, 38]	0.020686	0.156450	1	763
4	0.044667	0.706000	[]	0.020686	0.163335	1	417
37	0.688667	0.976667	[]	0.156450	0.267754	3	341
38	0.688667	0.958667	[]	0.156450	0.252876	3	94

```

cluster counts: [273 417 341 94]
leaf indices: [2, 4, 37, 38]

```



```

## Assign background points
fc = utl.assignBackgroundPoints(X.reshape((n, -1)), uc, method='knn', k=9)
print "final cluster counts:", np.bincount(fc[:, 1])

```

```

final cluster counts: [289 455 536 220]

```

Switch to a 2D example with finer control of the individual components of tree estimation.

```
## Add the stats package for kernel density estimation
import scipy.stats as spstat

## Re-set data parameters
n = 1500
p_k = 0.005
p_gamma = 0.01
mix = (0.5, 0.3, 0.2)

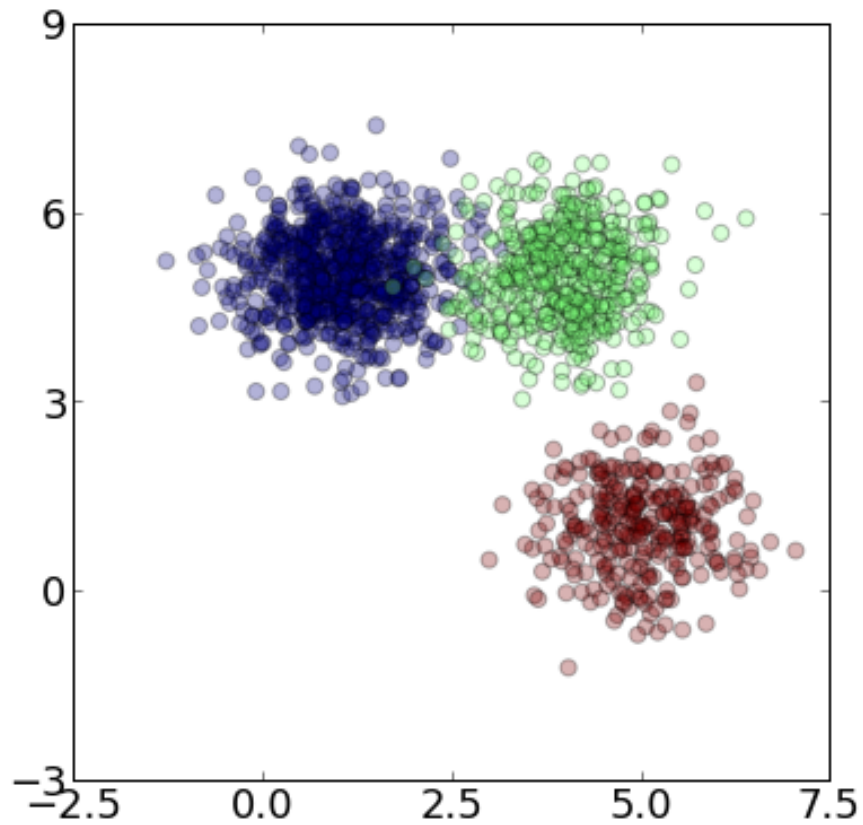
k = int(p_k * n)
gamma = int(p_gamma * n)

ctr = ((1, 5), (4, 5), (5, 1))
sdev = (0.5*np.eye(2),) * 3

## Generate data
membership = np.random.multinomial(n, pvals=mix)
p = len(ctr[0])
X = np.zeros((n, p), dtype=np.float)
g = np.zeros((n, ), dtype=np.int)
b = np.cumsum((0,) + tuple(membership))

for i, (size, mu, sigma) in enumerate(zip(membership, ctr, sdev)):
    ix = range(b[i], b[i+1])
    X[ix, :] = np.random.multivariate_normal(mu, sigma, size)
    g[ix] = i

## Scatterplot, to show the simulation worked
fig, ax = plt.subplots()
ax.scatter(X[:,0], X[:,1], s=50, c=g, alpha=0.3)
fig.show()
```



```
## Construct the similarity graph and density estimate
W, k_radius = utl.knnGraph(X, k, self_edge=False)
kernel = spstat.gaussian_kde(X.T)
fhat = kernel(X.T)

## Construct the level set tree
bg_sets, levels = utl.constructDensityGrid(fhat, mode='mass', n_grid=300)
tree = gtree.constructTree(W, levels, bg_sets, mode='density', verbose=True)
print tree
```

```
iteration 0
iteration 100
iteration 200
```

	alpha1	alpha2	children	lambda1	lambda2	parent	size
key							
0	0.000000	0.006667	[1, 2]	0.000000	0.003153	None	1500
1	0.006667	0.387333	[5, 6]	0.003153	0.038202	0	1201
2	0.006667	0.120000	[3, 4]	0.003153	0.017841	0	289
3	0.120000	0.404667	[7, 8]	0.017841	0.039015	2	201
4	0.120000	0.123333	[]	0.017841	0.018351	2	1
5	0.387333	0.882667	[11, 12]	0.038202	0.088096	1	589
6	0.387333	0.732000	[9, 10]	0.038202	0.065463	1	314

7	0.404667	0.418000	[]	0.039015	0.040040	3	5
8	0.404667	0.408000	[]	0.039015	0.039220	3	3
9	0.732000	0.738667	[]	0.065463	0.066172	6	2
10	0.732000	0.748667	[]	0.065463	0.067523	6	10
11	0.882667	0.922667	[13, 14]	0.088096	0.094461	5	175
12	0.882667	0.896000	[]	0.088096	0.090142	5	1
13	0.922667	0.976000	[15, 16]	0.094461	0.101287	11	115
14	0.922667	0.926000	[]	0.094461	0.095457	11	1
15	0.976000	1.000000	[]	0.101287	0.105238	13	31
16	0.976000	0.979333	[17, 18]	0.101287	0.101752	13	5
17	0.979333	0.989333	[]	0.101752	0.103493	16	1
18	0.979333	0.989333	[]	0.101752	0.103493	16	2

```
## Save and/or load a tree (obviously redundant in this tutorial)
tree.save('test_tree')
tree = gtree.loadTree('test_tree')

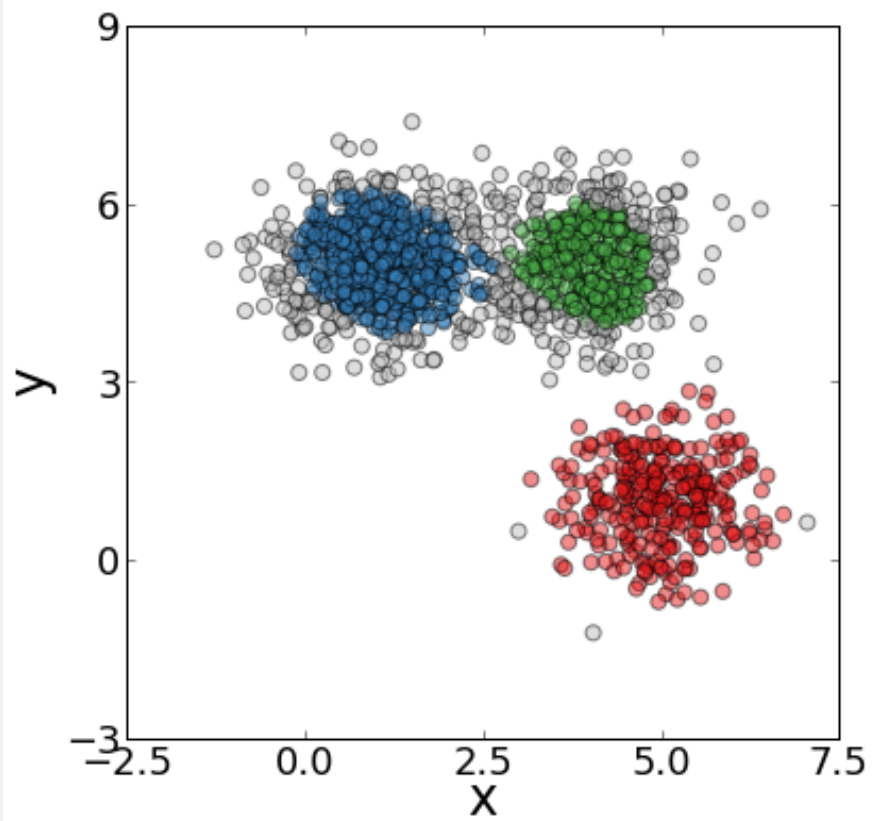
## Prune the tree
tree.mergeBySize(gamma)
print tree
```

	alpha1	alpha2	children	lambda1	lambda2	parent	size
key							
0	0.000000	0.006667	[1, 2]	0.000000	0.003153	None	1500
1	0.006667	0.387333	[5, 6]	0.003153	0.038202	0	1201
2	0.006667	0.418000	[]	0.003153	0.040040	0	289
5	0.387333	1.000000	[]	0.038202	0.105238	1	589
6	0.387333	0.748667	[]	0.038202	0.067523	1	314

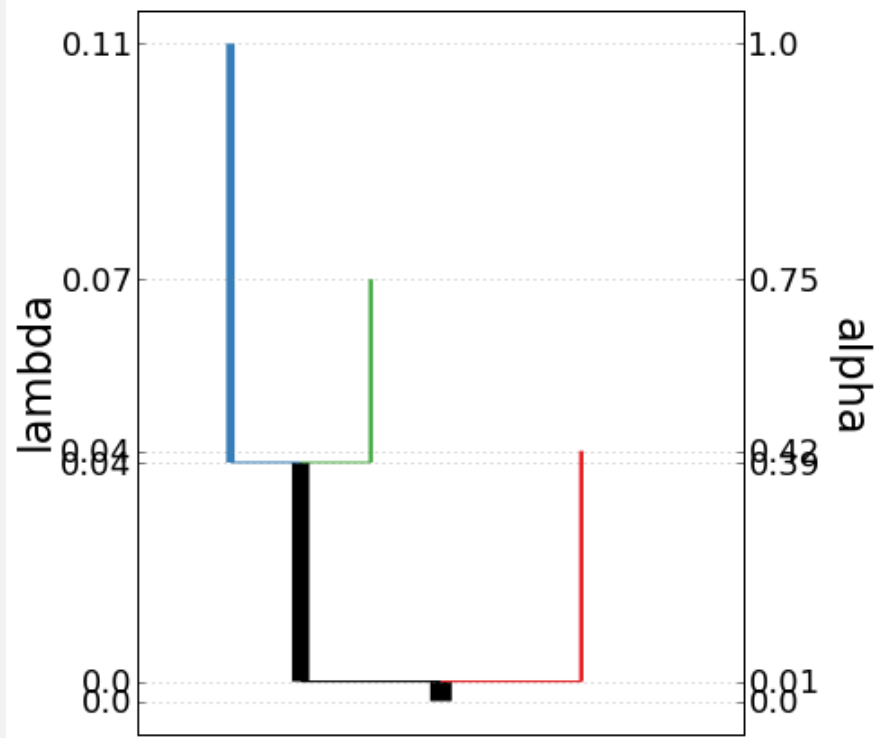
```
## Interactive tools
#tool = gtree.ComponentGUI(tree, X, form='alpha', width='mass', output=['scatter'])
#tool = gtree.ClusterGUI(tree, X, form='alpha', width='mass', output=['scatter'])
#tool.show()
```

```
## Get foreground clusters and plot them
uc, nodes = tree.getClusterLabels(method='first-k', k=3)

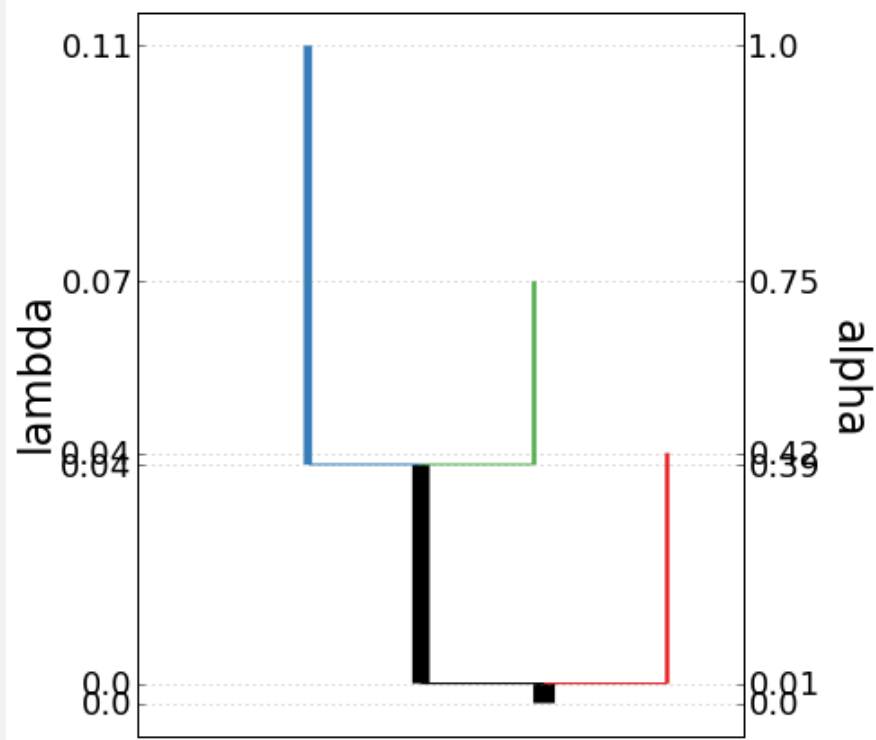
fig, ax = utl.plotForeground(X, uc, s=50, alpha=0.5)
fig.show()
```



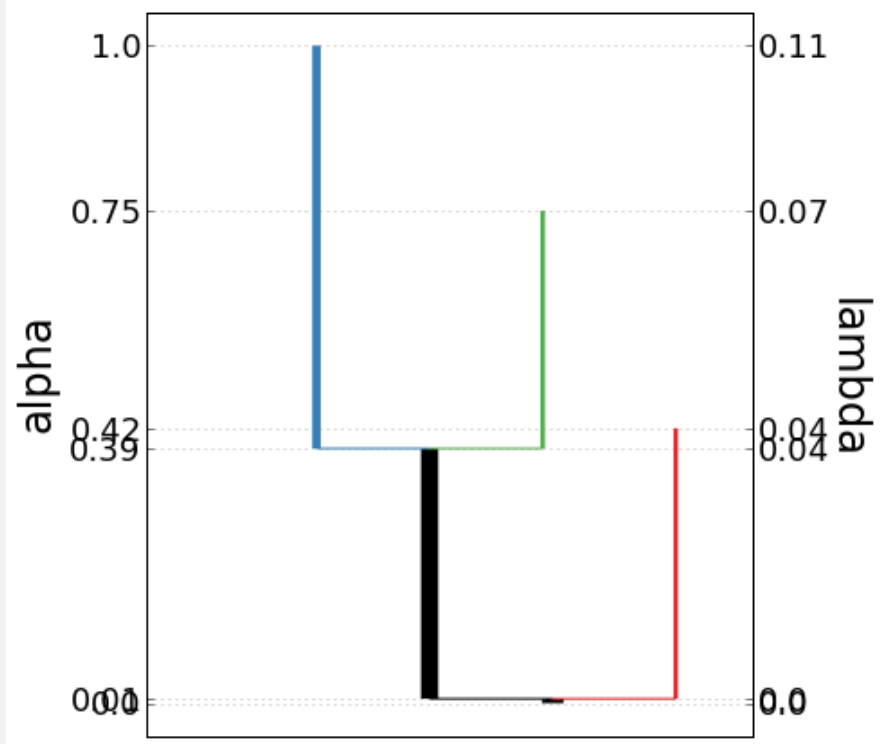
```
## Plot the basic lambda scale tree  
fig = tree.plot(form='lambda', width='uniform', color_nodes=nodes)[0]  
fig.show()
```



```
## Plot the basic level set tree with mass-based spacing
fig = tree.plot(form='lambda', width='mass', color_nodes=nodes)[0]
fig.show()
```

```
## Plot the level set tree with alpha scale
fig = tree.plot(form='alpha', width='mass', color_nodes=nodes)[0]
fig.show()
```



```
## Plot the level set tree with kappa scale
fig = tree.plot(form='kappa', width='mass', color_nodes=nodes)[0]
fig.show()
```

