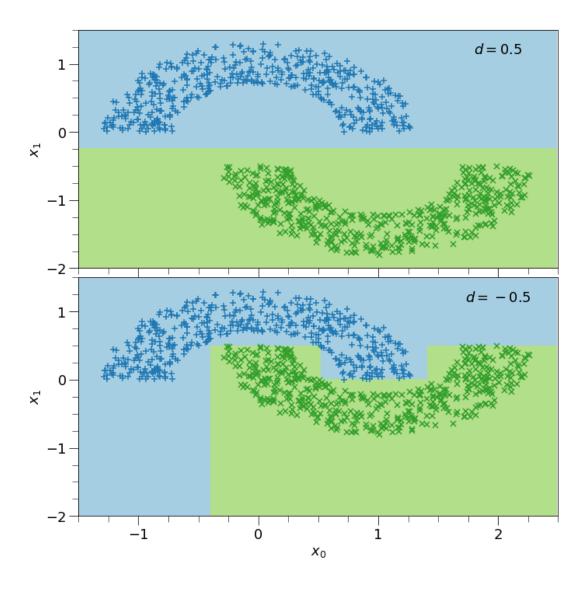
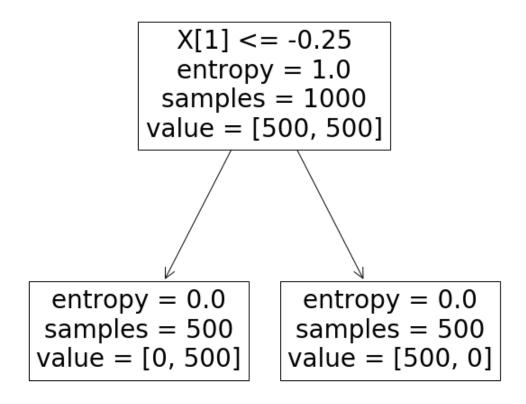
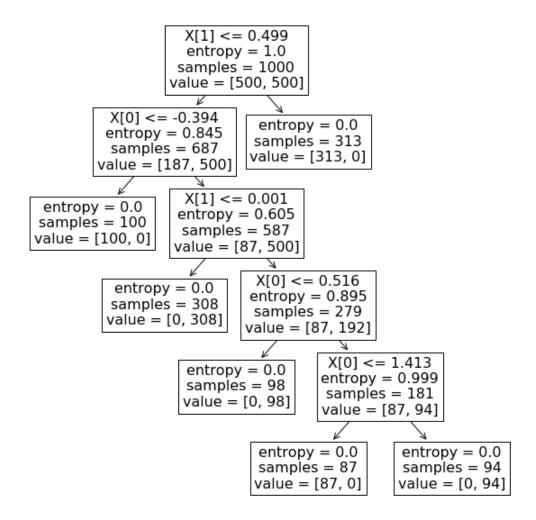
```
In [3]: #!/usr/bin/env python3
 # -*- coding: utf-8 -*-
Created on Tue Dec 3 21:49:59 2019
 @author: jorgeagr
 import numpy as np
 from sklearn.tree import DecisionTreeClassifier, plot_tree
from requiredFunctions.doubleMoon import doubleMoon
 import matplotlib as mpl
 import matplotlib.ticker as mtick
import matplotlib.pyplot as plt
width = 10
height = 10
mpl.rcParams['figure.figsize'] = (width, height)
mpl.rcParams['font.size'] = 18
mpl.rcParams['figure.titlesize'] = 'small'
mpl.rcParams['legend.fontsize'] = 'small'
mpl.rcParams['xtick.major.size'] = 12
mpl.rcParams['xtick.minor.size'] = 8
mpl.rcParams['xtick.labelsize'] = 18
mpl.rcParams['ytick.major.size'] = 12
mpl.rcParams['ytick.minor.size'] = 8
mpl.rcParams['ytick.labelsize'] = 18
cmap = plt.get_cmap('Paired')
cmap scatter = mpl.colors.ListedColormap(cmap((1, 3, 5)))
 cmap contour = mpl.colors.ListedColormap(cmap((0, 2)))
N = 1000
r = 1
w = 0.6
d_{vals} = [0.5, -0.5]
\overline{\text{titles}} = ['b', 'c']
```

```
In [4]: | figds, axds = plt.subplots(nrows=2, sharex=True)
for i, d in enumerate(d_vals):
     # Part B
     data = doubleMoon(N, w, r, d, seed=0)
     x_train = data[:,:2]
     y train = data[:,-1]
     tree = DecisionTreeClassifier(criterion='entropy', random state=1)
     tree.fit(x train, y train)
     # Predict for the data points and assign indeces of what's right a
 nd wrong
     y_pred = tree.predict(x_train)
     y_right = np.where(y_pred == y_train)[0]
     y wrong = np.where(y pred != y train)[0]
     blue_ind = y_right[np.where(y_right<N//2)]</pre>
     green_ind = y_right[np.where(y_right>=N//2)]
     # Make mesh for decision regions and predict for the points within
     x0 \text{ min}, x0 \text{ max} = -1.5, 2.5
     x1 \text{ min}, x1 \text{ max} = -2, 1.5
     xx0, xx1 = np.meshgrid(np.arange(x0 min, x0 max, 0.01),
                             np.arange(x1 min, x1 max, 0.01))
     cc = tree.predict(np.c [xx0.ravel(), xx1.ravel()]).reshape(xx0.sha
pe)
     # Plotting of decision regions and data points
     axds[i].contourf(xx0, xx1, cc, cmap=cmap_contour)
     axds[i].scatter(x train[:,0][blue ind], x train[:,1][blue ind], 5
 0,
                   c=[cmap_scatter(0)], marker='+')
     axds[i].scatter(x train[:,0][green ind], x train[:,1][green ind],
 50,
                   c=[cmap scatter(1)], marker='x')
     axds[i].scatter(x train[:,0][y wrong], x train[:,1][y wrong], 50,
                   c=[cmap scatter(2)], marker='*')
     axds[i].set_xlim(x0_min, x0_max)
     axds[i].set ylim(x1 min, x1 max)
     axds[i].xaxis.set_major_locator(mtick.MultipleLocator(1))
     axds[i].xaxis.set_minor_locator(mtick.MultipleLocator(0.25))
     axds[i].yaxis.set_major_locator(mtick.MultipleLocator(1))
     axds[i].yaxis.set_minor_locator(mtick.MultipleLocator(0.25))
     axds[i].text(2, 1.15, r'$d = {}$'.format(d vals[i]), horizontalali
 gnment='center')
     axds[i].set_ylabel(r'$x 1$')
     fig, ax = plt.subplots()
     plot_tree(tree, ax=ax)
     #fig.savefig('../prob1' + titles[i] + '.eps')
 axds[1].set xlabel(r'$x 0$')
 figds.tight layout(h_pad=0)
 #figds.savefig('../problds.eps',dpi=500)
```







In []: