Pruning Multiple neurons at one play

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Compute the performance of MAB methods of pruning Multiple neurons at one time
MAP for choosing multi arms at one time

In [1]:
```
import numpy as np
import time
import sys
from numpy import *
import matplotlib.pyplot as plt
from sklearn import metrics
%matplotlib inline
#plt.rcParams['figure.figsize'] = (15, 6)
```

1 Load Bokeh

In [2]:
```
from bokeh.layouts import row, gridplot
from bokeh.plotting import figure, output_notebook, show
from bokeh.models import Legend
TOOLS = 'box_zoom,box_select,crosshair,resize,reset,lasso_select,pan,save,poly_select,ta
output_notebook()
```

2 Load the data

In [3]:
```
X_train = np.load('./winequilty/X_train.npy')
y_train = np.load('./winequilty/y_train.npy')
X_test = np.load('./winequilty/X_test.npy')
y_test = np.load('./winequilty/y_test.npy')
X_deploy = np.load('./winequilty/X_deploy.npy')
y_deploy = np.load('./winequilty/y_deploy.npy')
print('Number of training examples',len(X_train))
print('Number of validation examples',len(X_test))
print('Number of testing examples',len(X_deploy))
```

```
Number of training examples 4157
Number of validation examples 1040
Number of testing examples 1300
```

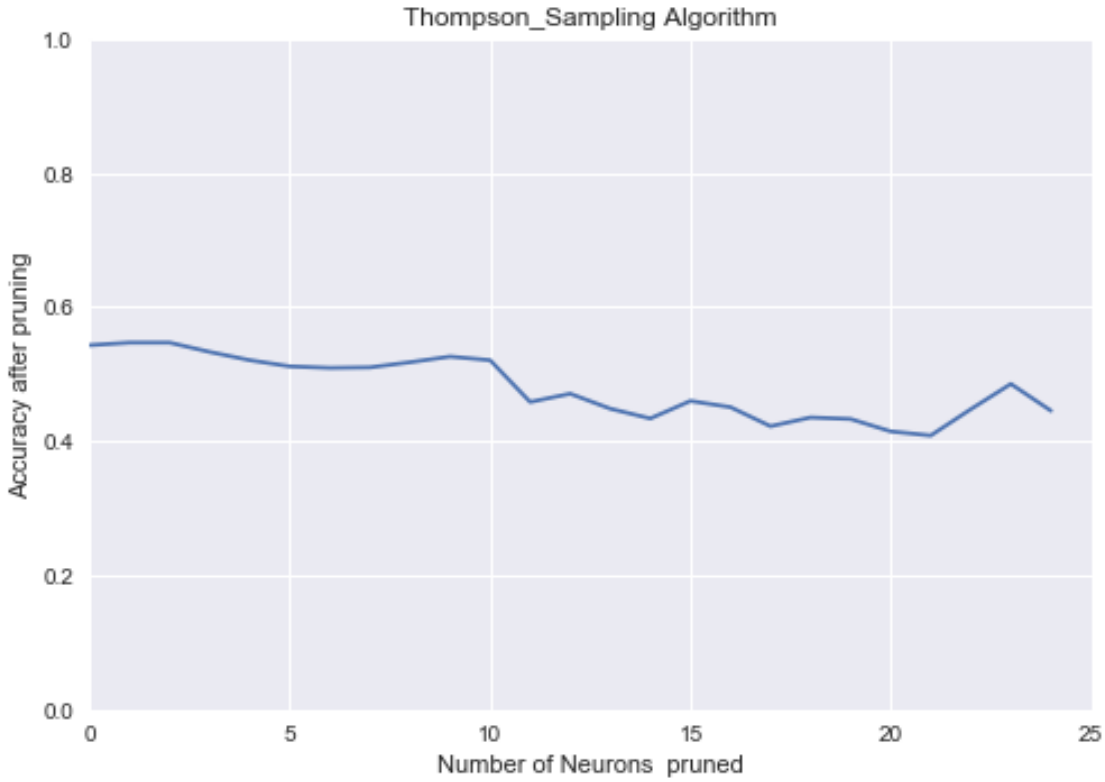
In [4]:
```
exec(open("core.py").read())  # pyhton 3x
```

2.1 Run Thompson Sampling pruning Algorithm

```
In [5]: algo = Thompson_Sampling([], [])
        Alg_name = 'Thompson_Sampling Algorithm'
        path = './Thompson_Sampling/'
        sys.path.append("./Thompson_Sampling")
        exec(open("mnist_cnnFORTESTING.py").read())
```

```
/Library/Frameworks/Python.framework/Versions/3.6/lib/python3.6/site-packages/sklearn/cross_vali
  "This module will be removed in 0.20.", DeprecationWarning)
Using Theano backend.


Test fraction correct (NN-Score) = 1.06
Test fraction correct (NN-Accuracy) = 0.55
The time for running this method is 0.17600369453430176 seconds
Finsh playing start pruining:
Test after pruning= 0.54
Test after pruning= 0.55
Test after pruning= 0.55
Test after pruning= 0.53
Test after pruning= 0.52
Test after pruning= 0.51
Test after pruning= 0.51
Test after pruning= 0.51
Test after pruning= 0.52
Test after pruning= 0.53
Test after pruning= 0.52
Test after pruning= 0.46
Test after pruning= 0.47
Test after pruning= 0.45
Test after pruning= 0.43
Test after pruning= 0.46
Test after pruning= 0.45
Test after pruning= 0.42
Test after pruning= 0.44
Test after pruning= 0.43
Test after pruning= 0.41
Test after pruning= 0.41
Test after pruning= 0.45
Test after pruning= 0.49
Test after pruning= 0.45
```

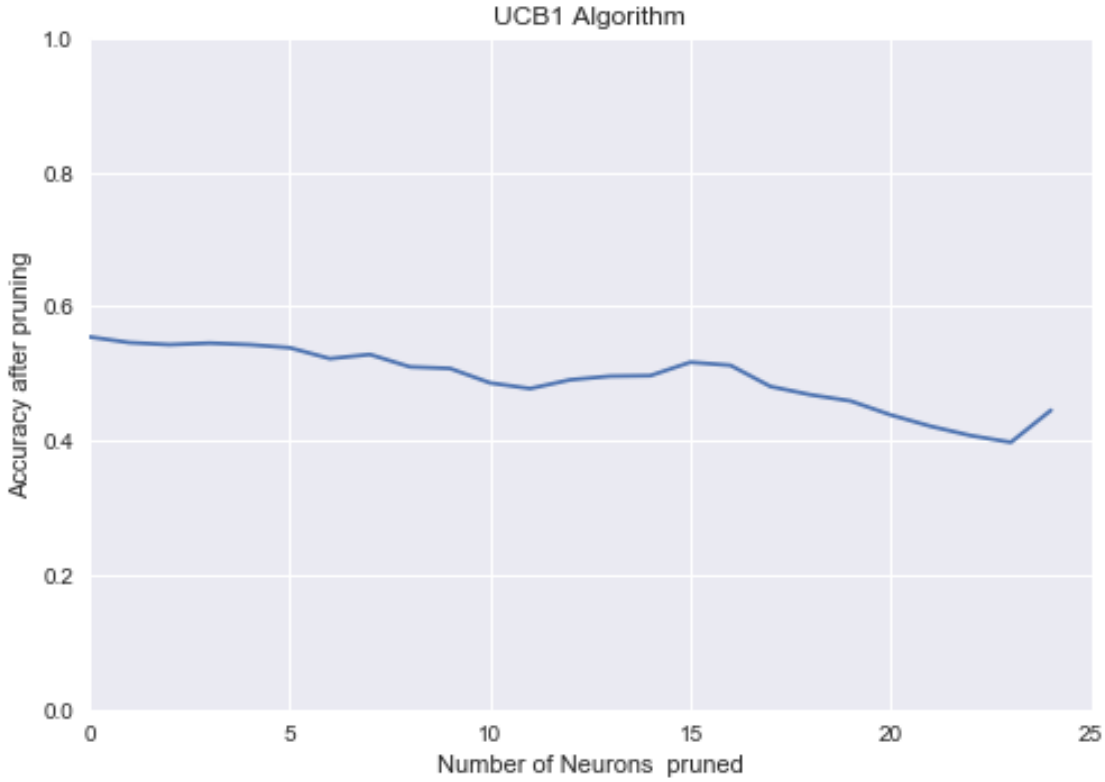


2.2 Run UCB1 pruning Algorithm

```
In [6]: algo = UCB1([], [])
        Alg_name = 'UCB1 Algorithm'
        path = './UCB1/'
        sys.path.append("./UCB1")
        exec(open("mnist_cnnFORTESTING.py").read())
```

```
Test fraction correct (NN-Score) = 1.06
Test fraction correct (NN-Accuracy) = 0.55
The time for running this method is 0.1891798973083496 seconds
Finsh playing start pruining:
Test after pruning= 0.55
Test after pruning= 0.55
Test after pruning= 0.54
Test after pruning= 0.55
Test after pruning= 0.54
Test after pruning= 0.54
Test after pruning= 0.52
Test after pruning= 0.53
Test after pruning= 0.51
Test after pruning= 0.51
```

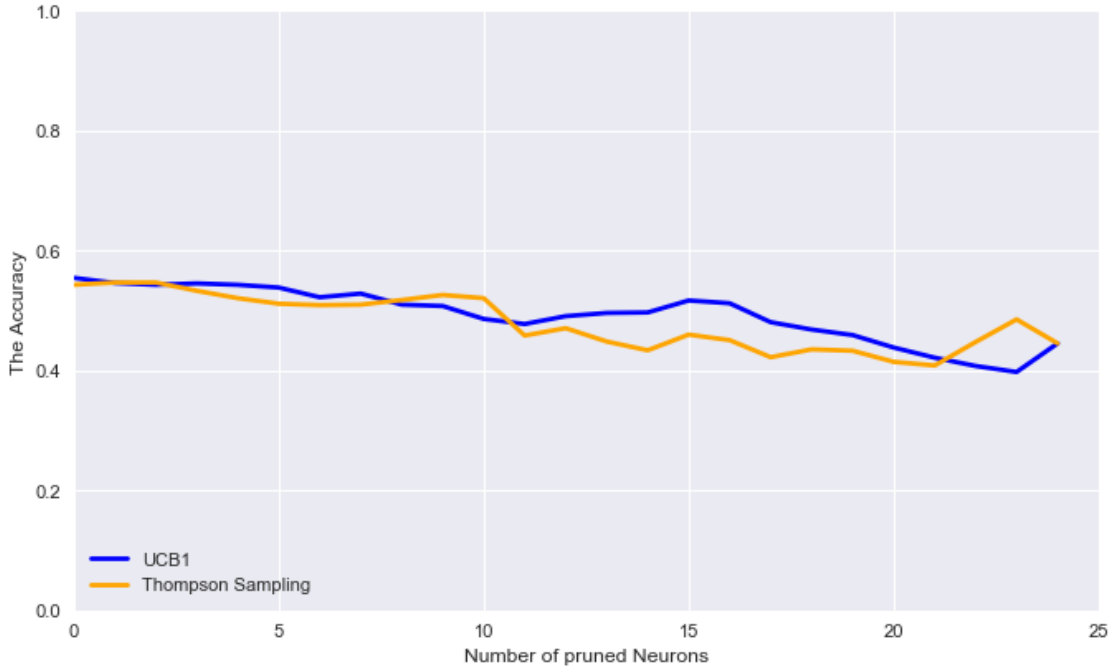
```
Test after pruning= 0.49
Test after pruning= 0.48
Test after pruning= 0.49
Test after pruning= 0.50
Test after pruning= 0.50
Test after pruning= 0.52
Test after pruning= 0.51
Test after pruning= 0.48
Test after pruning= 0.47
Test after pruning= 0.46
Test after pruning= 0.44
Test after pruning= 0.42
Test after pruning= 0.41
Test after pruning= 0.40
Test after pruning= 0.45
```



3 Compare the accuracy

```
In [7]: ucb1 = np.load('./UCB1/AccuracyAftrerPrune.npy')
        ThompsonSampling = np.load('./Thompson_Sampling/AccuracyAftrerPrune.npy')
        Accuracy = np.load('AccuracyBeforePruning.npy')
```

```
In [8]: fig = plt.figure(figsize=(10, 6), dpi=80)
        ax = fig.add_subplot(111)
        N = len(ucb1)
        ind = np.arange(N)                # the x locations for the groups
        plt.plot(ind , ucb1 , color="blue", linewidth=2.5, linestyle="-", label="UCB1")
        plt.plot(ind , ThompsonSampling, color="orange", linewidth=2.5, linestyle="-", label="Th
        plt.legend(loc = 3)
        plt.axis([0, 25, 0, 1])
        plt.xlabel('Number of pruned Neurons')
        plt.ylabel('The Accuracy')
        plt.grid(True)
        plt.show()
```

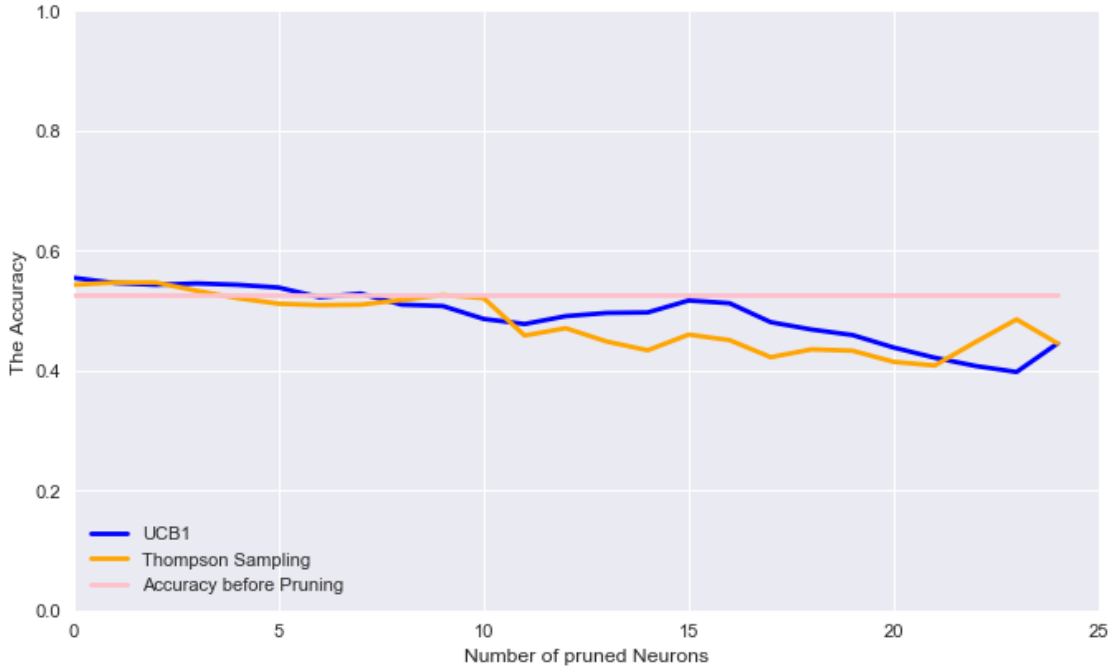



```
In [9]: p1 = figure(title="The Performance over the number of neurons' pruned", tools=TOOLS)
        p1.line(ind, ucb1, legend="ucb1", line_color="blue", line_width=2)
        p1.line(ind, ThompsonSampling, legend="Thompson Sampling", line_color="red", line_width=
        p1.title.align = "center"
        show(p1)
```

4 Comparing All algorithms with the model before pruning

```
In [10]: fig = plt.figure(figsize=(10, 6), dpi=80)
         ax = fig.add_subplot(111)
         N = len(ucb1)
         Acc = [Accuracy for col in range(N)]
```

```
ind = np.arange(N)                # the x locations for the groups
plt.plot(ind , ucb1 , color="blue", linewidth=2.5, linestyle="-", label="UCB1")
plt.plot(ind , ThompsonSampling, color="orange", linewidth=2.5, linestyle="-", label="T
plt.plot(ind , Acc, color="pink", linewidth=2.5, linestyle="-", label="Accuracy before
plt.legend(loc = 3)
plt.axis([0, 25, 0, 1])
plt.xlabel('Number of pruned Neurons')
plt.ylabel('The Accuracy')
plt.grid(True)
plt.show()
```



In [11]:
```
p1 = figure(title="The Performance over the number of neurons' pruned", tools=TOOLS)
p1.line(ind, ucb1, legend="ucb1", line_color="green", line_width=2)
p1.line(ind, ThompsonSampling, legend="Thompson Sampling", line_color="red", line_width
p1.line(ind, Acc, legend="Accuracy", line_dash=(4, 4), line_color="blue", line_width=2)
p1.title.align = "center"
show(p1)
```