

# Compression Normal

March 28, 2017

Compute the performance of MAB methods

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

## 0.1 Load BOKEH libariry

```
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,tap'
output_notebook()
```

## 1 Compare the accuracy of the models

### 1.1 Load the pruned algorithm from normal prune

```
In [3]: ucb1 = np.load('/Users/saleameen/Desktop/banditsbook/python_iris/UCB1/AccuracyAfterPrune.npy')
Accuracy = np.load('/Users/saleameen/Desktop/banditsbook/python_iris/AccuracyBeforePrune.npy')
```

### 1.2 Load the pruned algorithm from Multiple prune

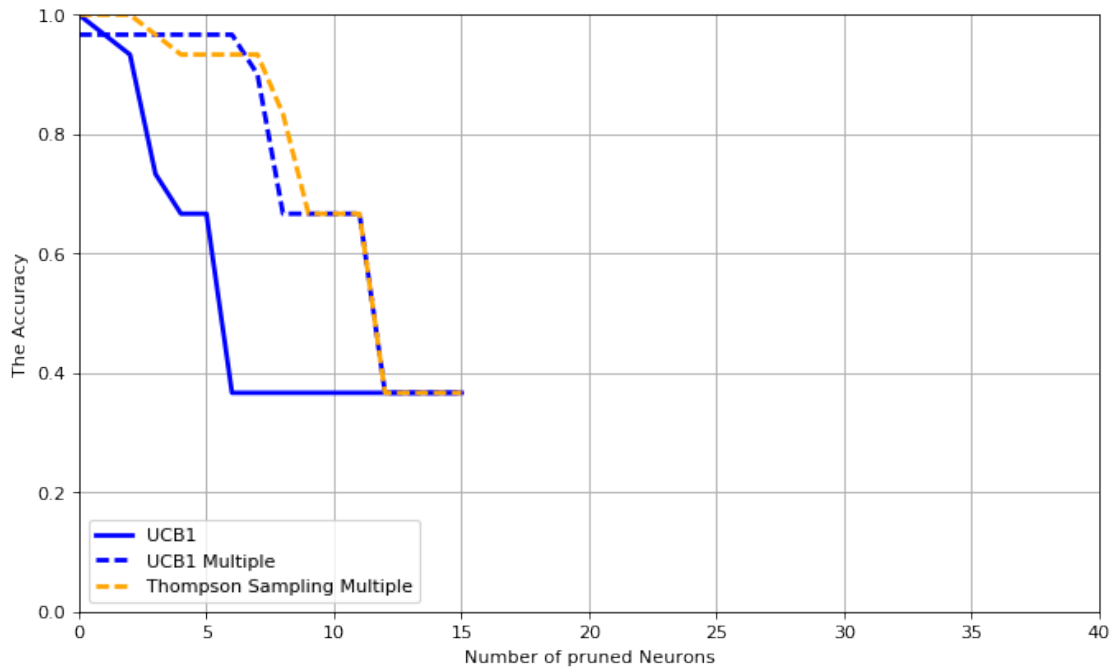
```
In [4]: ucb1_Multiple = np.load('/Users/saleameen/Desktop/banditsbook_Prune_many_one_play/python_iris/UCB1_Multiple.npy')
ThompsonSampling_Multiple = np.load('/Users/saleameen/Desktop/banditsbook_Prune_many_one_play/ThompsonSampling_Multiple.npy')
```

```
In [5]: fig = plt.figure(figsize=(10, 6), dpi=80)
ax = fig.add_subplot(111)
N = len(ucb1)
## necessary variables
ind = np.arange(N) # the x locations for the groups
### Normal algoritms
plt.plot(ind, ucb1, color="blue", linewidth=2.5, linestyle="--", label="UCB1")
### Absuluate once algoritms
```

```

plt.plot(ind , ucb1_Multiple , color="blue", linewidth=2.5, linestyle="--", label="UCB1
plt.plot(ind , ThompsonSampling_Multiple, color="orange", linewidth=2.5, linestyle="--",
#####
plt.legend(loc = 3)
plt.axis([0, 40, 0, 1])
plt.xlabel('Number of pruned Neurons')
plt.ylabel('The Accuracy')
plt.grid(True)
plt.show()

```



```

In [6]: p1 = figure(title="The Performance over the number of neurons' pruned", tools=TOOLS)
p1.line(ind, ucb1, legend="ucb1", line_color="orange", line_width=2)
##### =====
# Multiple pruned
p1.circle(ind, ucb1_Multiple, legend="ucb1 Multiple", line_color="orange", line_width=2)
p1.line(ind, ucb1_Multiple, legend="ucb1 Multiple", line_color="orange", line_width=2)
p1.circle(ind, ThompsonSampling_Multiple, legend="Thompson Sampling Multiple", line_color="orange", line_width=2)
p1.line(ind, ThompsonSampling_Multiple, legend="Thompson Sampling Multiple", line_color="orange", line_width=2)
p1.title.align = "center"
show(p1)

```

### 1.3 Comparing All algorithms with the model before pruning

```

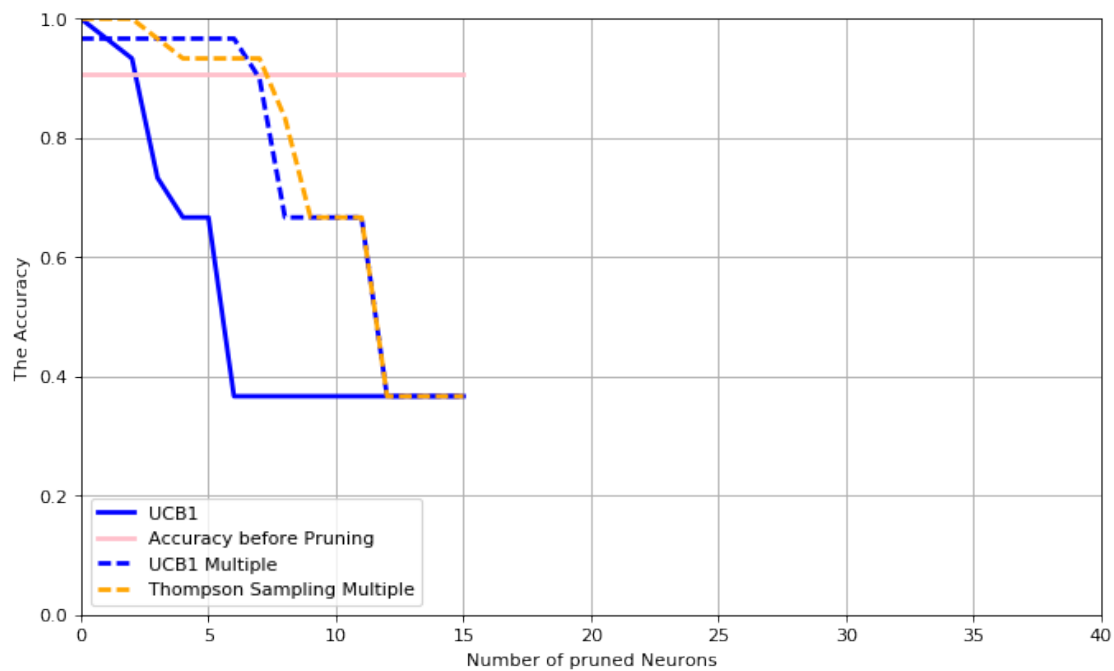
In [7]: fig = plt.figure(figsize=(10, 6), dpi=80)
ax = fig.add_subplot(111)

```

```

N = len(ucb1)
Acc = [Accuracy for col in range(N)]
## necessary variables
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 , Acc, color="pink", linewidth=2.5, linestyle="-", label="Accuracy before P
### Absuluate once algoritms
plt.plot(ind , ucb1_Multiple , color="blue", linewidth=2.5, linestyle="--", label="UCB1
plt.plot(ind , ThompsonSampling_Multiple, color="orange", linewidth=2.5, linestyle="--",
plt.legend(loc = 3)
plt.axis([0, 40, 0, 1])
plt.xlabel('Number of pruned Neurons')
plt.ylabel('The Accuracy')
plt.grid(True)
plt.show()

```



```

In [8]: p1 = figure(title="The Performance over the number of neurons' pruned", tools=TOOLS)
# Normal pruned
#p1.circle(ind, ucb1, legend="ucb1", color="orange")
p1.line(ind, ucb1, legend="ucb1", line_color="orange", line_width=2)

##### =====
# Multiple pruned
# Multiple pruned
p1.circle(ind, ucb1_Multiple, legend="ucb1 Multiple", line_color="red", line_width=2)

```

```

p1.line(ind, ucb1_Multiple, legend="ucb1 Multiple", line_color="red", line_width=2)

p1.circle(ind, ThompsonSampling_Multiple, legend="Thompson Sampling Multiple", line_color="red", line_width=2)
p1.line(ind, ThompsonSampling_Multiple, legend="Thompson Sampling Multiple", line_color="red", line_width=2)
p1.line(ind, Acc, legend="Accuracy", line_dash=(4, 4), line_color="orange", line_width=2)
p1.title.align = "center"
show(p1)

```

In [ ]: