DQN_v1

August 15, 2020

0.1 Jason's toy model in Q-learning

Thanks for this useful link:https://www.tensorflow.org/agents/tutorials/1_dqn_tutorial So far, Reinforcement learning algorithms are not well GPU optimized since it's hard to do parallization in updating stages + reward. Thus, the DQN part is CPU only :) Thanks to this useful link : https://adventuresinmachinelearning.com/reinforcement-learning-tensorflow/ Here we didn't consider the Volume

```
In [1]: import numpy as np # linear algebra
        import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
        import matplotlib
        import tensorflow as tf
        from matplotlib import colors as mcolors
        import matplotlib.pyplot as plt
        color_array = list(mcolors.CSS4_COLORS.keys())
        root_path = 'Data/Stocks/'
        #root_path = "/kaggle/input/price-volume-data-for-all-us-stocks-etfs/ETFs/"

        ## Here we got 2 sets of data: Google stock price daily and Google stock price in 5 da

        df = pd.read_csv(root_path + "googl.us.txt")
        df_hf = pd.read_csv("Data/High_frequency_data/GOOG_stock_1minute_sample/GOOG_sample.tx
```

```
In [2]: df.head()
```

Out[2]:

	Date	Open	High	Low	Close	Volume	OpenInt
0	2004-08-19	50.000	52.03	47.980	50.170	44703800	0
1	2004-08-20	50.505	54.54	50.250	54.155	22857200	0
2	2004-08-23	55.375	56.74	54.525	54.700	18274400	0
3	2004-08-24	55.620	55.80	51.785	52.435	15262600	0
4	2004-08-25	52.480	54.00	51.940	53.000	9197800	0

```
In [3]: df_hf.head()
```

Out[3]:

	DateTime	Open	High	Low	Close	Volume
0	2020-01-02 04:00:00	1342.00	1342.20	1342.00	1342.20	424
1	2020-01-02 04:02:00	1344.20	1344.20	1344.20	1344.20	177
2	2020-01-02 08:00:00	1337.02	1337.02	1337.02	1337.02	329
3	2020-01-02 08:09:00	1347.00	1347.00	1347.00	1347.00	155
4	2020-01-02 08:55:00	1348.00	1348.00	1348.00	1348.00	190

In [4]:
```
import matplotlib
from matplotlib.pylab import rc
font = {'family': 'normal','weight': 'bold',
        'size': 25}

matplotlib.rc('font', **font)
rc('axes', linewidth=3)


plt.subplot(1,1,1)
plt.plot(df["Open"],"k",label="Open",alpha=0.9)
plt.plot(df["Close"],"r",label="Close",alpha=0.9)


plt.xlabel("Day")
plt.ylabel(r"${\rm value}$")
plt.suptitle("Stock price vs Day Google")

fig = matplotlib.pyplot.gcf()

fig.set_size_inches(20,10)
plt.legend(fontsize=25,handlelength=5,ncol=3)
plt.show()
```

```
findfont: Font family ['normal'] not found. Falling back to DejaVu Sans.
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findfont: Font family ['normal'] not found. Falling back to DejaVu Sans.
```
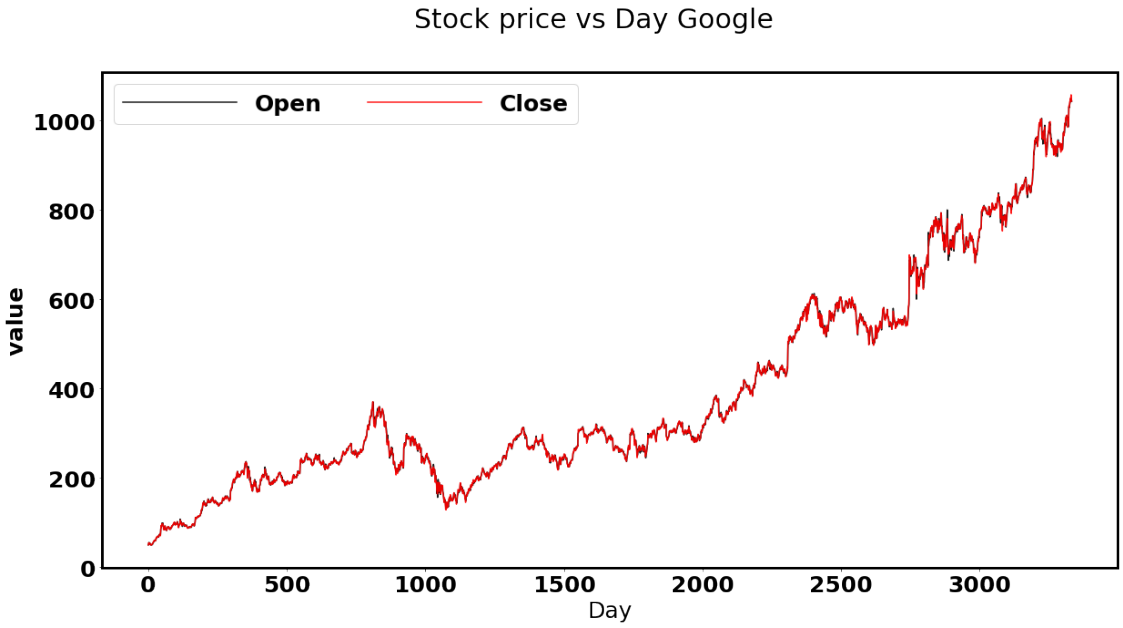
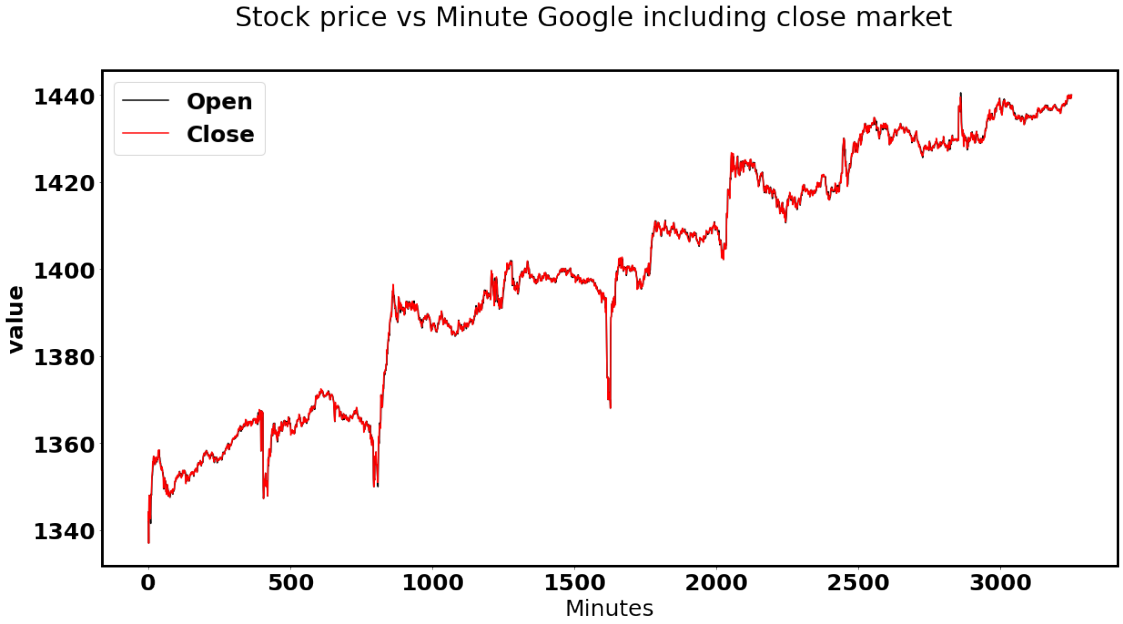



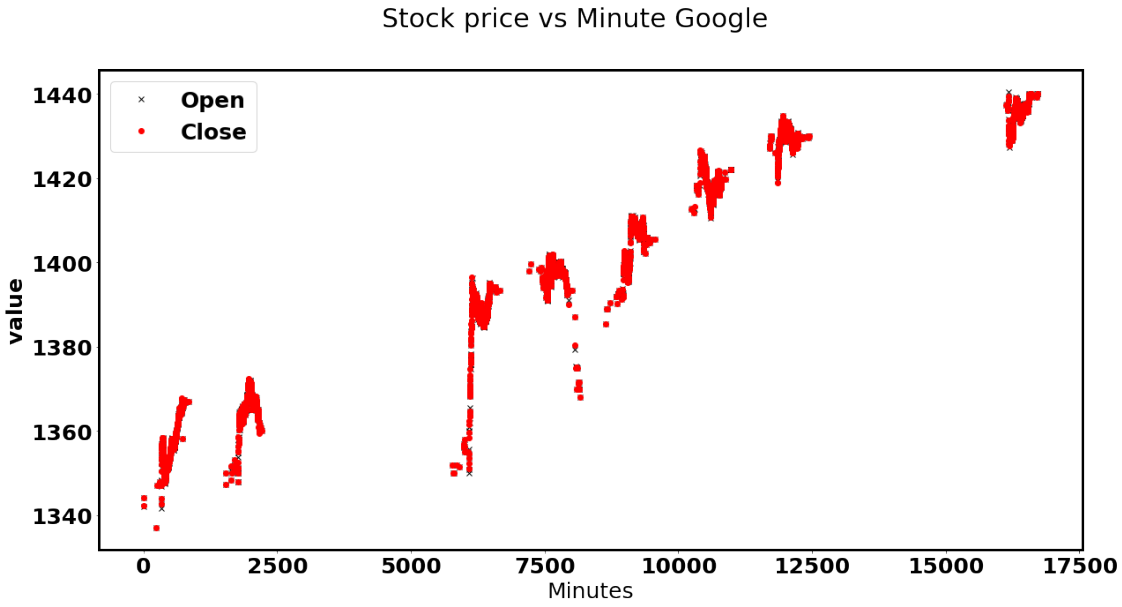
In [5]:
```
import matplotlib
from matplotlib.pylab import rc
font = {'family': 'normal','weight': 'bold',
        'size': 25}

matplotlib.rc('font', **font)
rc('axes', linewidth=3)

plt.plot(df_hf["Open"],"k",label= "Open")
plt.plot(df_hf["Close"],"r",label= "Close")

plt.xlabel("Minutes")
plt.ylabel(r"${\rm value}$")
plt.suptitle("Stock price vs Minute Google including close market")
plt.legend()

fig = matplotlib.pyplot.gcf()
fig.set_size_inches(20,10)
```



In [6]:
```
df_hf["DateTime"] = pd.to_datetime(df_hf["DateTime"])
delta_min = (df_hf["DateTime"]-df_hf["DateTime"][0]).dt.total_seconds()//60
```

In [7]:
```
import matplotlib
from matplotlib.pylab import rc
font = {'family': 'normal','weight': 'bold',
        'size': 25}
```

```
matplotlib.rc('font', **font)
rc('axes', linewidth=3)

plt.plot(delta_min,df_hf["Open"],"kx",label= "Open")
plt.plot(delta_min,df_hf["Close"],"ro",label= "Close")

plt.xlabel("Minutes")
plt.ylabel(r"${\rm value}$")
plt.suptitle("Stock price vs Minute Google")
plt.legend()

fig = matplotlib.pyplot.gcf()
fig.set_size_inches(20,10)
```



In [8]:
```
# This time Try DQN on df and df_hf

class Environment1:

    def __init__(self, data, history_t=90):
        self.data = data
        self.history_t = history_t
        self.reset()

    def reset(self):
        self.t = 0
        self.done = False
        self.profits = 0
```

```
        self.positions = []
        self.position_value = 0
        self.history = [0 for _ in range(self.history_t)]
        return [self.position_value] + self.history # obs

    def step(self, act):
        reward = 0

        # act = 0: stay, 1: buy, 2: sell
        if act == 1:
            self.positions.append(self.data.iloc[self.t, :]['Close'])
        elif act == 2: # sell
            if len(self.positions) == 0:
                reward = -1
            else:
                profits = 0
                for p in self.positions:
                    profits += (self.data.iloc[self.t, :]['Close'] - p)
                reward += profits
                self.profits += profits
                self.positions = []

        # set next time
        self.t += 1
        self.position_value = 0
        for p in self.positions:
            self.position_value += (self.data.iloc[self.t, :]['Close'] - p)
        self.history.pop(0)
        self.history.append(self.data.iloc[self.t, :]['Close'] - self.data.iloc[(self.

        # clipping reward
        if reward > 0:
            reward = 1
        elif reward < 0:
            reward = -1

        return [self.position_value] + self.history, reward, self.done
```

In [9]:
```
# df

df['Date'] = pd.to_datetime(df['Date'])
data = df.set_index('Date')
env = Environment1(df)


# df_hf
```

```
        data_hf = df_hf.set_index('DateTime')
        env_hf = Environment1(df_hf)
```

```
In [10]: # Train test split :) Here I didn't shuffle dataset since we want to check whether we

         delta_split = int(df.shape[0]*0.7)

         train = df[:delta_split]
         test = df[delta_split:]
         print("Training date length",len(train),"testing date length", len(test))
```

```
Training date length 2333 testing date length 1000
```

```
In [11]: delta_split = int(df_hf.shape[0]*0.7)
         train_hf = df_hf[:delta_split]
         test_hf = df_hf[delta_split:]
         print("Training date length hf",len(train_hf),"testing date length hf", len(test_hf))
```

```
Training date length hf 2275 testing date length hf 976
```

```
In [13]: import chainer
         import chainer.functions as F
         import chainer.links as L
         import copy
         import time



         def train_dqn(env):

             class Q_Network(chainer.Chain):

                 def __init__(self, input_size, hidden_size, output_size):
                     super(Q_Network, self).__init__(
                         fc1 = L.Linear(input_size, hidden_size),
                         fc2 = L.Linear(hidden_size, hidden_size),
                         fc3 = L.Linear(hidden_size, output_size)
                     )

                 def __call__(self, x):
                     h = F.relu(self.fc1(x))
                     h = F.relu(self.fc2(h))
                     y = self.fc3(h)
                     return y

                 def reset(self):
```

```
        self.zerograds()

Q = Q_Network(input_size=env.history_t+1, hidden_size=100, output_size=3)
# GPU Option
#Q.to_gpu()

Q_ast = copy.deepcopy(Q)
optimizer = chainer.optimizers.Adam()
optimizer.setup(Q)

# Hyper-parameters
epoch_num = 25
step_max = len(env.data)-1
memory_size = 200
batch_size = 20
epsilon = 1.0
epsilon_decrease = 1e-3
epsilon_min = 0.1
start_reduce_epsilon = 200
train_freq = 10
update_q_freq = 20
gamma = 0.97
show_log_freq = 5


memory = []
total_step = 0
total_rewards = []
total_losses = []

start = time.time()
for epoch in range(epoch_num):

    pobs = env.reset()
    step = 0
    done = False
    total_reward = 0
    total_loss = 0

    while not done and step < step_max:

        # select act
        pact = np.random.randint(3)
        if np.random.rand() > epsilon:
            pact = Q(np.array(pobs, dtype=np.float32).reshape(1, -1))
            pact = np.argmax(pact.data)

        # act
```

```
        obs, reward, done = env.step(pact)

        # add memory
        memory.append((pobs, pact, reward, obs, done))
        if len(memory) > memory_size:
            memory.pop(0)

        # train or update q
        if len(memory) == memory_size:
            if total_step % train_freq == 0:
                shuffled_memory = np.random.permutation(memory)
                memory_idx = range(len(shuffled_memory))
                for i in memory_idx[::batch_size]:
                    batch = np.array(shuffled_memory[i:i+batch_size])
                    b_pobs = np.array(batch[:, 0].tolist(), dtype=np.float32).resh
                    b_pact = np.array(batch[:, 1].tolist(), dtype=np.int32)
                    b_reward = np.array(batch[:, 2].tolist(), dtype=np.int32)
                    b_obs = np.array(batch[:, 3].tolist(), dtype=np.float32).resha
                    b_done = np.array(batch[:, 4].tolist(), dtype=np.bool)

                    q = Q(b_pobs)
                    maxq = np.max(Q_ast(b_obs).data, axis=1)
                    target = copy.deepcopy(q.data)
                    for j in range(batch_size):
                        target[j, b_pact[j]] = b_reward[j]+gamma*maxq[j]*(not b_do
                    Q.reset()
                    loss = F.mean_squared_error(q, target)
                    total_loss += loss.data
                    loss.backward()
                    optimizer.update()

                if total_step % update_q_freq == 0:
                    Q_ast = copy.deepcopy(Q)

        # epsilon
        if epsilon > epsilon_min and total_step > start_reduce_epsilon:
            epsilon -= epsilon_decrease

        # next step
        total_reward += reward
        pobs = obs
        step += 1
        total_step += 1

    total_rewards.append(total_reward)
    total_losses.append(total_loss)

    if (epoch+1) % show_log_freq == 0:
```

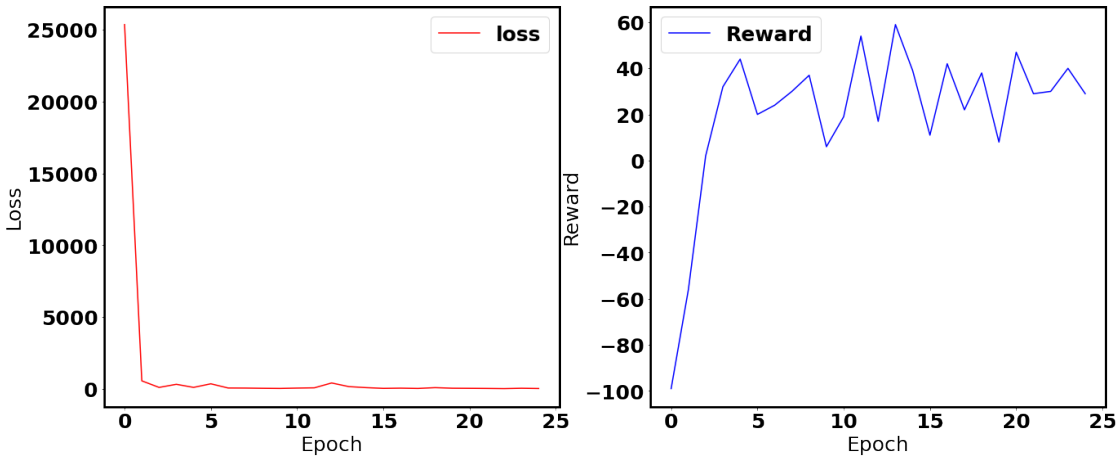
```
                    log_reward = sum(total_rewards[((epoch+1)-show_log_freq):])/show_log_freq
                    log_loss = sum(total_losses[((epoch+1)-show_log_freq):])/show_log_freq
                    elapsed_time = time.time()-start
                    print('\t'.join(map(str, [epoch+1, epsilon, total_step, log_reward, log_l
                    start = time.time()

            return Q, total_losses, total_rewards
```

In [14]:
```
# Train on google_daily:

# Q, total_losses, total_rewards = train_dqn(Environment1(train))
```

In [15]:
```
# Train on google_per_minute:
# Epoch, epsilon (Randomness in your strategy), steps,log[reward],log[loss], elapsed
Q, total_losses, total_rewards = train_dqn(Environment1(train_hf))
```

```
5       0.0999999999999992      11370       -15.4       5297.414914591704       113.0948
10       0.0999999999999992      22740       23.4       121.3875683060789       111.3312
15       0.0999999999999992      34110       37.6       169.700724235503       114.35927
20       0.0999999999999992      45480       24.2       62.293562510469926       111.294
25       0.0999999999999992      56850       35.0       44.13690594714135       113.9022
```

In [20]:
```
plt.subplot(1,2,1)
plt.plot(total_losses,"r",label="loss")
plt.xlabel("Epoch")
plt.ylabel("Loss")
plt.legend()

plt.subplot(1,2,2)
plt.plot(total_rewards,"b",label="Reward")
plt.xlabel("Epoch")
plt.ylabel("Reward")

plt.legend()



fig = matplotlib.pyplot.gcf()

fig.set_size_inches(22,9)
plt.legend()
plt.show()
```

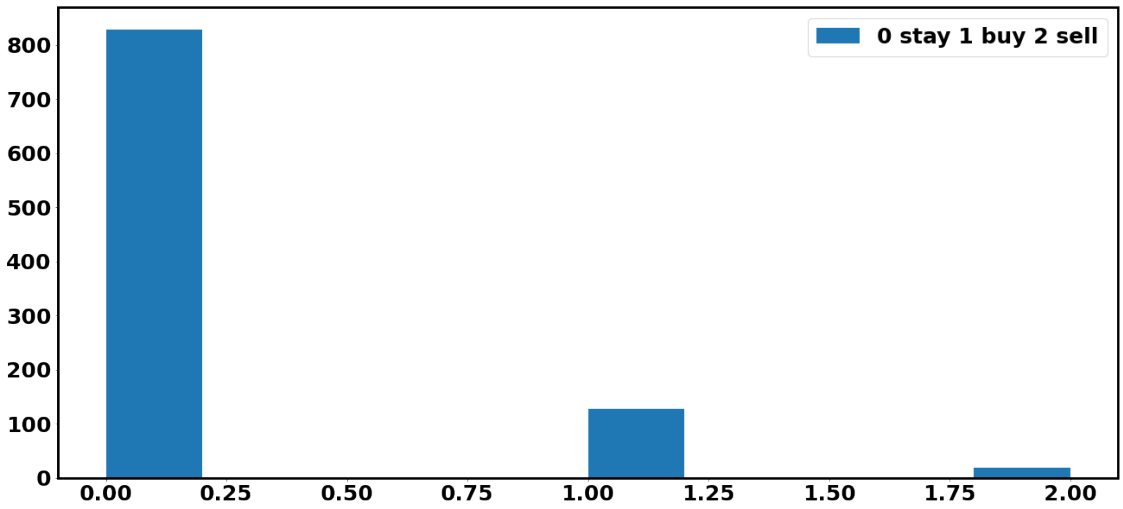


In [40]:
```
# Testing :)

# test
test_env = Environment1(test_hf)


# test
pobs = test_env.reset()
test_acts = []
test_rewards = []

for _ in range(len(test_env.data) - 1):
    pact = Q(np.array(pobs, dtype=np.float32).reshape(1, -1))
    pact = np.argmax(pact.data)
    test_acts.append(pact)

    obs, reward, done = test_env.step(pact)
    test_rewards.append(reward)

    pobs = obs

test_profits = test_env.profits


print("Test profit =%.4f in %d steps start price %.4f"%(test_profits,test_env.data.sha
```

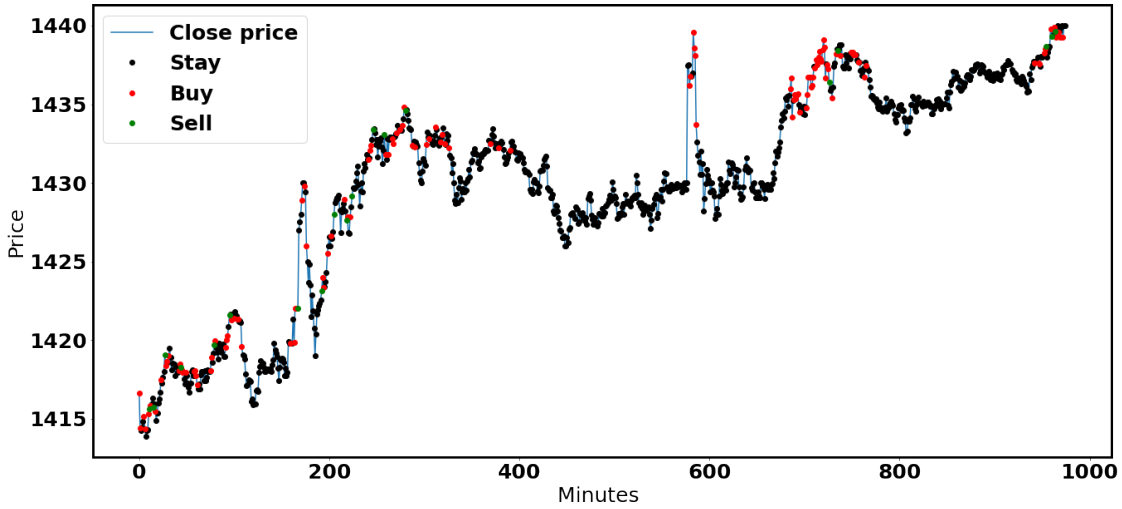
```
Test profit =117.8347 in 976 steps start price 1416.0400
```

In [69]:
```
# Our actions in testing set:
# Here 0 means stay, 1 means buy 2 means sell
plt.hist(test_acts,label = "0 stay 1 buy 2 sell")
```

```
fig = matplotlib.pyplot.gcf()

fig.set_size_inches(22,10)
plt.legend()
plt.show()
```



In [68]:
```
plt.subplot(1,1,1)
y = test_env.data["Close"].values[1:]

test_acts = np.array(test_acts)

m0 = test_acts==0
m1 = test_acts==1
m2 = test_acts==2
plt.plot(np.arange(0,len(y),1),y,label = "Close price")
plt.plot(np.arange(0,len(y),1)[m0],y[m0],"ko",label = "Stay")
plt.plot(np.arange(0,len(y),1)[m1],y[m1],"ro",label = "Buy")
plt.plot(np.arange(0,len(y),1)[m2],y[m2],"go",label = "Sell")

plt.xlabel("Minutes ")
plt.ylabel("Price")



fig = matplotlib.pyplot.gcf()

fig.set_size_inches(22,10)
plt.legend()
plt.show()
```


Close price
Stay
Buy
Sell
1440
1435
1430
1425
1420
1415
Price
0
200
400
600
800
1000
Minutes

In [ ]:

In [ ]:

In [ ]:

In [ ]: