Assignment4

March 26, 2025

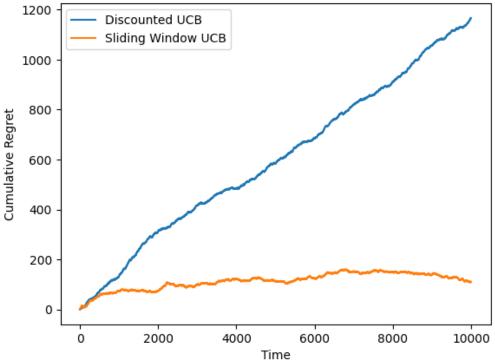
0.1 Code Assignment 1

```
[]: # Deepseek assignment 1
     import numpy as np
     import matplotlib.pyplot as plt
     class DynamicPricingEnv:
         def __init__(self, prices, change_points, reward_probabilities):
             self.prices = prices
             self.change_points = change_points
             self.reward_probabilities = reward_probabilities
             self.current_phase = 0
             self.t = 0
         def step(self, price_index):
             if self.t in self.change_points:
                 self.current_phase = min(self.current_phase + 1, len(self.
      →reward_probabilities) - 1)
             self.t += 1
             return np.random.rand() < self.reward_probabilities[self.</pre>
      →current_phase][price_index]
         def reset(self):
             self.current_phase = 0
             self.t = 0
     class DiscountedUCB:
         def __init__(self, n_arms, gamma=0.99):
             self.n_arms = n_arms
             self.gamma = gamma
             self.counts = np.zeros(n_arms)
             self.values = np.zeros(n_arms)
         def select_arm(self):
             for arm in range(self.n_arms):
                 if self.counts[arm] == 0:
                     return arm
             total = np.sum(self.counts)
```

```
ucb_values = (self.values / self.counts) + np.sqrt((2 * np.log(total)) /
 ⇔ self.counts)
       return np.argmax(ucb_values)
   def update(self, arm, reward):
       self.counts *= self.gamma
       self.values *= self.gamma
        self.counts[arm] += 1
        self.values[arm] += reward
class SlidingWindowUCB:
   def __init__(self, n_arms, window_size=50):
       self.n_arms = n_arms
        self.window_size = window_size
        self.rewards = [[] for _ in range(n_arms)]
   def select_arm(self):
        for arm in range(self.n_arms):
            if len(self.rewards[arm]) == 0:
                return arm
       total pulls = sum(len(rewards) for rewards in self.rewards)
       ucb values = []
        for arm in range(self.n_arms):
            cnt = len(self.rewards[arm])
            avg = sum(self.rewards[arm]) / cnt
            ucb = avg + np.sqrt((2 * np.log(total_pulls)) / cnt)
            ucb_values.append(ucb)
        return np.argmax(ucb_values)
   def update(self, arm, reward):
        self.rewards[arm].append(reward)
        if len(self.rewards[arm]) > self.window size:
            self.rewards[arm].pop(0)
# Experiment Setup
prices = [5, 10, 15, 20]
change_points = [2000, 4000, 6000, 8000]
reward_probabilities = [
    [0.3, 0.5, 0.2, 0.1], # Phase 1
    [0.2, 0.6, 0.3, 0.15], # Phase 2
    [0.1, 0.4, 0.5, 0.3], # Phase 3
    [0.25, 0.35, 0.3, 0.2], # Phase 4
    [0.15, 0.5, 0.25, 0.4] # Phase 5
]
env = DynamicPricingEnv(prices, change_points, reward_probabilities)
d_ucb = DiscountedUCB(len(prices))
```

```
sw_ucb = SlidingWindowUCB(len(prices))
T = 10000 # Total time steps
regrets_d = []
regrets_sw = []
for t in range(T):
    # Discounted UCB
    arm_d = d_ucb.select_arm()
    reward_d = env.step(arm_d)
    d_ucb.update(arm_d, reward_d)
    regrets_d.append(max(reward_probabilities[env.current_phase]) - reward_d)
    # Sliding Window UCB
    arm_sw = sw_ucb.select_arm()
    reward_sw = env.step(arm_sw)
    sw_ucb.update(arm_sw, reward_sw)
    regrets_sw.append(max(reward_probabilities[env.current_phase]) - reward_sw)
# Plot Regrets
plt.plot(np.cumsum(regrets_d), label='Discounted UCB')
plt.plot(np.cumsum(regrets_sw), label='Sliding Window UCB')
plt.xlabel("Time")
plt.ylabel("Cumulative Regret")
plt.legend()
plt.title("Comparison of Discounted UCB and Sliding Window UCB in Dynamic UCB and Sliding Window UCB in Dynamic
 →Pricing")
plt.show()
```

Comparison of Discounted UCB and Sliding Window UCB in Dynamic Pricing



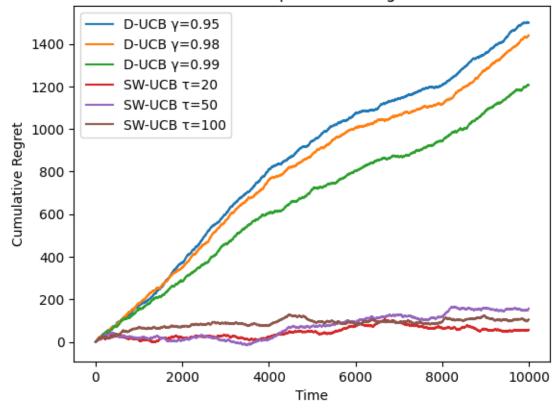
0.2 Assignment 4.2

```
[8]: # ===== Q2 =====
     # Q2: Test different values of gamma (D-UCB) and tau (SW-UCB)
     gammas = [0.95, 0.98, 0.99]
     taus = [20, 50, 100]
     for gamma in gammas:
         d_ucb = DiscountedUCB(len(prices), gamma=gamma)
         env.reset()
         regrets = []
         for t in range(T):
             arm = d_ucb.select_arm()
             reward = env.step(arm)
             d_ucb.update(arm, reward)
             regrets.append(max(reward_probabilities[env.current_phase]) - reward)
         plt.plot(np.cumsum(regrets), label=f'D-UCB ={gamma}')
     for tau in taus:
         sw_ucb = SlidingWindowUCB(len(prices), window_size=tau)
         env.reset()
         regrets = []
```

```
for t in range(T):
    arm = sw_ucb.select_arm()
    reward = env.step(arm)
    sw_ucb.update(arm, reward)
    regrets.append(max(reward_probabilities[env.current_phase]) - reward)
    plt.plot(np.cumsum(regrets), label=f'SW-UCB ={tau}')

plt.xlabel("Time")
plt.ylabel("Cumulative Regret")
plt.title("Effect of and on Regret")
plt.legend()
plt.show()
```

Effect of γ and τ on Regret

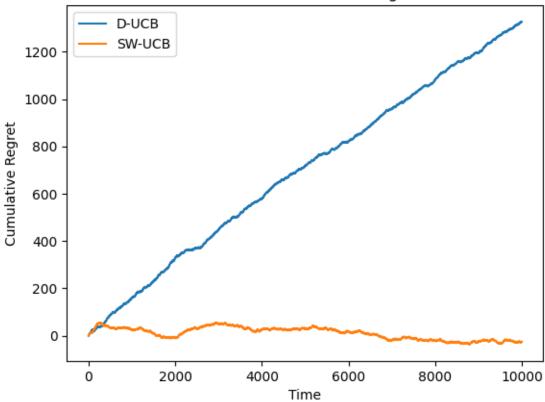


0.3 Assignment 4.3

```
[9]: # ===== Q3 =====
  # Q3: Irregular (unknown) change points
  import random
  change_points = sorted(random.sample(range(500, 9500), 4))
```

```
env = DynamicPricingEnv(prices, change_points, reward_probabilities)
sw_ucb = SlidingWindowUCB(len(prices), window_size=50)
d_ucb = DiscountedUCB(len(prices), gamma=0.99)
regrets_sw, regrets_d = [], []
for t in range(T):
   # SW-UCB
   arm_sw = sw_ucb.select_arm()
   reward_sw = env.step(arm_sw)
   sw_ucb.update(arm_sw, reward_sw)
   regrets_sw.append(max(reward_probabilities[env.current_phase]) - reward_sw)
   # D-UCB
   arm_d = d_ucb.select_arm()
   reward_d = env.step(arm_d)
   d_ucb.update(arm_d, reward_d)
   regrets_d.append(max(reward_probabilities[env.current_phase]) - reward_d)
plt.plot(np.cumsum(regrets_d), label="D-UCB")
plt.plot(np.cumsum(regrets_sw), label="SW-UCB")
plt.title("Robustness to Unknown Change Points")
plt.xlabel("Time")
plt.ylabel("Cumulative Regret")
plt.legend()
plt.show()
```

Robustness to Unknown Change Points



0.4 Assignment 4.4

```
[10]: # ===== Q4 =====
  # Q4: Increase number of arms
  prices = [5, 10, 15, 20, 25, 30]
  reward_probabilities = [
      [0.3, 0.5, 0.2, 0.1, 0.05, 0.02],
      [0.2, 0.6, 0.3, 0.15, 0.1, 0.03],
      [0.1, 0.4, 0.5, 0.3, 0.2, 0.1],
      [0.25, 0.35, 0.3, 0.2, 0.15, 0.12],
      [0.15, 0.5, 0.25, 0.4, 0.3, 0.18]
]
  env = DynamicPricingEnv(prices, change_points, reward_probabilities)
  d_ucb = DiscountedUCB(len(prices), gamma=0.99)
  sw_ucb = SlidingWindowUCB(len(prices), window_size=50)

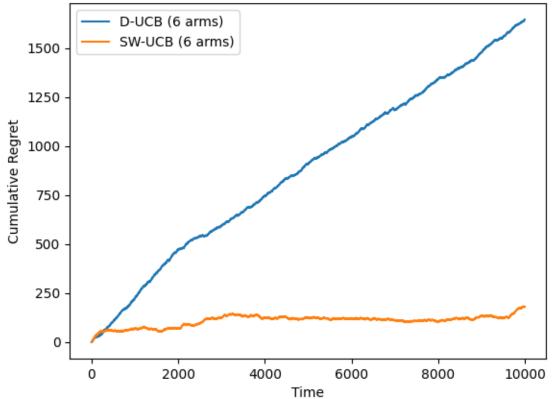
  regrets_sw, regrets_d = [], []
  for t in range(T):
      arm_sw = sw_ucb.select_arm()
```

```
reward_sw = env.step(arm_sw)
    sw_ucb.update(arm_sw, reward_sw)
    regrets_sw.append(max(reward_probabilities[env.current_phase]) - reward_sw)

arm_d = d_ucb.select_arm()
    reward_d = env.step(arm_d)
    d_ucb.update(arm_d, reward_d)
    regrets_d.append(max(reward_probabilities[env.current_phase]) - reward_d)

plt.plot(np.cumsum(regrets_d), label="D-UCB (6 arms)")
plt.plot(np.cumsum(regrets_sw), label="SW-UCB (6 arms)")
plt.title("Effect of More Price Options on Regret")
plt.xlabel("Time")
plt.ylabel("Cumulative Regret")
plt.legend()
plt.show()
```

Effect of More Price Options on Regret

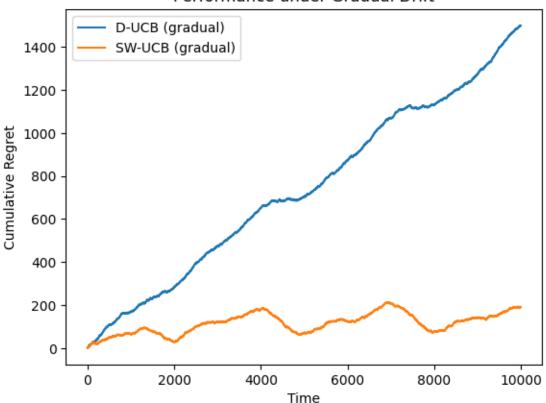


0.5 Assignment 4.5

```
[11]:  # ===== Q5 =====
      # Q5: Gradually changing reward probabilities
      class GradualEnv:
          def __init__(self, prices):
              self.prices = prices
              self.t = 0
          def step(self, price_index):
              probs = self.get_probs(self.t)
              self.t += 1
              return np.random.rand() < probs[price_index]</pre>
          def get_probs(self, t):
              return [
                  0.3 + 0.2 * np.sin(2 * np.pi * t / 5000),
                  0.5 + 0.1 * np.cos(2 * np.pi * t / 3000),
                  0.2 + 0.1 * np.sin(2 * np.pi * t / 2000),
                  0.1 + 0.1 * np.cos(2 * np.pi * t / 2500)
              ]
          def reset(self):
              self.t = 0
      env = GradualEnv(prices=[5, 10, 15, 20])
      d_ucb = DiscountedUCB(4, gamma=0.99)
      sw_ucb = SlidingWindowUCB(4, window_size=50)
      regrets_d, regrets_sw = [], []
      for t in range(T):
          probs = env.get_probs(t)
          arm_d = d_ucb.select_arm()
          reward d = env.step(arm d)
          d_ucb.update(arm_d, reward_d)
          regrets_d.append(max(probs) - reward_d)
          arm_sw = sw_ucb.select_arm()
          reward_sw = env.step(arm_sw)
          sw_ucb.update(arm_sw, reward_sw)
          regrets_sw.append(max(probs) - reward_sw)
      plt.plot(np.cumsum(regrets_d), label="D-UCB (gradual)")
      plt.plot(np.cumsum(regrets_sw), label="SW-UCB (gradual)")
      plt.title("Performance under Gradual Drift")
      plt.xlabel("Time")
```

```
plt.ylabel("Cumulative Regret")
plt.legend()
plt.show()
```

Performance under Gradual Drift



```
[13]: !jupyter nbconvert --to pdf Assignment4.ipynb

[NbConvertApp] Converting notebook Assignment4.ipynb to pdf
[NbConvertApp] Support files will be in Assignment4_files\
[NbConvertApp] Making directory .\Assignment4_files
[NbConvertApp] Writing 41481 bytes to notebook.tex
[NbConvertApp] Building PDF
[NbConvertApp] Running xelatex 3 times: ['xelatex', 'notebook.tex', '-quiet']
[NbConvertApp] Running bibtex 1 time: ['bibtex', 'notebook']
[NbConvertApp] WARNING | b had problems, most likely because there were no citations
[NbConvertApp] PDF successfully created
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[NbConvertApp] Writing 212257 bytes to Assignment4.pdf