```
import numpy as np
    import matplotlib.pyplot as plt
    import random
    class ThreeArmedBanditEnv:
        def __init__(self):
             self.probabilities = [0.4, 0.4, 0.4]
             self.n_arms = len(self.probabilities)
             self.best_action = np.argmax(self.probabilities)
        def step(self, action):
    if action < 0 or action >= self.n_arms:
                raise ValueError("Invalid action")
             reward = 1 if np.random.rand() < self.probabilities[action] else 0</pre>
             return None, reward, False, False, {}
         def reset(self):
             return None
    def run_ucb(env, num_rounds=1000):
         n_arms = env.n_arms
         Q = np.zeros(n_arms)
         N = np.zeros(n_arms)
         rewards = []
        arm_selections = np.zeros(n_arms)
         for t in range(1, num_rounds + 1):
    ucb_values = np.zeros(n_arms)
             for a in range(n_arms):
                if N[a] == 0:
                     ucb_values[a] = float('inf')
                 else:
                     ucb_values[a] = Q[a] + 2 * np.sqrt(np.log(t) / N[a])
             action = np.argmax(ucb_values)
             _, reward, _, _, _ = env.step(action)
N[action] += 1
             Q[action] += (reward - Q[action]) / N[action]
             rewards.append(reward)
             arm_selections[action] += 1
         return Q, rewards, arm_selections
    def run_thompson_sampling(env, num_rounds=1000):
         n_arms = env.n_arms
         alpha = np.ones(n_arms)
         beta = np.ones(n_arms)
         rewards = []
         arm_selections = np.zeros(n_arms)
         for _ in range(num_rounds):
             sampled_probabilities = [np.random.beta(alpha[a], beta[a]) for a in range(n_arms)]
             action = np.argmax(sampled_probabilities)
             _, reward, _, _, _ = env.step(action)
if reward == 1:
```

```
if reward == 1:
                 alpha[action] += 1
             else:
                  beta[action] += 1
             rewards.append(reward)
             arm_selections[action] += 1
         Q = alpha / (alpha + beta)
         return Q, rewards, arm_selections
     env = ThreeArmedBanditEnv()
     num_rounds = 1000
     q_ucb, rewards_ucb, selections_ucb = run_ucb(env, num_rounds=num_rounds)
     q_ts, rewards_ts, selections_ts = run_thompson_sampling(env, num_rounds=num_rounds)
     print("UCB Results:")
     print(" Estimated Q-values:", q_ucb)
print(" Arm Selections:", selections_ucb)
     print(" Total Reward:", sum(rewards_ucb))
     print("\nThompson Sampling Results:")
     print(" Estimated Q-values:", q_ts)
     print(" Arm Selections:", selections_ts)
     print(" Total Reward:", sum(rewards_ts))
→ UCB Results:
       Estimated Q-values: [0.42342342 0.41233766 0.43454039]
       Arm Selections: [333. 308. 359.]
       Total Reward: 424
     Thompson Sampling Results:
       Estimated Q-values: [0.38557214 0.38
                                                    0.36881188]
       Arm Selections: [400. 198. 402.]
       Total Reward: 377
[7] import matplotlib.pyplot as plt
    import numpy as np
    plt.figure(figsize=(12, 6))
    plt.plot(np.cumsum(rewards_ucb), label='UCB')
    plt.plot(np.cumsum(rewards_ts), label='Thompson Sampling')
    plt.xlabel('Round')
    plt.ylabel('Cumulative Reward')
    plt.title('Cumulative Reward over Rounds: UCB vs. Thompson Sampling')
    plt.legend()
    plt.grid(True)
    plt.show()
    arms = ['Video Lectures', 'Interactive Quizzes', 'Gamified Modules']
    x = np.arange(len(arms))
    fig, ax = plt.subplots(1, 2, figsize=(14, 6))
    ax[0].bar(x, selections_ucb, tick_label=arms)
    ax[0].set_ylabel('Number of Selections')
    ax[0].set_title('UCB Arm Selections')
    ax[1].bar(x, selections_ts, tick_label=arms, color='orange')
    ax[1].set_ylabel('Number of Selections')
    ax[1].set_title('Thompson Sampling Arm Selections')
    plt.tight_layout()
    plt.show()
```

