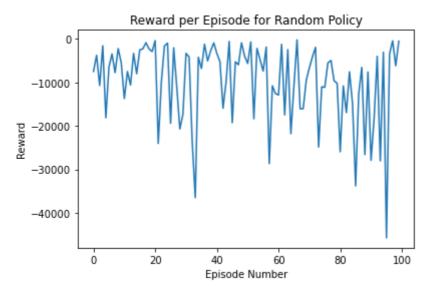
```
import gym
In [3]:
          import random
          import numpy as np
          import time
          import math
          import matplotlib.pyplot as plt
          from sklearn.preprocessing import KBinsDiscretizer
In [4]:
         #Tax Random Policy
         env = gym.make("Taxi-v3").env
         env.reset()
Out[4]: 173
In [5]:
         # Shows space sizes
         env.render()
          # We can see below that the pipe presentes the walls, the yellow block represents th
          #The RGYB represent the possible pickup/destinations.
          #Printing Action Space and State Sapce
          print("Action Space {}".format(env.action_space))
          print("State Space {}".format(env.observation_space))
         |R: | : :G|
         | : | : : |
          ::::|
          |:|:
         |Y| : |B: |
         Action Space Discrete(6)
         State Space Discrete(500)
         # (taxi row, taxi column, passenger index, destination index)
In [6]:
          # Here we find the state of where our Taxi is currently at.
          state = env.encode(3, 3, 2, 0)
          print("State:", state)
         env.s = state
          env.render()
         State: 368
         +----+
         |R: | : :G|
         | : | : : |
          ::::|
          | | : | : |
         |Y| : |B: |
In [7]:
         #reward table structure {action: [(probability, nextstate, reward, done)]}
         env.P[368]
Out[7]: {0: [(1.0, 468, -1, False)],
          1: [(1.0, 268, -1, False)],
          2: [(1.0, 388, -1, False)],
         3: [(1.0, 368, -1, False)],
4: [(1.0, 368, -10, False)],
5: [(1.0, 368, -10, False)]}
         # set environment to illustration's state
In [8]:
          env.env.s = 368
```

```
epochs = 0
          penalties, reward = 0, 0
          frames = [] # for animation
          done = False
          while not done:
              action = env.action_space.sample()
              state, reward, done, info = env.step(action)
              if reward == -10:
                  penalties += 1
              # Put each rendered frame into dict for animation
              frames.append({
                  'frame': env.render(mode='ansi'),
                  'state': state,
                  'action': action,
                  'reward': reward
              epochs += 1
          print("Timesteps taken: {}".format(epochs))
          print("Penalties incurred: {}".format(penalties))
         Timesteps taken: 594
         Penalties incurred: 197
          from IPython.display import clear_output
 In [9]:
          from time import sleep
          def print_frames(frames):
              for i, frame in enumerate(frames):
                  clear_output(wait=True)
                  print(frame['frame'])#.getvalue())
                  print(f"Timestep: {i + 1}")
                  print(f"State: {frame['state']}")
                  print(f"Action: {frame['action']}")
                  print(f"Reward: {frame['reward']}")
                  sleep(.1)
          print_frames(frames)
          |R: | : :G| |
          |\cdot|\cdot|\cdot|
          |Y| : |B: |
           (Dropoff)
         Timestep: 594
         State: 0
         Action: 5
         Reward: 20
In [12]:
         #Random Policy for Testing Random Agent
          env.s = 368
          total epochs = 0
          total penalties = 0
```

```
episodes = 100
          rand_episode_reward_list = []
          episode reward = 0
          #loop through 100 episodes (100 different passengers)
          for _ in range(episodes):
              #set current state to starting environment
              state = env.reset()
              epochs, penalties, reward = 0, 0, 0
              #done = have successfully dropped off passenger/completed episode
              done = False
              #while episode incomplete
              while not done:
                  #select a random action
                  action = env.action_space.sample()
                  state, reward, done, info = env.step(action)
                  episode_reward = episode_reward + reward
                  #set penalty based on reward value for action
                  if reward == -10:
                      penalties += 1
                  #increment episode
                  epochs += 1
              #once 100 episode are complete, sum total penalties and episodes
              total_penalties += penalties
              total_epochs += epochs
              rand_episode_reward_list.append(episode_reward)
              episode reward = 0
          #store metrics
          rand_average_time = total_epochs / episodes
          rand_average_pen = total_penalties / episodes
          #print evaluation of agents performance
          print(f"Results after {episodes} episodes:")
          print(f"Average timesteps per episode: {rand average time}")
          print(f"Average penalties per episode: {rand_average_pen}")
         Results after 100 episodes:
         Average timesteps per episode: 2511.69
         Average penalties per episode: 812.03
In [14]:
         #plot the reward for each episode - random policy
          plt.plot(rand_episode_reward_list)
          plt.ylabel('Reward')
          plt.xlabel('Episode Number')
          plt.title("Reward per Episode for Random Policy")
          plt.show()
```



In []: