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In [1]: !pip install gym
        import gym
In [2]:
        import random
        import numpy as np
        import torch
        import torch.nn as nn
        import torch.optim as optim
In [3]: # Define Deep Q-Network (DQN) model
        class DQN(nn.Module):
            def __init__(self, input_size, hidden_size, output_size):
                super(DQN, self).__init__()
                self.fc1 = nn.Linear(input_size, hidden_size)
                self.fc2 = nn.Linear(hidden size, hidden size)
                self.fc3 = nn.Linear(hidden_size, output_size)
            def forward(self, x):
                x = torch.relu(self.fc1(x))
                x = torch.relu(self.fc2(x))
                return self.fc3(x)
In [4]: # Define replay buffer
        class ReplayBuffer:
            def __init__(self, capacity):
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class ReplayBuffer:
    def __init__(self, capacity):
        self.capacity = capacity
        self.buffer = []

    def add(self, state, action, reward, next_state, done):
        experience = (state, action, reward, next_state, done)
        self.buffer.append(experience)
        if len(self.buffer) > self.capacity:
            self.buffer.pop(0)

    def sample(self, batch_size):
        return random.sample(self.buffer, batch_size)
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In [5]: # Define the DQN Agent
        class DQNAgent:
            def __init__(self, input_size, output_size, hidden_size=64, learning_ra
                         epsilon_start=1.0, epsilon_decay=0.995, epsilon_min=0.01,
                         batch_size=64):
                self.input_size = input_size
                self.output_size = output_size
                self.hidden_size = hidden_size
                self.learning_rate = learning_rate
                self.gamma = gamma
                self.epsilon = epsilon_start
                self.epsilon decay = epsilon decay
                self.epsilon_min = epsilon_min
                self.replay_buffer = ReplayBuffer(replay_buffer_capacity)
                self.batch_size = batch_size
                self.q_network = DQN(input_size, hidden_size, output_size)
                self.target_network = DQN(input_size, hidden_size, output_size)
                self.target_network.load_state_dict(self.q_network.state_dict())
                self.optimizer = optim.Adam(self.q_network.parameters(), lr=learnin
                self.loss_function = nn.MSELoss()
            def epsilon_greedy_action(self, state):
                if np.random.rand() < self.epsilon:</pre>
                    return np.random.choice(range(self.output_size))
                else:
                    with torch.no_grad():
                        q_values = self.q_network(torch.FloatTensor(state))
                        return torch.argmax(q_values).item()
            def train(self, state, action, reward, next_state, done):
                self.replay_buffer.add(state, action, reward, next_state, done)
                if len(self.replay_buffer.buffer) > self.batch_size:
                    batch = self.replay_buffer.sample(self.batch_size)
                    states, actions, rewards, next_states, dones = zip(*batch)
                    states = torch.FloatTensor(states)
                    actions = torch.LongTensor(actions)
                    rewards = torch.FloatTensor(rewards)
                    next_states = torch.FloatTensor(next_states)
                    dones = torch.FloatTensor(dones)
                    q values = self.q network(states)
                    next_q_values = self.target_network(next_states).max(1)[0]
                    target_q_values = rewards + (1 - dones) * self.gamma * next_q_v
                    q_values = q_values.gather(1, actions.unsqueeze(1)).squeeze(1)
                    loss = self.loss_function(q_values, target_q_values.detach())
                    self.optimizer.zero_grad()
                    loss.backward()
                    self.optimizer.step()
                self.epsilon = max(self.epsilon * self.epsilon_decay, self.epsilon_
            def update_target_network(self):
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self.target_network.load_state_dict(self.q_network.state_dict())
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In [6]: # Initialize environment and agent
        env = gym.make('CartPole-v1')
        input size = env.observation space.shape[0]
        output size = env.action space.n
        agent = DQNAgent(input size, output size)
In [7]: # Training
        num_episodes = 1000
        for episode in range(num_episodes):
            state = env.reset()
            if isinstance(state, tuple): # For Gymnasium compatibility
                state = state[0]
            total reward = 0
            done = False
            while not done:
                action = agent.epsilon_greedy_action(state)
                next_state, reward, done, truncated, _ = env.step(action) if hasatt
                if isinstance(next_state, tuple): # For Gymnasium compatibility
                    next_state = next_state[0]
                agent.train(state, action, reward, next state, done)
                state = next_state
                total_reward += reward
            if episode % 100 == 0:
                agent.update target network()
                print(f"Episode {episode}, Total Reward: {total_reward}")
        # Close the environment
        env.close()
        C:\Python310\lib\site-packages\gym\utils\passive_env_checker.py:233: Depre
        cationWarning: `np.bool8` is a deprecated alias for `np.bool `. (Deprecat
        ed NumPy 1.24)
          if not isinstance(terminated, (bool, np.bool8)):
        C:\Users\dell\AppData\Local\Temp\ipykernel_19552\133867816.py:38: UserWarn
        ing: Creating a tensor from a list of numpy.ndarrays is extremely slow. Pl
        ease consider converting the list to a single numpy.ndarray with numpy.arr
        ay() before converting to a tensor. (Triggered internally at C:\actions-ru
        nner\_work\pytorch\pytorch\torch\csrc\utils\tensor_new.cpp:257.)
          states = torch.FloatTensor(states)
        Episode 0, Total Reward: 23.0
        Episode 100, Total Reward: 15.0
        Episode 200, Total Reward: 19.0
        Episode 300, Total Reward: 30.0
        Episode 400, Total Reward: 8.0
        Episode 500, Total Reward: 38.0
        Episode 600, Total Reward: 40.0
        Episode 700, Total Reward: 241.0
        Episode 800, Total Reward: 189.0
        Episode 900, Total Reward: 232.0
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In []: