

```
import random
import torch
import numpy as np
```

```
class ReplayBufferDQN:
```

```
    def __init__(self, buffer_size:int, seed:int=42):
        self.buffer_size = buffer_size
        self.seed = seed
        self.buffer = []
        random.seed(self.seed)
```

```
    def add(self, state:np.ndarray, action:int, reward:float, next_state:np.ndarray
        , done:bool):
```

```
        """
```

```
        Add a new experience to the buffer
```

```
        Args:
```

```
            state (np.ndarray): the current state of shape [n_c,h,w]
```

```
            action (int): the action taken
```

```
            reward (float): the reward received
```

```
            next_state (np.ndarray): the next state of shape [n_c,h,w]
```

```
            done (bool): whether the episode is done
```

```
        """
```

```
        self.buffer.append((state, action, reward, next_state, done))
```

```
        if len(self.buffer) > self.buffer_size:
```

```
            self.buffer.pop(0)
```

```
    def sample(self, batch_size:int, device='cpu'):
```

```
        """
```

```
        Randomly sample a batch of experiences from the replay buffer.
```

```
        Args:
```

```
            batch_size (int): the number of samples to take
```

```
        Returns:
```

```
            states (torch.Tensor): Tensor of shape (batch_size, n_channels, height, width),
```

```
            dtype torch.float32.
```

```
            actions (torch.Tensor): Tensor of shape (batch_size,), dtype torch.int64
```

```
(converted via `.long()`).
```

```
            rewards (torch.Tensor): Tensor of shape (batch_size,), dtype torch.float32.
```

```
            next_states (torch.Tensor): Tensor of shape (batch_size, n_channels, height,
```

```
width), dtype torch.float32.
```

```
            dones (torch.Tensor): Tensor of shape (batch_size,), dtype torch.bool.
```

```
        Notes:
```

```
            1. Use `random.sample` for uniform sampling without replacement.
```

```
            2. Convert NumPy arrays to torch tensors with the correct dtype before moving to
```

```
`device`.
```

```
            3. Use `torch.stack` to combine individual tensors into a batch dimension.
```

```
            4. Keep the output shapes and dtypes consistent.
```

```
        """
```

```
        # ===== YOUR CODE HERE =====
```

```
        # TODO:
```

```
        # 1. sample random indices
```

```
        # 2. collect experiences using the sampled indices
```

```
        # 3. stack and move batches to the specified device, making sure to convert to the
correct dtype
```

```
        # =====
```

```
        # step 1
```

```
        batch = random.sample(self.buffer, batch_size)
```

```
        # step 2
```

```
        states, actions, rewards, next_states, dones = zip(*batch)
```

```
# step 3
states = torch.stack([torch.tensor(s, dtype=torch.float32) for s in
states]).to(device)
actions = torch.tensor(actions, dtype=torch.int64).to(device)
rewards = torch.tensor(rewards, dtype=torch.float32).to(device)
next_states = torch.stack([torch.tensor(ns, dtype=torch.float32) for ns in
next_states]).to(device)
dones = torch.tensor(dones, dtype=torch.bool).to(device)

# ===== YOUR CODE ENDS =====

return states, actions, rewards, next_states, dones

def __len__(self):
    return len(self.buffer)
```