## Q-learning from colab notebook

## Setup

if tight\_layout:

```
In [43]:
          # Python ≥3.5 is required
          import sys
          assert sys.version_info >= (3, 5)
          # Is this notebook running on Colab or Kaggle?
          IS_COLAB = "google.colab" in sys.modules
          IS_KAGGLE = "kaggle_secrets" in sys.modules
          if IS_COLAB or IS_KAGGLE:
              !apt update && apt install -y libpq-dev libsdl2-dev swig xorg-dev xvfb
              %pip install -U tf-agents pyvirtualdisplay
              %pip install -U gym>=0.21.0
              %pip install -U gym[box2d,atari,accept-rom-license]
          # Scikit-Learn ≥0.20 is required
          import sklearn
          assert sklearn.__version__ >= "0.20"
          # TensorFlow ≥2.0 is required
          import tensorflow as tf
          from tensorflow import keras
          assert tf.__version__ >= "2.0"
          if not tf.config.list physical devices('GPU'):
              print("No GPU was detected. CNNs can be very slow without a GPU.")
              if IS_COLAB:
                  print("Go to Runtime > Change runtime and select a GPU hardware accelerator.")
              if IS KAGGLE:
                  print("Go to Settings > Accelerator and select GPU.")
          # Common imports
          import numpy as np
          import os
          # to make this notebook's output stable across runs
          np.random.seed(42)
          tf.random.set_seed(42)
          # To plot pretty figures
          %matplotlib inline
          import matplotlib as mpl
          import matplotlib.pyplot as plt
          mpl.rc('axes', labelsize=14)
          mpl.rc('xtick', labelsize=12)
          mpl.rc('ytick', labelsize=12)
          # To get smooth animations
          import matplotlib.animation as animation
          mpl.rc('animation', html='jshtml')
          # Where to save the figures
          PROJECT_ROOT_DIR = "."
          CHAPTER_ID = "rl"
          IMAGES_PATH = os.path.join(PROJECT_ROOT_DIR, "images", CHAPTER_ID)
          os.makedirs(IMAGES_PATH, exist_ok=True)
          def save_fig(fig_id, tight_layout=True, fig_extension="png", resolution=300):
              path = os.path.join(IMAGES_PATH, fig_id + "." + fig_extension)
              print("Saving figure", fig_id)
```

```
plt.tight_layout()
     plt.savefig(path, format=fig_extension, dpi=resolution)
Get:1 https://cloud.r-project.org/bin/linux/ubuntu bionic-cran40/ InRelease [3,626 B]
Hit:2 http://ppa.launchpad.net/c2d4u.team/c2d4u4.0+/ubuntu bionic InRelease
Get:3 http://security.ubuntu.com/ubuntu bionic-security InRelease [88.7 kB]
Hit:4 http://archive.ubuntu.com/ubuntu bionic InRelease
Get:5 http://archive.ubuntu.com/ubuntu bionic-updates InRelease [88.7 kB]
Hit:6 http://ppa.launchpad.net/cran/libgit2/ubuntu bionic InRelease
Ign:7 https://developer.download.nvidia.com/compute/cuda/repos/ubuntu1804/x86 64 InRelease
Hit:8 http://ppa.launchpad.net/deadsnakes/ppa/ubuntu bionic InRelease
Get:9 http://archive.ubuntu.com/ubuntu bionic-backports InRelease [74.6 kB]
Hit:10 http://ppa.launchpad.net/graphics-drivers/ppa/ubuntu bionic InRelease
Ign:11 https://developer.download.nvidia.com/compute/machine-learning/repos/ubuntu1804/x86_64
InRelease
Hit:12 https://developer.download.nvidia.com/compute/cuda/repos/ubuntu1804/x86_64 Release
Hit:13 https://developer.download.nvidia.com/compute/machine-learning/repos/ubuntu1804/x86_64
Fetched 256 kB in 2s (117 kB/s)
Reading package lists... Done
Building dependency tree
Reading state information... Done
58 packages can be upgraded. Run 'apt list --upgradable' to see them.
Reading package lists... Done
Building dependency tree
Reading state information... Done
swig is already the newest version (3.0.12-1).
libpq-dev is already the newest version (10.19-0ubuntu0.18.04.1).
xorg-dev is already the newest version (1:7.7+19ubuntu7.1).
libsdl2-dev is already the newest version (2.0.8+dfsg1-1ubuntu1.18.04.4).
xvfb is already the newest version (2:1.19.6-1ubuntu4.9).
0 upgraded, 0 newly installed, 0 to remove and 58 not upgraded.
Requirement already satisfied: tf-agents in /usr/local/lib/python3.7/dist-packages (0.11.0)
Requirement already satisfied: pyvirtualdisplay in /usr/local/lib/python3.7/dist-packages (2.2)
Requirement already satisfied: absl-py>=0.6.1 in /usr/local/lib/python3.7/dist-packages (from t
f-agents) (0.12.0)
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s) (7.1.2)
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gents) (1.15.0)
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m tf-agents) (0.5.0)
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tf-agents) (3.17.3)
Requirement already satisfied: cloudpickle>=1.3 in /usr/local/lib/python3.7/dist-packages (from
tf-agents) (1.3.0)
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-agents) (1.13.3)
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ges (from gym>=0.17.0->tf-agents) (4.8.2)
Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-packages (from import
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orflow-probability>=0.14.1->tf-agents) (0.4.0)
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flow-probability>=0.14.1->tf-agents) (4.4.2)
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ow-probability>=0.14.1->tf-agents) (0.1.6)
Requirement already satisfied: EasyProcess in /usr/local/lib/python3.7/dist-packages (from pyvi
rtualdisplay) (0.3)
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Requirement already satisfied: gym[accept-rom-license,atari,box2d] in /usr/local/lib/python3.7/
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Requirement already satisfied: cloudpickle>=1.2.0 in /usr/local/lib/python3.7/dist-packages (fr
om gym[accept-rom-license,atari,box2d]) (1.3.0)
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Requirement already satisfied: box2d-py==2.3.5 in /usr/local/lib/python3.7/dist-packages (from
gym[accept-rom-license,atari,box2d]) (2.3.5)
Requirement already satisfied: autorom[accept-rom-license]~=0.4.2 in /usr/local/lib/python3.7/d
ist-packages (from gym[accept-rom-license,atari,box2d]) (0.4.2)
Requirement already satisfied: importlib-resources in /usr/local/lib/python3.7/dist-packages (f
rom ale-py~=0.7.1->gym[accept-rom-license,atari,box2d]) (5.4.0)
Requirement already satisfied: click in /usr/local/lib/python3.7/dist-packages (from autorom[ac
cept-rom-license \[ \] ~= 0.4.2-\[ gym[accept-rom-license, atari, box2d] \] (7.1.2)
Requirement already satisfied: tqdm in /usr/local/lib/python3.7/dist-packages (from autorom[acc
ept-rom-license]~=0.4.2->gym[accept-rom-license,atari,box2d]) (4.62.3)
Requirement already satisfied: requests in /usr/local/lib/python3.7/dist-packages (from autorom
[accept-rom-license]~=0.4.2->gym[accept-rom-license,atari,box2d]) (2.23.0)
Requirement already satisfied: AutoROM.accept-rom-license in /usr/local/lib/python3.7/dist-pack
ages (from autorom[accept-rom-license]~=0.4.2->gym[accept-rom-license,atari,box2d]) (0.4.2)
Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-packages (from import
lib-metadata>=4.8.1->gym[accept-rom-license,atari,box2d]) (3.6.0)
Requirement already satisfied: typing-extensions>=3.6.4 in /usr/local/lib/python3.7/dist-packag
es (from importlib-metadata>=4.8.1->gym[accept-rom-license,atari,box2d]) (3.10.0.2)
Requirement already satisfied: future in /usr/local/lib/python3.7/dist-packages (from pyglet>=
1.4.0->gym[accept-rom-license,atari,box2d]) (0.16.0)
Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from req
uests->autorom[accept-rom-license]~=0.4.2->gym[accept-rom-license,atari,box2d]) (2.10)
Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /usr/local/lib/python
3.7/dist-packages (from requests->autorom[accept-rom-license]~=0.4.2->gym[accept-rom-license,at
ari,box2d]) (1.24.3)
Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packages (fro
m requests->autorom[accept-rom-license]~=0.4.2->gym[accept-rom-license,atari,box2d]) (3.0.4)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packages (fr
om requests->autorom[accept-rom-license]~=0.4.2->gym[accept-rom-license,atari,box2d]) (2021.10.
No GPU was detected. CNNs can be very slow without a GPU.
```

Go to Runtime > Change runtime and select a GPU hardware accelerator.

```
In [44]:
          def update_scene(num, frames, patch):
              patch.set_data(frames[num])
              return patch,
          def plot animation(frames, repeat=False, interval=40):
              fig = plt.figure()
              patch = plt.imshow(frames[0])
              plt.axis('off')
              anim = animation.FuncAnimation(
                  fig, update_scene, fargs=(frames, patch),
                  frames=len(frames), repeat=repeat, interval=interval)
              plt.close()
              return anim
```

## Deep Q-Network

Let's build the DQN. Given a state, it will estimate, for each possible action, the sum of discounted future rewards it can expect after it plays that action (but before it sees its outcome):

```
keras.backend.clear_session()
tf.random.set_seed(42)
np.random.seed(42)

env = gym.make("CartPole-v0")
input_shape = env.observation_space.shape
print(env.observation_space.shape)
n_outputs = 2 # == env.action_space.n

model = keras.models.Sequential([
    keras.layers.Dense(32, activation="elu", input_shape=input_shape),
    keras.layers.Dense(32, activation="elu"),
    keras.layers.Dense(n_outputs)
])
```

(4,)

To select an action using this DQN, we just pick the action with the largest predicted Q-value. However, to ensure that the agent explores the environment, we choose a random action with probability epsilon.

```
def epsilon_greedy_policy(state, epsilon=0):
    if np.random.rand() < epsilon:
        return np.random.randint(n_outputs)
    else:
        Q_values = model.predict(state[np.newaxis])
        return np.argmax(Q_values[0])</pre>
```

We will also need a replay memory. It will contain the agent's experiences, in the form of tuples: (obs, action, reward, next\_obs, done). We can use the deque class for that (but make sure to check out DeepMind's excellent Reverb library for a much more robust implementation of experience replay):

```
In [47]:
    from collections import deque
    replay_memory = deque(maxlen=2000)
```

And let's create a function to sample experiences from the replay memory. It will return 5 NumPy arrays: [obs, actions, rewards, next\_obs, dones].

```
def sample_experiences(batch_size):
    indices = np.random.randint(len(replay_memory), size=batch_size)
    batch = [replay_memory[index] for index in indices]
    states, actions, rewards, next_states, dones = [
          np.array([experience[field_index] for experience in batch])
          for field_index in range(5)]
    return states, actions, rewards, next_states, dones
```

Now we can create a function that will use the DQN to play one step, and record its experience in the replay memory:

```
def play_one_step(env, state, epsilon):
    action = epsilon_greedy_policy(state, epsilon)
    next_state, reward, done, info = env.step(action)
    replay_memory.append((state, action, reward, next_state, done))
    return next_state, reward, done, info
```

Lastly, let's create a function that will sample some experiences from the replay memory and perform a training step:

Notes:

- The first 3 releases of the 2nd edition were missing the reshape() operation which converts target\_Q\_values to a column vector (this is required by the loss\_fn()).
- The book uses a learning rate of 1e-3, but in the code below I use 1e-2, as it significantly improves training. I also tuned the learning rates of the DQN variants below.

```
In [50]:
          batch_size = 32
          discount_rate = 0.95
          optimizer = keras.optimizers.Adam(learning_rate=1e-2)
          loss_fn = keras.losses.mean_squared_error
          def training_step(batch_size):
              experiences = sample_experiences(batch_size)
              states, actions, rewards, next_states, dones = experiences
              next_Q_values = model.predict(next_states)
              max_next_Q_values = np.max(next_Q_values, axis=1)
              target_Q_values = (rewards +
                                  (1 - dones) * discount_rate * max_next_Q_values)
              target_Q_values = target_Q_values.reshape(-1, 1)
              mask = tf.one_hot(actions, n_outputs)
              with tf.GradientTape() as tape:
                  all_Q_values = model(states)
                  Q_values = tf.reduce_sum(all_Q_values * mask, axis=1, keepdims=True)
                  loss = tf.reduce_mean(loss_fn(target_Q_values, Q_values))
              grads = tape.gradient(loss, model.trainable_variables)
              optimizer.apply_gradients(zip(grads, model.trainable_variables))
```

```
And now, let's train the model!
In [51]:
          env.seed(42)
          np.random.seed(42)
          tf.random.set seed(42)
          rewards = []
          best score = 0
In [52]:
          %%bash
          apt-get install swig
          # install required system dependencies
          apt-get install -y xvfb x11-utils
          # install required python dependencies (might need to install additional gym extras depending)
          pip install gym[box2d] pyvirtualdisplay PyOpenGL PyOpenGL-accelerate
          pip install piglet
         Reading package lists...
         Building dependency tree...
         Reading state information...
         swig is already the newest version (3.0.12-1).
         0 upgraded, 0 newly installed, 0 to remove and 58 not upgraded.
         Reading package lists...
         Building dependency tree...
         Reading state information...
         x11-utils is already the newest version (7.7+3build1).
         xvfb is already the newest version (2:1.19.6-1ubuntu4.9).
         0 upgraded, 0 newly installed, 0 to remove and 58 not upgraded.
         Requirement already satisfied: gym[box2d] in /usr/local/lib/python3.7/dist-packages (0.21.0)
         Requirement already satisfied: pyvirtualdisplay in /usr/local/lib/python3.7/dist-packages (2.2)
         Requirement already satisfied: PyOpenGL in /usr/local/lib/python3.7/dist-packages (3.1.5)
         Requirement already satisfied: PyOpenGL-accelerate in /usr/local/lib/python3.7/dist-packages
         (3.1.5)
         Requirement already satisfied: EasyProcess in /usr/local/lib/python3.7/dist-packages (from pyvi
         rtualdisplay) (0.3)
         Requirement already satisfied: cloudpickle>=1.2.0 in /usr/local/lib/python3.7/dist-packages (fr
```

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Requirement already satisfied: numpy>=1.18.0 in /usr/local/lib/python3.7/dist-packages (from gy
         m[box2d]) (1.19.5)
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         ges (from gym[box2d]) (4.8.2)
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         gym[box2d]) (2.3.5)
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         m[box2d]) (1.5.0)
         Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-packages (from import
         lib-metadata>=4.8.1->gym[box2d]) (3.6.0)
         Requirement already satisfied: typing-extensions>=3.6.4 in /usr/local/lib/python3.7/dist-packag
         es (from importlib-metadata>=4.8.1->gym[box2d]) (3.10.0.2)
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         1.4.0 - \text{ym}[box2d]) (0.16.0)
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         piglet) (1.2.0)
         Requirement already satisfied: astunparse in /usr/local/lib/python3.7/dist-packages (from pigle
         t-templates->piglet) (1.6.3)
         Requirement already satisfied: pyparsing in /usr/local/lib/python3.7/dist-packages (from piglet
         -templates->piglet) (3.0.6)
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         plates->piglet) (21.2.0)
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         t-templates->piglet) (2.0.1)
         Requirement already satisfied: wheel<1.0,>=0.23.0 in /usr/local/lib/python3.7/dist-packages (fr
         om astunparse->piglet-templates->piglet) (0.37.0)
         Requirement already satisfied: six<2.0,>=1.6.1 in /usr/local/lib/python3.7/dist-packages (from
         astunparse->piglet-templates->piglet) (1.15.0)
In [53]:
          import pyvirtualdisplay
          from pyvirtualdisplay import Display
          display = Display(visible=0, size=(1400, 900))
          display.start()
          !apt install xvfb -y
          !pip install pyvirtualdisplay
          !pip install piglet
          from pyvirtualdisplay import Display
          display = Display(visible=0, size=(1400, 900))
          display.start()
          #_display = pyvirtualdisplay.Display(visible=False, # use False with Xvfb
                                              size=(1400, 900))
          # = display.start()
         Reading package lists... Done
         Building dependency tree
         Reading state information... Done
         xvfb is already the newest version (2:1.19.6-1ubuntu4.9).
         0 upgraded, 0 newly installed, 0 to remove and 58 not upgraded.
         Requirement already satisfied: pyvirtualdisplay in /usr/local/lib/python3.7/dist-packages (2.2)
         Requirement already satisfied: EasyProcess in /usr/local/lib/python3.7/dist-packages (from pyvi
         rtualdisplay) (0.3)
         Requirement already satisfied: piglet in /usr/local/lib/python3.7/dist-packages (1.0.0)
         Requirement already satisfied: piglet-templates in /usr/local/lib/python3.7/dist-packages (from
         piglet) (1.2.0)
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         -templates->piglet) (3.0.6)
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         t-templates->piglet) (2.0.1)
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         plates->piglet) (21.2.0)
         Requirement already satisfied: astunparse in /usr/local/lib/python3.7/dist-packages (from pigle
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om gym[box2d]) (1.3.0)

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Requirement already satisfied: six<2.0,>=1.6.1 in /usr/local/lib/python3.7/dist-packages (from
          astunparse->piglet-templates->piglet) (1.15.0)
          Requirement already satisfied: wheel<1.0,>=0.23.0 in /usr/local/lib/python3.7/dist-packages (fr
          om astunparse->piglet-templates->piglet) (0.37.0)
          <pyvirtualdisplay.display.Display at 0x7f1991963890>
Out[53]:
In [54]:
          for episode in range(600):
               obs = env.reset()
               for step in range(200):
                   epsilon = max(1 - episode / 500, 0.01)
                   obs, reward, done, info = play_one_step(env, obs, epsilon)
                   if done:
                       break
               rewards.append(step) # Not shown in the book
               if step >= best_score: # Not shown
                   best_weights = model.get_weights() # Not shown
                   best_score = step # Not shown
               print("\rEpisode: {}, Steps: {}, eps: {:.3f}".format(episode, step + 1, epsilon), end="") #
               if episode > 50:
                   training_step(batch_size)
          model.set_weights(best_weights)
          Episode: 599, Steps: 160, eps: 0.010
In [60]:
           plt.figure(figsize=(8, 4))
          plt.plot(rewards)
          print(max(rewards))
          plt.xlabel("Episode", fontsize=14)
          plt.ylabel("Sum of rewards", fontsize=14)
          save_fig("dqn_rewards_plot")
          plt.show()
          199
          Saving figure dqn_rewards_plot
             200
             175
            150
          Sum of rewards
             125
             100
              75
              50
              25
               0
                              100
                                         200
                                                    300
                                                               400
                                                                           500
                                                                                      600
                                                  Episode
In [59]:
           env.seed(42)
          state = env.reset()
          frames = []
```

t-templates->piglet) (1.6.3)

for step in range(1000):

frames.append(img)

if done:
 break

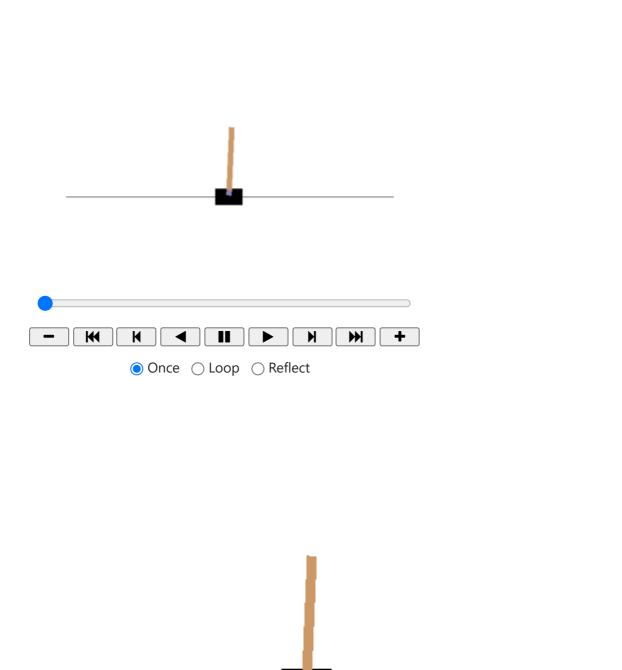
action = epsilon\_greedy\_policy(state)

img = env.render(mode="rgb\_array")

state, reward, done, info = env.step(action)

```
plot_animation(frames)
```

Out[59]:



## **Double DQN**

```
In [15]:
    keras.backend.clear_session()
    tf.random.set_seed(42)
    np.random.seed(42)

model = keras.models.Sequential([
         keras.layers.Dense(32, activation="elu", input_shape=[4]),
         keras.layers.Dense(32, activation="elu"),
         keras.layers.Dense(n_outputs)
])

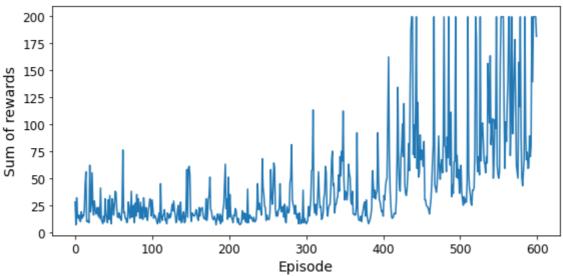
target = keras.models.clone_model(model)
target.set_weights(model.get_weights())
```

```
In [16]:
          batch_size = 32
          discount_rate = 0.95
          optimizer = keras.optimizers.Adam(learning_rate=6e-3)
          loss_fn = keras.losses.Huber()
          def training_step(batch_size):
              experiences = sample_experiences(batch_size)
              states, actions, rewards, next_states, dones = experiences
              next_Q_values = model.predict(next_states)
              best_next_actions = np.argmax(next_Q_values, axis=1)
              next_mask = tf.one_hot(best_next_actions, n_outputs).numpy()
              next_best_Q_values = (target.predict(next_states) * next_mask).sum(axis=1)
              target_Q_values = (rewards +
                                  (1 - dones) * discount_rate * next_best_Q_values)
              target_Q_values = target_Q_values.reshape(-1, 1)
              mask = tf.one_hot(actions, n_outputs)
              with tf.GradientTape() as tape:
                  all_Q_values = model(states)
                  Q_values = tf.reduce_sum(all_Q_values * mask, axis=1, keepdims=True)
                  loss = tf.reduce_mean(loss_fn(target_Q_values, Q_values))
              grads = tape.gradient(loss, model.trainable_variables)
              optimizer.apply_gradients(zip(grads, model.trainable_variables))
In [17]:
          replay memory = deque(maxlen=2000)
In [18]:
          env.seed(42)
          np.random.seed(42)
          tf.random.set_seed(42)
          rewards = []
          best_score = 0
          for episode in range(600):
              obs = env.reset()
              for step in range(200):
                  epsilon = max(1 - episode / 500, 0.01)
                  obs, reward, done, info = play_one_step(env, obs, epsilon)
                  if done:
                      break
              rewards.append(step)
              if step >= best score:
                  best_weights = model.get_weights()
                  best_score = step
              print("\rEpisode: {}, Steps: {}, eps: {:.3f}".format(episode, step + 1, epsilon), end="")
              if episode >= 50:
                  training step(batch size)
                  if episode % 50 == 0:
                      target.set_weights(model.get_weights())
              # Alternatively, you can do soft updates at each step:
              #if episode >= 50:
                  #target_weights = target.get_weights()
                  #online_weights = model.get_weights()
                  #for index in range(len(target_weights)):
                       target_weights[index] = 0.99 * target_weights[index] + 0.01 * online_weights[index
                  #target.set weights(target weights)
          model.set_weights(best_weights)
         Episode: 599, Steps: 182, eps: 0.010
```

In [42]: plt.figure(figsize=(8, 4))
 plt.plot(rewards)
 print(max(rewards))
 plt.xlabel("Episode", fontsize=14)

```
plt.ylabel("Sum of rewards", fontsize=14)
save_fig("double_dqn_rewards_plot")
plt.show()
```

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Saving figure double\_dqn\_rewards\_plot



```
In [21]:
    env.seed(43)
    state = env.reset()

    frames = []

    for step in range(1000):
        action = epsilon_greedy_policy(state)
        state, reward, done, info = env.step(action)
        if done:
            break
        img = env.render(mode="rgb_array")
        frames.append(img)

    plot_animation(frames)
```

Out[21]:

