Results

July 31, 2024

```
import json
import pickle

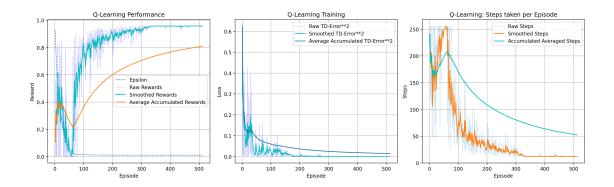
import matplotlib.animation as animation
import matplotlib.pyplot as plt
from IPython.display import HTML

from Utility.Plots import plot_live_data, plot_train_data, plot_steps_taken
from Utility.TabularEpsilonStrategy import EpsilonGreedy

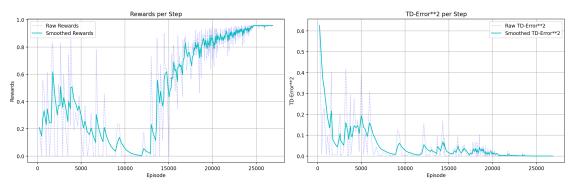
%matplotlib inline
%config InlineBackend.figure_format = 'retina'
```

0.1 Q-Learning

```
[2]: with open('Q_Learning/QLearning_live_plot.json', 'r') as f:
       data = json.load(f)
    plt.figure(figsize=(16, 5))
    plt.subplot(131)
    plot_live_data(data['episode'], data['rewards'], epsilon=data['epsilons'],
                 title={'title': "Q-Learning Performance", 'raw': 'Raw Rewards', |
     'acc' : 'Average Accumulated Rewards', 'xlab':
     plt.subplot(132)
    plot_train_data(data['episode'], data['losses'],
                  {'title': "Q-Learning Training", 'raw': 'Raw TD-Error**2', __
     'acc' : 'Average Accumulated TD-Error**2', 'xlab': 'Episode',
     plt.subplot(133)
    plot_steps_taken(data['episode'], data['steps_taken'], "Q-Learning")
    plt.tight layout()
    plt.savefig('report/figures/QLearning_episode.pdf')
    plt.savefig('report/figures/QLearning_episode.png')
    plt.show()
```



```
[3]: with open('Q_Learning/QLearning_train.json', 'r') as f:
        data = json.load(f)
    plt.figure(figsize=(16, 5))
    plt.subplot(121)
    plot_live_data(data['steps_done'], data['rewards'], avg=False,
                   title={'title': "Rewards per Step", 'raw': 'Raw Rewards', |
      ⇔'smooth': 'Smoothed Rewards',
                          'acc' : 'Cumulative Average Rewards', 'xlab': 'Episode',
      plt.subplot(122)
    plot_train_data(data['steps_done'], data['td_error_sq'], avg=False,
                    title={'title' : "TD-Error**2 per Step", 'raw': 'Rawu
      →TD-Error**2'.
                           'smooth': 'Smoothed TD-Error**2', 'acc': 'Cumulative⊔
      →Average TD-Error**2',
                           'xlab' : 'Episode', 'ylab': 'TD-Error**2'})
    plt.tight_layout()
    plt.savefig('report/figures/QLearning_steps.pdf')
    plt.savefig('report/figures/QLearning_steps.png')
    plt.show()
```



```
[4]: rewards = data['rewards']
    steps_taken = data['steps_taken']
    finish_counter = data['finish_counter']
    episodes = len(rewards)
    print('===== TRAIN SUMMARY ======')
    print(f'Episodes : {episodes}')
    print(f"Completion rate: {finish_counter / episodes}")
    print(f"Average Reward : {sum(rewards) / episodes:.3f}")
    print(f"Average steps : {sum(steps_taken) / episodes:.3f}")
    ===== TRAIN SUMMARY =====
    Episodes
                   : 512
    Completion rate: 0.9296875
    Average Reward: 0.808
    Average steps : 52.510
[5]: from Q_Learning.QLearning import eval as qlearning_eval
    from Utility.MiniGrid import MiniGridHash
    env = MiniGridHash(render_mode='rgb_array')
    with open('Q_Learning/q_learning_table.pkl', 'rb') as f:
        glearning table = pickle.load(f)
    qlearning_eval(env, qlearning_table, EpsilonGreedy(0, 3), 1000)
    pygame 2.6.0 (SDL 2.28.4, Python 3.10.14)
    Hello from the pygame community. https://www.pygame.org/contribute.html
    Evaluation...
    ===== EVALUATION SUMMARY =====
    Evaluation episodes: 1000
    Completion rate : 1.0
    Average Reward
                      : 0.958
    Average steps
                      : 12.000
[6]: imgs = []
    fig = plt.figure(figsize=(5, 5))
    strategy = EpsilonGreedy(0, 3)
    current_state = env.reset()
    plt.axis('off')
    im = plt.imshow(env.render(), animated=True)
    imgs.append([im])
    for step in range(env.maxSteps):
        action = strategy.select_action(current_state, qlearning_table)
        next_state, reward, done, truncated = env.step(action)
         im = plt.imshow(env.render(), animated=True)
```

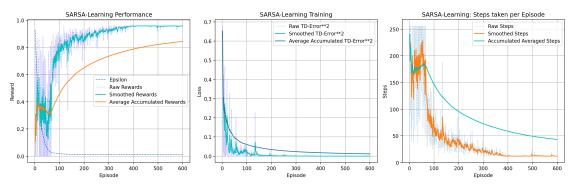
```
imgs.append([im])
if done or truncated:
    break
    current_state = next_state
ani = animation.ArtistAnimation(fig, imgs, interval=100, repeat_delay=1000)
ani.save('Q_Learning/q_learning.gif', writer='imagemagickasd')
plt.close()
HTML(ani.to_jshtml())
```

MovieWriter imagemagickasd unavailable; using Pillow instead.

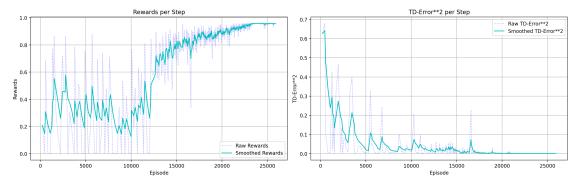
[6]: <IPython.core.display.HTML object>

0.2 SARSA

```
[7]: with open('SARSA/SARSA_live_plot.json', 'r') as f:
        data = json.load(f)
    plt.figure(figsize=(16, 5))
    plt.subplot(131)
    plot_live_data(data['episode'], data['rewards'], epsilon=data['epsilons'],
                  title={'title': "SARSA-Learning Performance", 'raw': 'Rawu
     →Rewards', 'smooth': 'Smoothed Rewards',
                         'acc' : 'Average Accumulated Rewards', 'xlab':
     ⇔'Episode', 'ylab': 'Reward'})
    plt.subplot(132)
    plot_train_data(data['episode'], data['losses'],
                   {'title': "SARSA-Learning Training", 'raw': 'Raw TD-Error**2',
      'acc' : 'Average Accumulated TD-Error**2', 'xlab': 'Episode',
     plt.subplot(133)
    plot_steps_taken(data['episode'], data['steps_taken'], "SARSA-Learning")
    plt.tight_layout()
    plt.savefig('report/figures/SARSALearning_episode.pdf')
    plt.savefig('report/figures/SARSALearning_episode.png')
    plt.show()
```



```
[8]: with open('SARSA/SARSA_train.json', 'r') as f:
        data = json.load(f)
    plt.figure(figsize=(16, 5))
    plt.subplot(121)
    plot_live_data(data['steps_done'], data['rewards'], avg=False,
                   title={'title': "Rewards per Step", 'raw': 'Raw Rewards', |
      ⇔'smooth': 'Smoothed Rewards',
                          'acc' : 'Cumulative Average Rewards', 'xlab': 'Episode',
      plt.subplot(122)
    plot_train_data(data['steps_done'], data['td_error_sq'], avg=False,
                    title={'title' : "TD-Error**2 per Step", 'raw': 'Rawu
      →TD-Error**2',
                           'smooth': 'Smoothed TD-Error**2', 'acc': 'Cumulative
      →Average TD-Error**2',
                           'xlab' : 'Episode', 'ylab': 'TD-Error**2'})
    plt.tight_layout()
    plt.savefig('report/figures/SARSALearning_steps.pdf')
    plt.savefig('report/figures/SARSALearning_steps.png')
    plt.show()
```



```
[9]: rewards = data['rewards']
    steps_taken = data['steps_taken']
    finish_counter = data['finish_counter']
    episodes = len(rewards)
    print('====== TRAIN SUMMARY ======')
    print(f'Episodes : {episodes}')
    print(f"Completion rate: {finish_counter / episodes}")
```

```
print(f"Average Reward : {sum(rewards) / episodes:.3f}")
print(f"Average steps : {sum(steps_taken) / episodes:.3f}")
```

===== TRAIN SUMMARY =====

Episodes : 600

Completion rate: 0.95666666666666667

Average Reward : 0.844 Average steps : 43.185

0.2.1 Evaluation

```
[10]: from SARSA.SARSA import eval as sarsa_eval
    from Utility.MiniGrid import MiniGridHash

env = MiniGridHash(render_mode='rgb_array')
    with open('SARSA/sarsa_learning_table.pkl', 'rb') as f:
        sarsa_table = pickle.load(f)
    sarsa_eval(env, qlearning_table, EpsilonGreedy(0, 3), 1000)
```

```
Evaluation...
```

===== EVALUATION SUMMARY ======

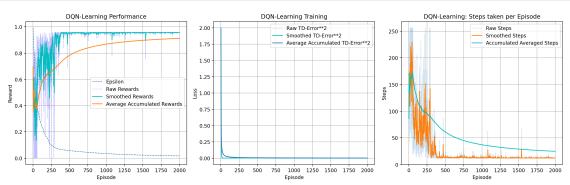
Evaluation episodes: 1000 Completion rate : 1.0 Average Reward : 0.958 Average steps : 12.000

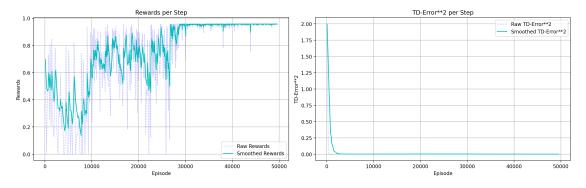
```
[11]: imgs = []
      fig = plt.figure(figsize=(5, 5))
      strategy = EpsilonGreedy(0, 3)
      current_state = env.reset()
      plt.axis('off')
      im = plt.imshow(env.render(), animated=True)
      imgs.append([im])
      for step in range(env.maxSteps):
          action = strategy.select_action(current_state, sarsa_table)
          next state, reward, done, truncated = env.step(action)
          im = plt.imshow(env.render(), animated=True)
          imgs.append([im])
          if done or truncated:
              break
          current_state = next_state
      ani = animation.ArtistAnimation(fig, imgs, interval=100, repeat_delay=1000)
      ani.save('SARSA/sarsa.gif', writer='imagemagick')
      plt.close()
      HTML(ani.to_jshtml())
```

MovieWriter imagemagick unavailable; using Pillow instead.

0.3 DQN

```
[12]: with open('DQN/DQN_live_plot.json', 'r') as f:
         data = json.load(f)
     plt.figure(figsize=(16, 5))
     plt.subplot(131)
     plot_live_data(data['episode'], data['rewards'], epsilon=data['epsilons'],
                    title={'title': "DQN-Learning Performance", 'raw': 'Rawu
      →Rewards', 'smooth': 'Smoothed Rewards',
                          'acc' : 'Average Accumulated Rewards', 'xlab':
      plt.subplot(132)
     plot_train_data(data['episode'], data['losses'],
                     {'title': "DQN-Learning Training", 'raw': 'Raw TD-Error**2',
      ⇔'smooth': 'Smoothed TD-Error**2',
                     'acc' : 'Average Accumulated TD-Error**2', 'xlab': 'Episode',
      plt.subplot(133)
     plot_steps_taken(data['episode'], data['steps_taken'], "DQN-Learning")
     plt.tight_layout()
     plt.savefig('report/figures/DQNLearning_episode.pdf')
     plt.savefig('report/figures/DQNLearning_episode.png')
     plt.show()
```





```
rewards = data['rewards']
steps_taken = data['steps_taken']
finish_counter = data['finish_counter']
episodes = len(rewards)
print('===== TRAIN SUMMARY ======')
print(f'Episodes : {episodes}')
print(f"Completion rate: {finish_counter / episodes}")
print(f"Average Reward : {sum(rewards) / episodes:.3f}")
print(f"Average steps : {sum(steps_taken) / episodes:.3f}")
```

===== TRAIN SUMMARY =====

Episodes : 2000 Completion rate: 0.992 Average Reward : 0.912 Average steps : 24.794

0.4 Evaluation

```
[15]: from DQN.DQN import eval as dqn_eval
     from DQN.DQN import DQN
     from Utility.DQNEpsilonStrategy import EpsilonGreedy as DQNEpsilonGreedy
     from Utility.MiniGrid import MiniGridRaw
     import torch
     from torchinfo import summary
     env_dqn = MiniGridRaw(render_mode='rgb_array')
     policy_net = DQN(env_dqn.numStates, env_dqn.numActions, (64, 32))
[16]: a = env_dqn.reset()
     print(summary(policy_net, input_data=a, col_names=['input_size', 'output_size', _

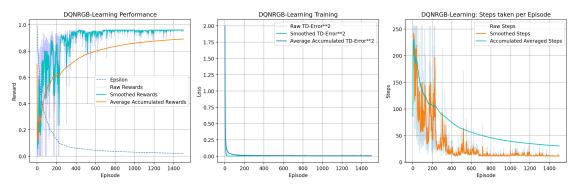
¬'num_params', 'trainable']))
    ______
    _____
    Layer (type:depth-idx)
                                                               Output Shape
                                        Input Shape
    Param #
                           Trainable
    DQN
                                        [1, 49]
                                                              [1, 3]
                           True
                                                              [1, 3]
                                       [1, 49]
     Sequential: 1-1
                           True
                                       [1, 49]
         Linear: 2-1
                                                              [1, 64]
    3,200
                           True
                                       [1, 64]
                                                              [1, 64]
         ReLU: 2-2
         Linear: 2-3
                                       [1, 64]
                                                              [1, 32]
    2,080
                           True
                                       [1, 32]
                                                              [1, 32]
         ReLU: 2-4
                                       [1, 32]
         Linear: 2-5
                                                              [1, 3]
    99
                           True
    Total params: 5,379
    Trainable params: 5,379
    Non-trainable params: 0
    Total mult-adds (M): 0.01
    ______
    Input size (MB): 0.00
    Forward/backward pass size (MB): 0.00
    Params size (MB): 0.02
    Estimated Total Size (MB): 0.02
```

```
[17]: policy_net.load_state_dict(torch.load('DQN/dqn.pth'))
      dqn_eval(env_dqn, policy_net, DQNEpsilonGreedy(0, 3), 1000)
     Evaluation...
     ===== EVALUATION SUMMARY =====
     Evaluation episodes: 1000
     Completion rate : 1.0
     Average Reward
                        : 0.958
     Average steps
                        : 12.000
[18]: | imgs = []
      fig = plt.figure(figsize=(5, 5))
      strategy_dqn = DQNEpsilonGreedy(0, 3)
      current_state = env_dqn.reset()
      plt.axis('off')
      imgs.append([plt.imshow(env_dqn.render(), animated=True)])
      for step in range(env.maxSteps):
          action = strategy_dqn.select_action(current_state, policy_net)
          next_state, reward, done, truncated = env_dqn.step(action)
          imgs.append([plt.imshow(env_dqn.render(), animated=True)])
          if done or truncated:
              break
          current_state = next_state
      ani = animation.ArtistAnimation(fig, imgs, interval=100, repeat_delay=1000)
      ani.save('DQN/dqn.gif', writer='imagemagick')
      plt.close()
      HTML(ani.to_jshtml())
```

MovieWriter imagemagick unavailable; using Pillow instead.

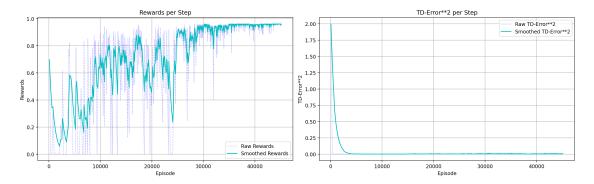
[18]: <IPython.core.display.HTML object>

0.5 DQN RGB Technique



```
[20]: with open('DQN/DQNIMAGE_train.json', 'r') as f:
         data = json.load(f)
     plt.figure(figsize=(16, 5))
     plt.subplot(121)
     plot_live_data(data['steps_done'], data['rewards'], avg=False,
                    title={'title': "Rewards per Step", 'raw': 'Raw Rewards', |
       ⇔'smooth': 'Smoothed Rewards',
                           'acc' : 'Cumulative Average Rewards', 'xlab': 'Episode',
       plt.subplot(122)
     plot_train_data(data['steps_done'], data['td_error_sq'], avg=False,
                     title={'title' : "TD-Error**2 per Step", 'raw': 'Rawu
       ⇔TD-Error**2',
                            'smooth': 'Smoothed TD-Error**2', 'acc': 'Cumulative⊔
       →Average TD-Error**2',
                            'xlab' : 'Episode', 'ylab': 'TD-Error**2'})
     plt.tight_layout()
     plt.savefig('report/figures/DQNRGBLearning_steps.pdf')
     plt.savefig('report/figures/DQNRGBLearning_steps.png')
```

plt.show()



```
[21]: rewards = data['rewards']
    steps_taken = data['steps_taken']
    finish_counter = data['finish_counter']
    episodes = len(rewards)
    print('===== TRAIN SUMMARY ======')
    print(f'Episodes : {episodes}')
    print(f"Completion rate: {finish_counter / episodes}")
    print(f"Average Reward : {sum(rewards) / episodes:.3f}")
    print(f"Average steps : {sum(steps_taken) / episodes:.3f}")
```

===== TRAIN SUMMARY =====

Episodes : 1500 Completion rate: 0.98 Average Reward : 0.892 Average steps : 30.068

0.6 Evaluation

```
[22]: from DQN.DQNIMAGE import eval as dqn_image_eval
    from DQN.DQNIMAGE import CNN_DQN, FrameStack
    from Utility.DQNEpsilonStrategy import EpsilonGreedy as DQNEpsilonGreedy
    from Utility.MiniGrid import MiniGridImage, get_device
    import torch

device = get_device()

env_dqn_img = MiniGridImage(render_mode='rgb_array')
    policy_net_img = CNN_DQN(56, 56, 3).to(device)
```

```
[23]: summary(policy_net_img, input_size=(1,4,56,56), device=device,__

col_names=['input_size', 'kernel_size', 'output_size', 'num_params',__

'trainable'])
```

Layer (type:depth-idx) Output Shape	Param #	Input Shape Trainable	Kernel Shape
===== CNN_ DON		[4 4 50 50]	
CNN_DQN [1, 3]		[1, 4, 56, 56] True	
•			
Sequential: 1-1		[1, 4, 56, 56]	
[1, 512]		True	רס סו
Conv2d: 2-1	F7 <i>C</i>	[1, 4, 56, 56]	[3, 3]
[1, 16, 27, 27]	576	True	
BatchNorm2d: 2-2	20	[1, 16, 27, 27]	
[1, 16, 27, 27]	32	True	
ReLU: 2-3		[1, 16, 27, 27]	
[1, 16, 27, 27]			[O O]
Conv2d: 2-4		[1, 16, 27, 27]	[3, 3]
[1, 32, 13, 13]	4,608	True	
BatchNorm2d: 2-5		[1, 32, 13, 13]	
[1, 32, 13, 13]	64	True	
ReLU: 2-6		[1, 32, 13, 13]	
[1, 32, 13, 13]			
Conv2d: 2-7		[1, 32, 13, 13]	[3, 3]
[1, 64, 6, 6]	18,432	True	
BatchNorm2d: 2-8		[1, 64, 6, 6]	
[1, 64, 6, 6]	128	True	
ReLU: 2-9		[1, 64, 6, 6]	
[1, 64, 6, 6]			
Conv2d: 2-10		[1, 64, 6, 6]	[3, 3]
[1, 128, 2, 2]	73,728	True	
BatchNorm2d: 2-11		[1, 128, 2, 2]	
[1, 128, 2, 2]	256	True	
Flatten: 2-12		[1, 128, 2, 2]	
[1, 512]			
Sequential: 1-2		[1, 512]	
[1, 3]		True	
Linear: 2-13		[1, 512]	
[1, 64]	32,832	True	
ReLU: 2-14		[1, 64]	
[1, 64]			
Linear: 2-15		[1, 64]	
[1, 3]	195	True	

=====

```
Total params: 130,851
     Trainable params: 130,851
     Non-trainable params: 0
     Total mult-adds (M): 2.19
     ______
     Input size (MB): 0.05
     Forward/backward pass size (MB): 0.32
     Params size (MB): 0.52
     Estimated Total Size (MB): 0.89
     ______
     ____
[24]: policy_net_img.load_state_dict(torch.load('DQN/dqn_image.pth',_
      →map_location=device))
     dqn_image_eval(env_dqn_img, policy_net_img, DQNEpsilonGreedy(0, 3), 1000)
    Evaluation...
    /Users/lucien/.pyenv/versions/3.10.14/envs/torch/lib/python3.10/site-
    packages/gymnasium/core.py:311: UserWarning: WARN: env.get_frame to get
    variables from other wrappers is deprecated and will be removed in v1.0, to get
    this variable you can do `env.unwrapped.get_frame` for environment variables or
     `env.get wrapper attr('get frame')` that will search the reminding wrappers.
      logger.warn(
    ===== EVALUATION SUMMARY ======
    Evaluation episodes: 1000
    Completion rate : 1.0
    Average Reward : 0.961
    Average steps
                    : 11.000
[25]: | imgs = []
     fig = plt.figure(figsize=(5, 5))
     frame_stack = FrameStack(56, 56, 4)
     strategy_dqn_img = DQNEpsilonGreedy(0, 3)
     current_state = env_dqn_img.reset()
     current_stack = frame_stack.push(current_state, True)
     plt.axis('off')
     imgs.append([plt.imshow(env_dqn_img.render(), animated=True)])
     for step in range(env_dqn_img.maxSteps):
         action = strategy_dqn_img.select_action(current_stack, policy_net_img)
         next_state, reward, done, truncated = env_dqn_img.step(action.item())
         next_stack = frame_stack.push(next_state, False)
         imgs.append([plt.imshow(env_dqn_img.render(), animated=True)])
```

```
if done or truncated:
          break
     current_stack = next_stack
ani = animation.ArtistAnimation(fig, imgs, interval=100, repeat_delay=1000)
ani.save('DQN/dqn_img.gif', writer='imagemagick')
plt.close()
HTML(ani.to_jshtml())
```

MovieWriter imagemagick unavailable; using Pillow instead.

[25]: <IPython.core.display.HTML object>