

## About the code

hw3.m - Made with Matlab R2019a

## About the algorithm

**Parameters** used:

- $\alpha = 0.95$
- $\gamma = 0.05$
- $\epsilon = 0.05$

The **action selection policy** used was epsilon-greedy, as described in this pseudocode:

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**Algorithm 2:** Epsilon-Greedy Action Selection

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**Data:**  $Q$ :  $Q$ -table generated so far,  $\epsilon$ : a small number,  $S$ : current state

**Result:** Selected action

**Function** *SELECT-ACTION*( $Q, S, \epsilon$ ) **is**

```
     $n \leftarrow$  uniform random number between 0 and 1;  
    if  $n < \epsilon$  then  
        |  $A \leftarrow$  random action from the action space;  
    else  
        |  $A \leftarrow \max Q(S, \cdot)$ ;  
    end  
    return selected action  $A$ ;  
end
```

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The **actions** a robot can take in any state is to go up, down, right, or left, encoded by the numbers 1-4.

The **states** a robot can be in corresponds with the cells of a grid, state 1 being the top, leftmost cell. Cells/states are numbered from left to right, top to bottom. For example, the numbering schema for a 5x5 grid is pictured below.

1	2	3	4	5
6	7	8	9	10
11	12	13	14	15
16	17	18	19	20
21	22	23	24	25

The **reward system** is slightly different than the system suggested in the assignment guidelines. Reward for actions were given as such:

- Action that makes the robot tend to go out of the grid will get a reward of -1 (when the robot is in the border cells)
- Action that makes the robot reach the goal will get a reward of 100
- Action that brings the robot to a cell previously explored in the current episode will get a reward of 0
- All other actions will get a reward of 0

The **start** cell of each grid is colored **green**, while the **goal** cell of each grid is colored **red**.

The **Q table** was organized so that columns are actions (1-4) and rows are states (1-25, assuming a 5x5 grid).

The **total rewards plot** has black vertical lines indicating the ends of learning episodes.

To clearly see the directions of the **arrows** in the grid, it is recommended to view the plots in full screen. Note that arrows may overlap.

## Run #1

- 5x5 grid
- **start** = 1 (top left cell)
- **goal** = 25 (bottom right cell)

- number of episodes = 6
- number of iterations in each episode = 180

Figure 2, the first learning episode:

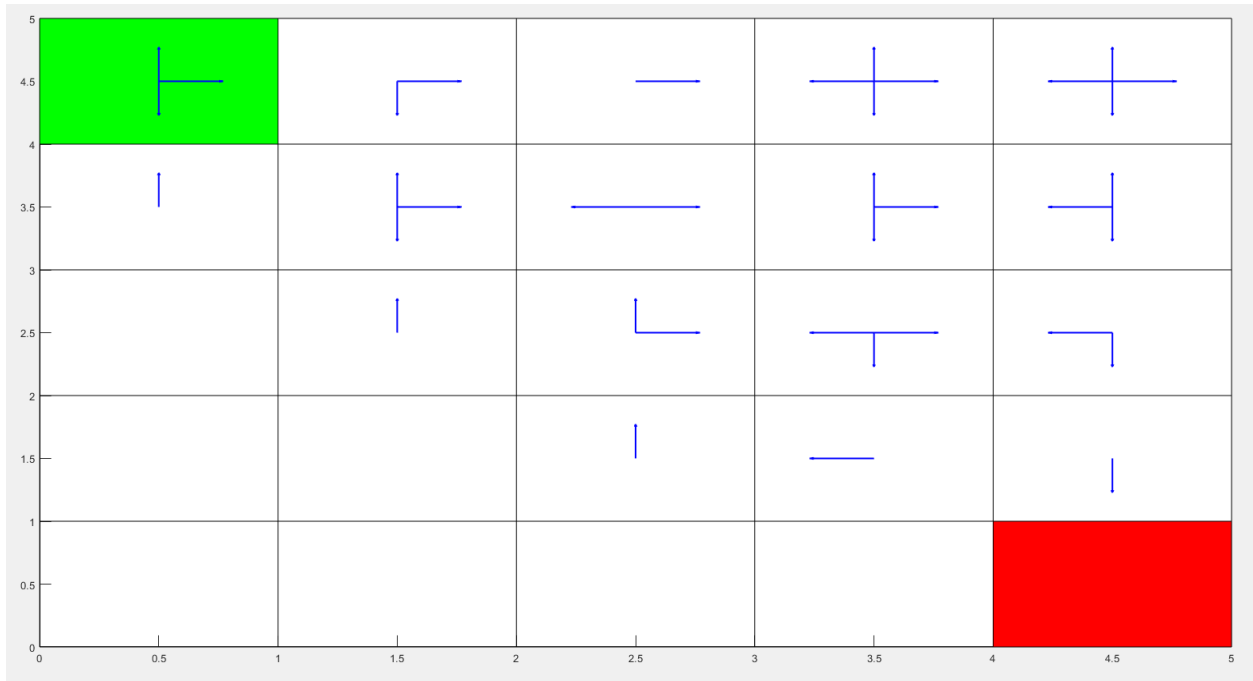
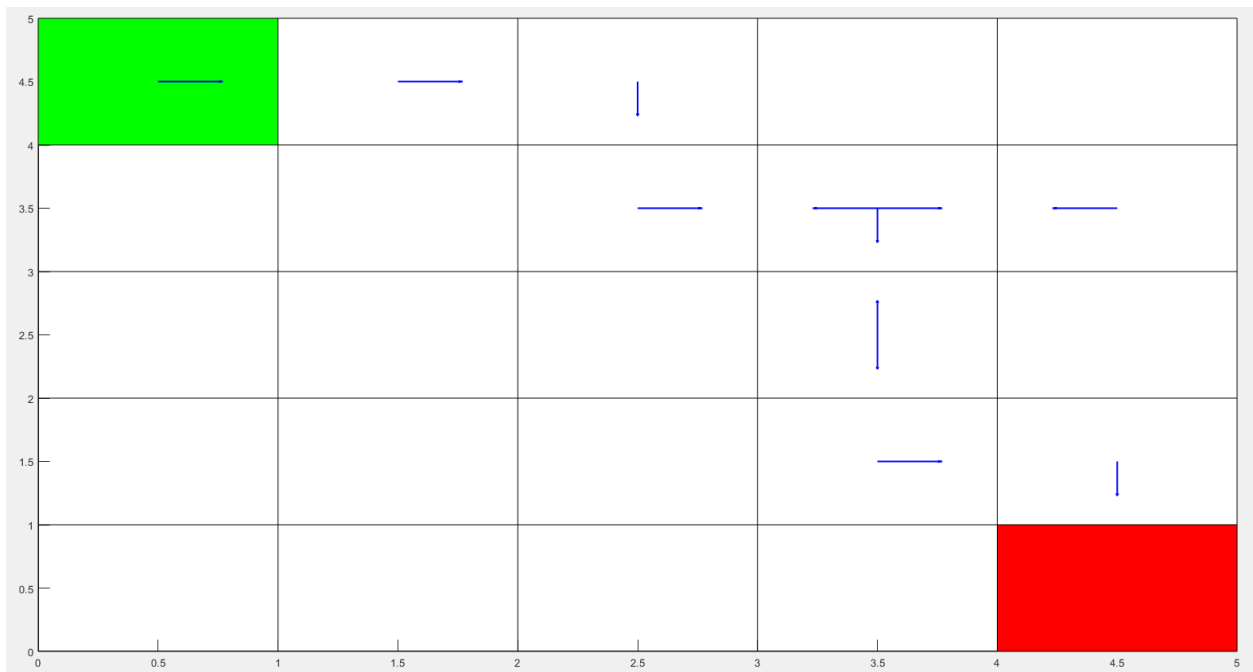


Figure 3, the final learning episode:

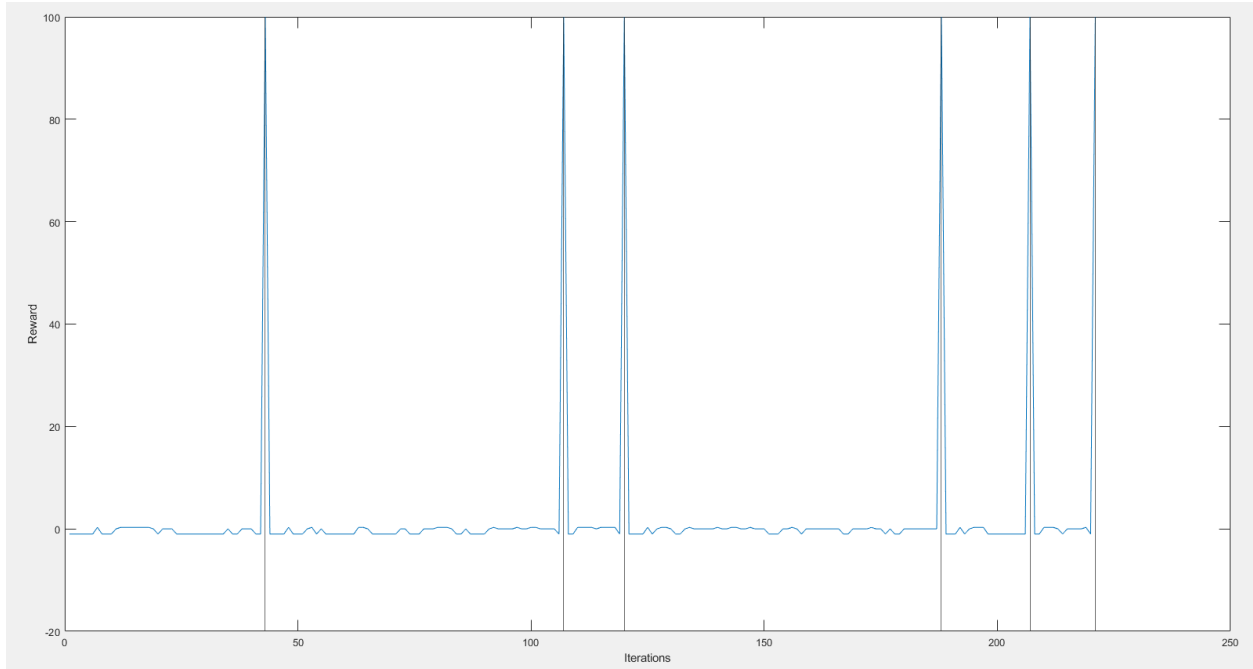


Q table for final episode:

```
Final_Q_Table =
```

-1.0322	-0.9966	-0.9830	-1.0407
-0.5600	0.3290	-0.7905	-0.5575
1.0000	0.4987	-0.9842	2.0000
-0.9876	0.0026	-1.0279	-0.7100
-0.8879	-0.8330	-1.0236	-0.9881
-0.8355	-0.8928	0.0152	-0.8525
-0.5625	0.0177	0.2879	-0.9850
-0.6100	0.0414	0.0302	0.0038
-0.9997	0.0208	-0.9571	0.0150
-0.8050	-0.7931	-0.6575	0.0364
-0.9950	-0.9356	0.1546	-0.7550
0.0328	0.0283	0.0399	-0.9869
0.0435	0.0412	0.3088	0.0671
0.0179	0.4795	-0.7955	0.1070
-0.6075	3.9878	-0.5600	0.2975
-0.9787	-0.6625	0.0524	-0.9388
0.0334	0.0021	0.2400	-0.9483
0.0498	0.5006	0.2950	0.0311
0.1196	1.0000	3.9986	0.5611
-0.6100	100.0000	3.9494	2.0000
-0.6525	-0.8781	0.0176	-0.7550
0.0331	0.0169	0.0281	-0.9926
0.0432	0.0975	0.0753	0.0522
0.6594	0.4875	95.0500	0.1776
0	0	0	0

Figure 1, reward for all episodes:



## Run #2

- 5x5 grid
- **start** = 7 (second row from top, second column)
- **goal** = 19 (fourth row from top, fourth column)
- number of episodes = 6
- number of iterations in each episode = 180

Figure 2, the first learning episode:

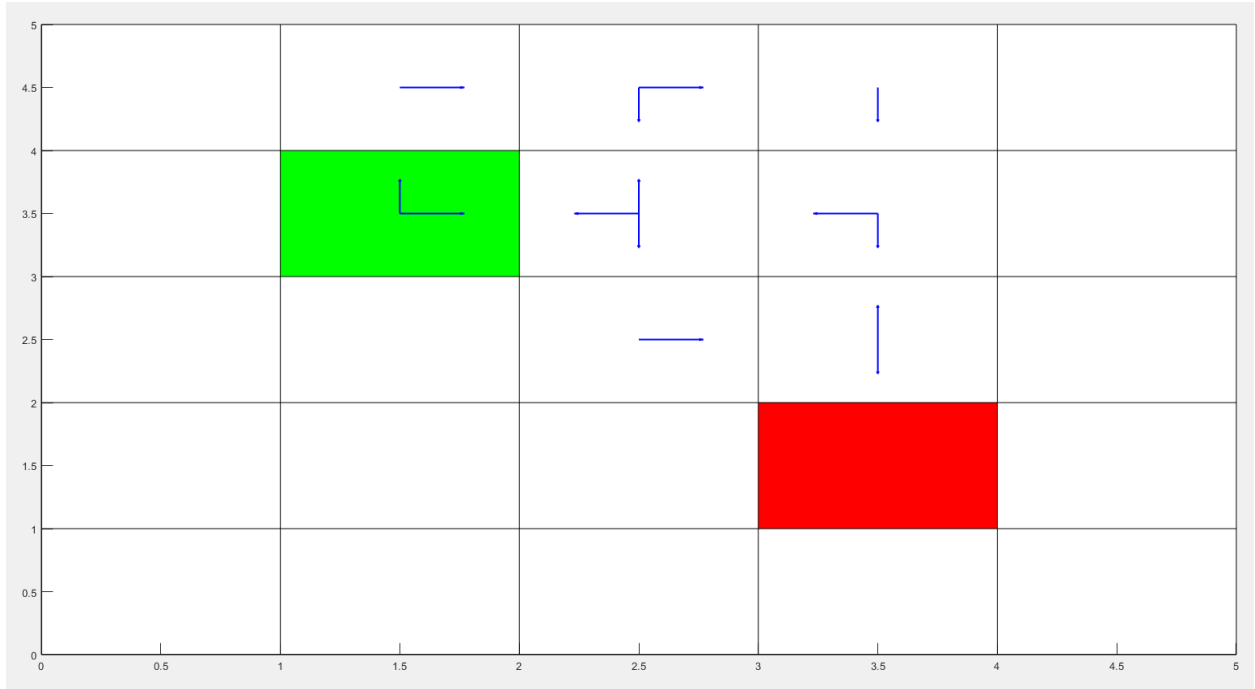
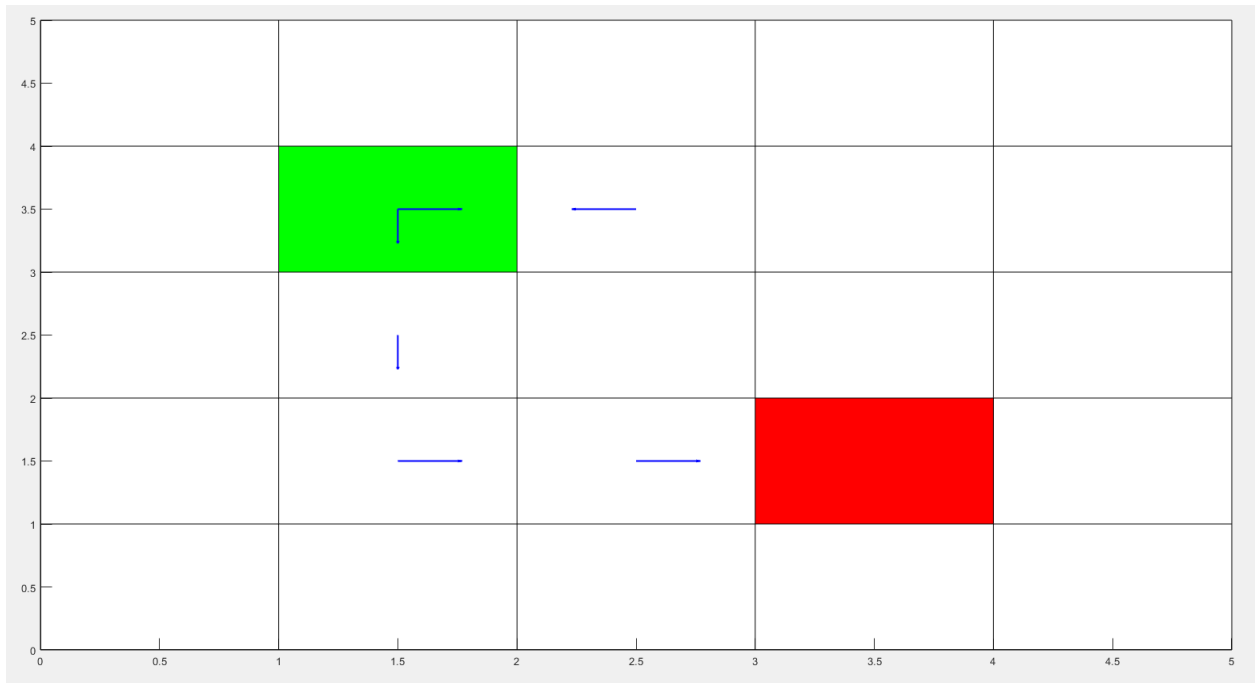


Figure 3, the final learning episode:

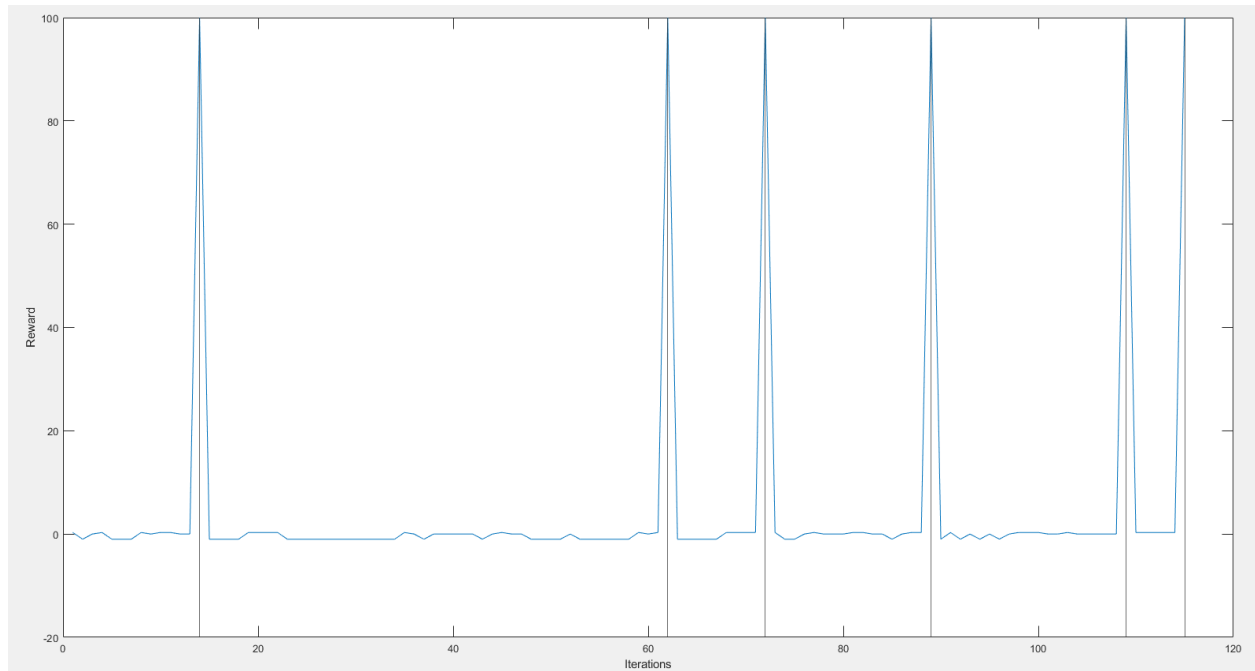


Q table for final episode:

Final\_Q\_Table =

-0.6125	-0.6575	-0.8025	-0.4625
-0.7550	0.0171	-0.9704	-0.4625
-0.5625	0.0563	-0.9364	-0.4625
-0.5600	0.2286	-0.5125	-0.5125
-0.7431	-0.8025	-0.4625	-0.5600
-0.6125	-0.5100	0.0468	-0.5625
-0.9977	0.3275	0.3155	-0.9642
-0.9547	0.5250	0.5750	0.3119
-0.5600	0.3425	-0.5600	0.3450
-0.8050	-0.7100	-0.9305	1.0000
3.0000	2.0000	0.0339	-0.8355
0.0184	0.5500	0.0523	-0.8354
0.4800	0.6225	0.7225	0.0277
0.4400	95.1500	2.0000	0.1450
3.0000	2.0000	1.0000	5.0046
1.0000	1.0000	2.0000	2.0000
0.1797	0.5725	5.2849	4.0000
2.0000	0.6355	100.0000	0.2503
0	0	0	0
4.0000	5.0000	5.0000	3.0000
4.0000	4.0000	0.4400	1.0000
2.0000	0.2144	0.7225	-0.5125
5.0131	5.0000	2.0000	3.0000
2.0000	2.0000	4.0000	5.0000
3.0000	1.0000	1.0000	2.0000

Figure 1, reward for all episodes:



## Run #3

- 5x5 grid
- **start** = 24 (bottom row, fourth column)
- **goal** = 2 (top row, second column)
- number of episodes = 6
- number of iterations in each episode = 180

Figure 2, the first learning episode:

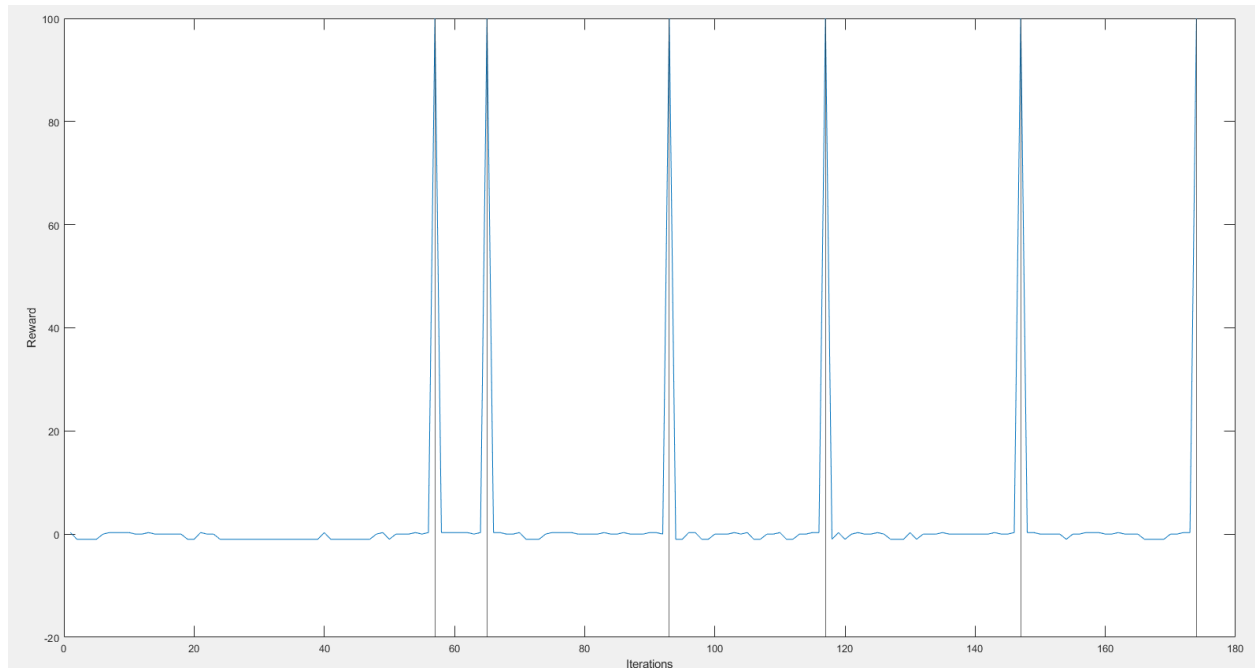




Final\_Q\_Table =

1.0000	1.0000	4.0000	2.0000
0	0	0	0
-0.4625	0.3425	-0.6704	3.0000
-0.9436	0.6225	-1.0021	-0.6625
-0.5600	-0.7955	-1.0192	-0.8781
5.0000	5.0000	2.0000	2.0000
100.0000	1.0000	0.3669	3.0000
-0.7100	0.6725	0.1950	5.2728
-0.7025	0.1297	-0.8050	0.5203
-0.9976	-0.8179	-0.5600	0.4375
2.0000	1.0000	5.0000	4.0000
5.2999	0.0768	0.1288	1.0000
0.3925	0.2119	0.1950	0.2621
0.0262	0.2950	-0.8381	0.0567
-0.5100	-0.7940	-0.7356	0.0507
1.0000	-0.9256	0.0457	1.0000
0.5506	0.0609	0.0401	-0.9306
0.1123	0.0038	0.0445	0.3116
0.0393	0.0278	-0.8902	0.3155
-0.6625	-0.7575	-0.7550	0.0303
-0.9493	-0.4625	0.0529	-0.6575
0.0163	0.0195	0.0150	-0.8331
0.0248	0.0416	0.0184	0.0243
0.3111	0.0022	-0.9977	0.0295
-0.7050	-0.7550	-0.6575	0.0037

Figure 1, reward for all episodes:



## Run #4

- 10x10 grid
- **start** = 1 (top row, first column)
- **goal** = 100 (bottom row, last column)
- number of episodes = 6
- number of iterations in each episode = 180

Figure 2, the first learning episode:

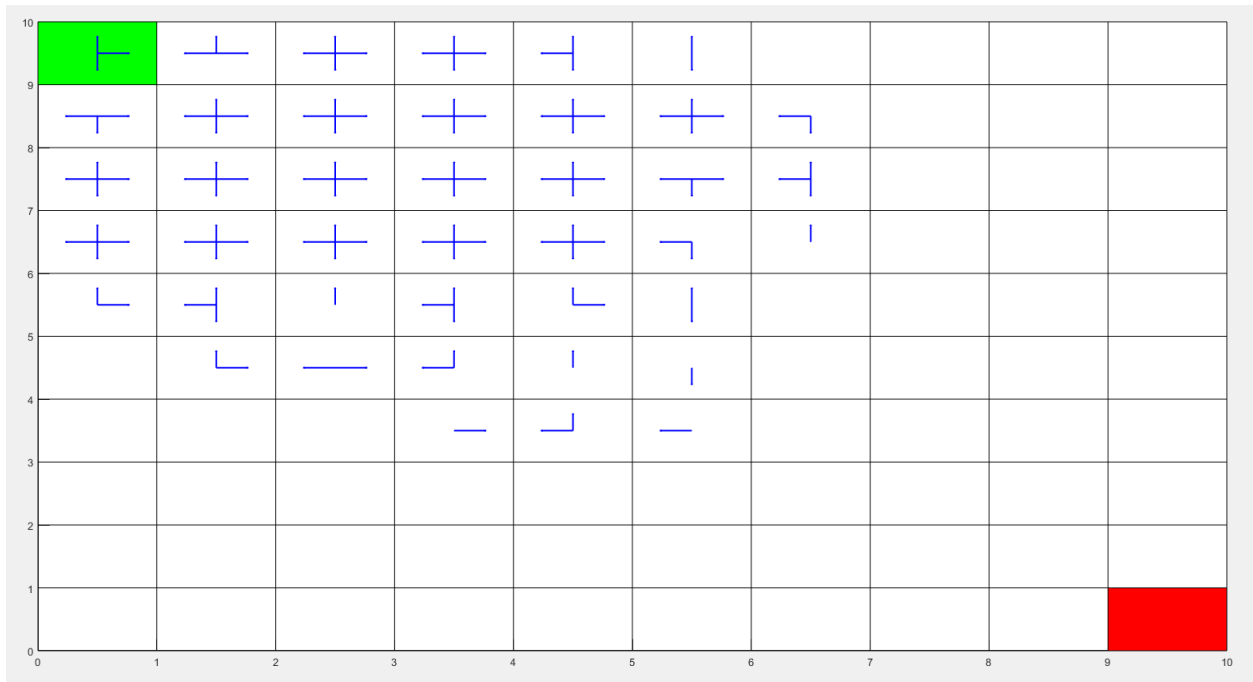
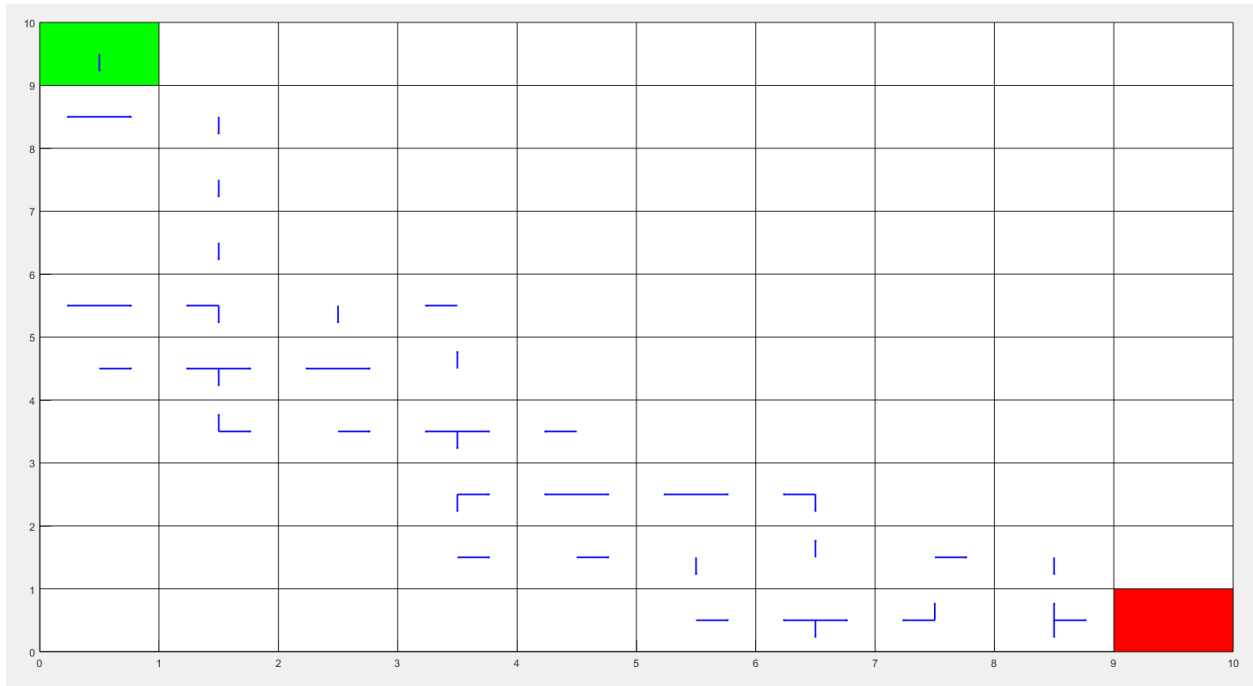


Figure 3, the final learning episode:



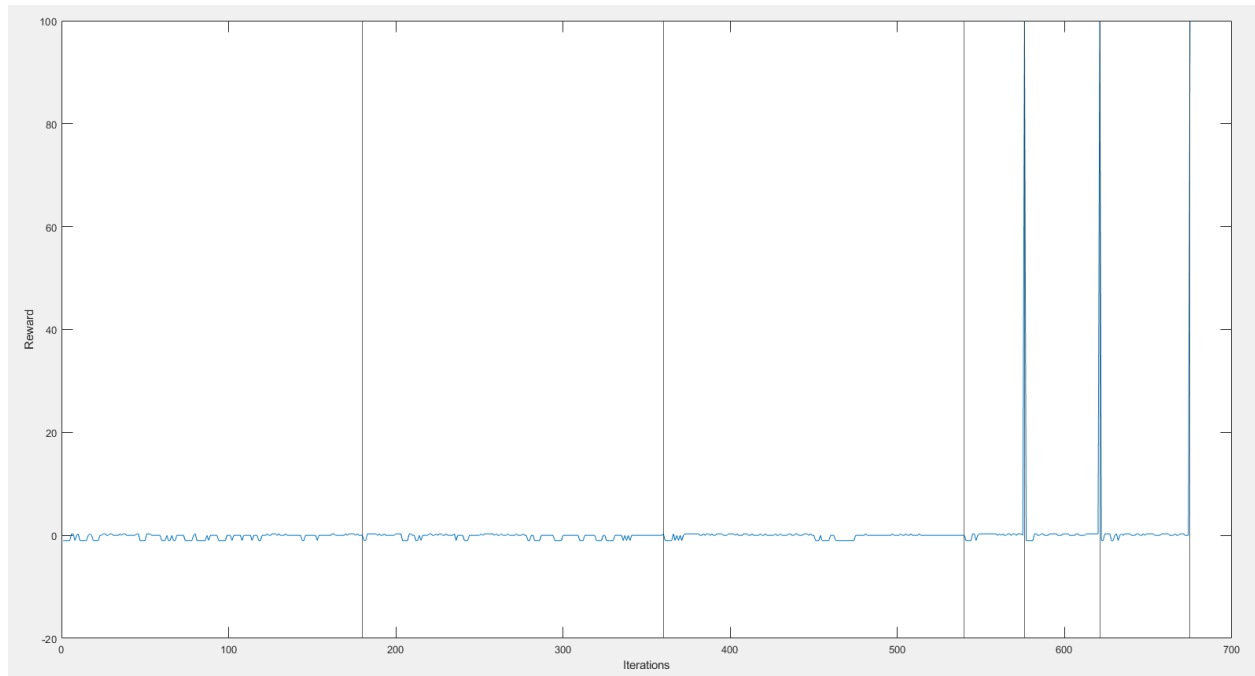
Q table for final episode:

	1	2	3	4
1	-1.0220	-0.9850	-0.9923	-1.0225
2	-0.5125	0.0337	-0.8781	-1.0129
3	-0.9862	0.0050	-0.7315	-0.6125
4	-0.6100	0.0190	-0.6550	-0.6100
5	-0.7075	0.3375	-0.8919	-0.8380
6	-0.4625	0.0174	-0.6100	-0.7525
7	-0.7955	0.2139	1	-0.9767
8	2	1	-0.5600	-0.7525
9	-0.8525	0.6725	-0.6075	-0.7575
10	3	2	1	-0.6575
11	-0.8220	-0.9986	0.3149	-0.9837
12	-0.9617	0.3149	0.2901	-0.9973
13	-0.9984	0.0403	0.2960	0.0077
14	-0.9680	0.3040	0.0198	0.0230
15	-0.9379	0.0207	0.0179	0.0301
16	-0.9991	0.0040	0.0042	0.0022
17	-0.6625	0.0418	0.0308	0.0200
18	-0.5100	0.0405	0.1450	0.0975
19	-0.8025	0.2263	-0.8379	0.4375
20	3	-0.6125	2	0.2020
21	-0.7575	-0.9913	0.3002	-0.9984
22	0.0513	0.3162	0.0309	-0.9909
23	0.0364	0.0307	0.0482	0.0288
24	0.0281	0.0838	0.0333	0.1640
25	0.0151	0.0545	0.2972	0.0352
26	0.0153	0.3031	0.0406	0.0194
27	0.4636	0.0371	0.0079	0.0207
28	0.0861	0.0796	0.0309	0.0263
29	0.1070	0.0313	-0.7931	0.1975
30	-0.5575	-0.5600	-0.7550	0.0310
31	-0.8495	-0.6125	0.0572	-0.9639
32	0.1264	0.3343	0.0223	-0.9893
33	0.0345	0.2875	0.3158	0.3042
34	0.1844	0.1513	0.3186	0.1450
35	0.2953	0.3245	0.3155	0.2425

	1	2	3	4
36	0.0975	0.3167	0.1073	0.0207
37	0.0239	0.2425	0.0442	0.0621
38	0.1975	0.0519	0.0418	0.1521
39	0.0191	0.6225	-0.7931	0.0814
40	-0.6550	-0.7905	-0.7831	0.1322
41	-0.6575	-0.7525	0.0708	-0.6575
42	0.1950	0.3664	0.5725	-0.8429
43	0.1289	0.1652	0.0236	2
44	0.1596	0.1014	0.0425	0.4362
45	0.1950	0.4015	0.0377	0.3096
46	0.0167	0.3234	0.0315	0.1975
47	0.0310	0.1717	0.3230	0.0176
48	0.0492	0.2211	0.1169	0.0388
49	0.3450	0.6225	-0.5600	0.0380
50	-0.5600	4	1	0.1169
51	-0.7075	-0.8906	0.0797	1
52	0.1450	0.0280	0.3348	-0.8525
53	0.1925	0.0291	0.3209	0.0283
54	0.3105	0.0358	0.1296	0.0305
55	0.1950	0.0607	0.6225	0.0451
56	0.3925	0.4506	0.0671	0.0796
57	0.0550	0.0772	0.2450	0.3875
58	0.2619	0.1298	2	0.6275
59	1	1	5	3
60	4	5	4	2
61	-0.7550	-0.6625	0.0330	-0.8928
62	0.0513	0.0192	0.3114	-0.8050
63	0.0766	0.1215	0.0436	0.0618
64	0.1950	0.4086	0.3245	0.0467
65	0.1190	0.2652	0.1689	0.0658
66	0.4825	0.5300	0.3210	0.1311
67	0.0513	0.1711	0.5036	0.3925
68	0.4300	0.1739	0.6275	0.4875
69	1	3	2	0.1475
70	5	1	2	1

	1	2	3	4
69	1	3	2	0.1475
70	5	1	2	1
71	-0.8854	-0.4625	0.1925	-0.5125
72	0.0378	0.1595	0.1450	-0.4625
73	0.2544	0.2425	0.2875	0.0664
74	0.2950	0.3611	0.4217	0.2975
75	0.2367	0.1950	0.4186	0.0516
76	0.6775	0.7725	0.3314	0.1232
77	0.0468	0.3541	0.3486	0.1263
78	0.3400	0.3284	0.5725	0.5775
79	4	0.5825	4	0.2425
80	4	3	3	3
81	-0.9760	-0.5600	5	4
82	0.7225	0.3425	0.2950	3
83	0.1475	0.7725	0.3400	0.3400
84	1	0.0838	0.3995	0.1811
85	0.2950	0.0291	0.5275	0.1475
86	0.4875	0.5300	0.6275	0.7225
87	0.0491	1	0.3495	0.2900
88	0.0975	0.2261	0.5329	0.2950
89	0.5725	5.0019	1	0.3900
90	1	1	2	3
91	-0.7406	4	4	-0.4625
92	0.7250	3	2	-0.6125
93	2	0.4875	0.1450	0.7725
94	0.2119	0.3400	0.5250	0.3375
95	0.2095	0.0670	0.2045	0.0749
96	0.6750	2	0.0338	0.2185
97	0.2425	0.0332	0.1150	0.1059
98	0.4024	0.4875	0.6750	0.0279
99	0.2625	4.7488	99.9880	4
100	0	0	0	0

Figure 1, reward for all episodes:



## Conclusion

In all runs, the number of actions taken by the robot to reach the goal is reduced from the first learning episode to the final learning episode. You can see this in the number of arrows in the first learning episode, compared to the lower number of arrows in the final episode for each run.