

A APPENDIX

Infinite horizon discounted LQR. Given a deterministic discrete-time linear system as in (6) and a quadratic γ -discounted cost function (1), the optimal policy is (see e.g., [4, Chapter 4.3])

$$\pi^*(x) = K^*(x - x^*), \quad K^* = -\gamma(R_Y + \gamma B^\top P B)^{-1} B^\top P A, \quad (12)$$

where K^* is the discounted optimal gain and P is the unique positive solution of the discounted Discrete-time Algebraic Riccati Equation (DARE)

$$P = Q_Y + \gamma A^\top (P - \gamma P B (R_Y + \gamma B^\top P B)^{-1} B^\top P) A. \quad (13)$$

From [6, Section 3], the value functions for the discounted LQR problem under the optimal controller are

$$J^*(x) = x^\top P x, \quad Q^*(x, u) = z^\top H z, \quad (14)$$

with $z := \text{col}(x, u) \in \mathbb{R}^{n+m}$ and H given by

$$H = \begin{pmatrix} Q_Y + \gamma A^\top P A & \gamma A^\top P B \\ \gamma B^\top P A & R_Y + \gamma B^\top P B \end{pmatrix}.$$

Table 4: Pendulum swing-up environment parameters

| Parameters | Training | Corrupted |
|------------------------------|----------|--------------------------------|
| pole mass m | 1 | 1.2 |
| pole length l | 1 | 1 |
| gravity acceleration g | 9.81 | 9.81 |
| episode max length T | 200 | 1000 |
| input bounds | [-2,2] | [-2,2] |
| noise w | 0 | $w \sim \mathcal{N}(0, 0.1)$ |
| disturbance d | 0 | $0.2 \sin(\frac{2\pi}{100} t)$ |
| steady-state threshold t_0 | - | 500 |

Table 6: Inverted pendulum swing-up environment parameters

| Parameters | Training | Corrupted |
|------------------------------|------------|--------------------------------|
| cart mass m_c | 10.47 | 10.47 |
| pole mass m_p | 5 | 6.53 |
| pole length l_p | 0.6 | 0.8 |
| rail bounds l_r | [-1,1] | [-1,1] |
| episode max length T | 1000 | 1000 |
| input bounds | [-100,100] | [-100,100] |
| noise w | 0 | $w \sim \mathcal{N}(0, 0.173)$ |
| disturbance d | 0 | $20 \sin(\frac{2\pi}{50} t)$ |
| steady-state threshold t_0 | - | 500 |

Table 7: Double inverted pendulum swing-up environment parameters

| Parameters | Training | Corrupted |
|------------------------------|------------|--------------------------------|
| cart mass m_c | 10 | 10 |
| first pole mass m_{p1} | 1 | 1 |
| second pole mass m_{p2} | 1 | 1.2 |
| first pole length l_{p1} | 0.6 | 0.6 |
| second pole length l_{p2} | 0.6 | 0.7 |
| rail bounds l_r | [-2,2] | [-2,2] |
| episode max length T | 1000 | 2000 |
| input bounds | [-200,200] | [-200,200] |
| noise w | 0 | $w \sim \mathcal{N}(0, 0.173)$ |
| disturbance d | 0 | $20 \sin(\frac{2\pi}{100} t)$ |
| steady-state threshold t_0 | - | 1500 |

Table 5: Hyperparameters. lr: learning rate, af: activation function, NN: hidden layer sizes

| | PSU | | | IPSU | | | DIPSU | | |
|----------|-------|------|---------|---------|------|---------|--------|------|---------|
| | lr | af | NN | lr | af | NN | lr | af | NN |
| DDPG | 0.001 | ReLU | 400,300 | 0.001 | ReLU | 400,300 | 0.0001 | ReLU | 400,300 |
| LAS-DDPG | 0.001 | ReLU | 400,300 | 0.001 | ReLU | 400,300 | 0.0001 | Tanh | 400,300 |
| PPO | 0.003 | Tanh | 64,64 | 0.00025 | Tanh | 256,256 | 0.0001 | Tanh | 64,64 |
| LAS-PPO | 0.003 | Tanh | 64,64 | 0.00025 | Tanh | 256,256 | 0.0001 | Tanh | 64,64 |
| TD3 | 0.001 | ReLU | 400,300 | 0.001 | ReLU | 400,300 | 0.0006 | ReLU | 400,300 |
| LAS-TD3 | 0.001 | ReLU | 400,300 | 0.001 | ReLU | 400,300 | 0.0002 | Tanh | 256,256 |
| SAC | 0.001 | ReLU | 256,256 | 0.001 | ReLU | 256,256 | 0.001 | ReLU | 256,256 |
| LAS-SAC | 0.001 | ReLU | 256,256 | 0.001 | ReLU | 256,256 | 0.0001 | Tanh | 256,256 |