

## Contents

<b>1 Notes for the project:</b>	<b>1</b>
1.1 MPC Control: . . . . .	1
1.1.1 MPC Control . . . . .	1
1.2 DDPG: . . . . .	2

## 1 Notes for the project:

### 1.1 MPC Control:

I have a state space model for the mpc control used to control the opening of a valve Out variables should control  $\text{last}_{\text{outputvelocity}}$ ,  $\text{last}_{\text{outputdisplacement}}$   
How does MPC control work?

$$\dot{x}(t) = Ax(t) + Bu(t) \quad (1)$$

$$y(t) = Cx(t) + Du(t) \quad (2)$$

#### 1.1.1 MPC Control

MPC is based on iterative, finite-horizon optimization of a plant model. At time  $t$  the current plant state is sampled and a cost minimizing control strategy is computed (via a numerical minimization algorithm) for a relatively short time horizon in the future:  $[t, t + T]$  Model predictive control is a multivariable control algorithm that uses:

- an internal dynamic model of the process
- a cost function  $J$  over the receding horizon
- an optimization algorithm minimizing the cost function  $J$  using the control input  $u$

#### 1. Equations:

$$n_x = 6, N = 1$$

- Variables  $x, u$
- constraints:
  - $x_{k+1} = A_d x_k + B_d u_k$

$$- u_k \geq -10, u_k < 0$$

Controlup outputs the predicted control signal  $u[:, 0]$

Adapt the control methods to provide the predicted state of the system to the DDPG framework

## 1.2 DDPG:

Reinforcement learning technique that strives to combine perception capabilities of Deep learning with Decision capabilities of conventional reinforcement learning.

- The perception step that gives information about the environment.
- The decision step that self actualizes to get an appropriate response.