Model Predictive Control Prof. Colin Jones

ME-425

# Exercise sheet 5 Tracking MPC

Consider the discrete-time system

$$x_{k+1} = \begin{pmatrix} 0.7115 & -0.4345 \\ 0.4345 & 0.8853 \end{pmatrix} x_k + \begin{pmatrix} 0.2173 \\ 0.0573 \end{pmatrix} u_k$$

$$y_k = \begin{pmatrix} 0 & 1 \end{pmatrix} x_k + d$$
(1)

where d is an *unknown* constant disturbance and  $x_0$  is unknown. The goal of this exercise sheet is to design a controller able to track a constant output reference while fulfilling input constraints

$$-3 \le u_k \le 3$$
.

## Exercise 1 (Observer design)

Since state and disturbances are unknown at time zero, we need to design an observer to estimate them. We call  $\hat{x}$  and  $\hat{d}$  the estimate of x and d, respectively. Design an observer for the given system, and test it for the condition  $x_0 = [1; 2]$ ,  $\hat{x}_0 = [3; 0]$ ,  $\hat{d}_0 = 0$ , d = 0.2 and u = 0.

#### Hints:

- Similarly to the previous exercise sheets, you can use YALMIP to implement the MPC controller
- To estimate the disturbance you will have to augment the state as seen in class
- Note the eigenvalious of (A + LC) are the same of (A' + C'L')
- The matlab function K = place(A,B,F) computes a state-feedback matrix K such that the eigenvalues of A BK are those specified in the vector F

## Deliverables:

- Matlab code computing the observer
- A plot showing that the estimates converge to the true values

## **Exercise 2** (Steady-state target computation)

Given the system above, and a reference r, use YALMIP to compute a steady state for the system that minimizes  $u^2$ .

## **Exercise 3** (MPC tracking)

Implement an MPC controller to track an output reference signal r.

## Hints:

• Use a terminal set of  $\mathcal{X}_f = \mathbb{R}^n$  and a terminal cost of  $V_f(x_N) = \Delta x_N' P \Delta x_N$  where P is the solution of P - A'PA = Q

These correspond to a terminal controller of u=0. Note that this is a valid terminal controller because the system is stable.

You can use the function P = dlyap(A,Q) to compute P.

- Good values for the horizon and stage costs are: N = 5, Q = I, R = 1
- In the previous exercise you designed an observer specifying the eigenvalues of the estimation error state-update matrix. Eigenvalues with a small norm will speed up the estimation process, but may increase the initial overshoot of the estimate  $\hat{d}$ . A large  $\hat{d}$  can cause the problem of computing the set-point to be infeasible. Use moderate eigenvalues (e.g. 0.5, 0.6, 0.7).

#### Deliverables:

- Matlab code computing your controller
- Plots showing for the references r = 1 and r = 0.5:
  - the estimates converge to the true values
  - the output converges to the reference
  - the input does not violate the constraints