

# 24 – Discrete control



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## Design for purely discrete systems

- Many methods are analogous (or identical) to continuous methods, so we do not mention them, just show them on examples.
- These methods are briefly described in additional presentation.
- Here we focus on differences in a discrete design.

## Discrete Control Design for Sampled Continuous Systems

- We come out of a discrete system model and use discrete design methods.
- CL stability is guaranteed (by a stabilizing discrete controller), unlike the emulation methods.
- Control works well at sampling times (within a reasonable period of time)
- On the contrary, we do not control the behavior between samples.
- Behavior between sampling times is fine unless action is "very wild"



- Typically discrete control design strategy, continuous (not exactly) work
- Goal = all poles „to zero“ i.e.  $p_{\text{new}}(z) = z^n$   
so for the result holds (Cayley-Hamilton)  $\mathbf{F}_{\text{new}}^n = 0$
- Because equation  $\mathbf{x}_{k+1} = \mathbf{F}_{\text{new}} \mathbf{x}_k$  has solution  $\mathbf{x}_k = \mathbf{F}_{\text{new}}^k \mathbf{x}_0$ ,  
and  $\mathbf{x}_k = 0, \forall k \geq n, \forall \mathbf{x}_0$
- Thus, regardless of the initial state, the system is completely at the steady state with n-th step (all states and output are zero)!
- In case of disturbance, it also gets to steady state (on latest in the n-th step after the disturbance has vanished)

- The resulting system is valid for

$$\mathbf{x}(z) = (z\mathbf{I} - \mathbf{F}_{\text{new}})^{-1} \mathbf{G}u(z) + z(z\mathbf{I} - \mathbf{F}_{\text{new}})^{-1} \mathbf{x}_0 = \frac{\text{adj}(z\mathbf{I} - \mathbf{F}_{\text{new}}) \mathbf{G}}{z^n} u(z) + \frac{z \text{adj}(z\mathbf{I} - \mathbf{F}_{\text{new}})}{z^n} \mathbf{x}_0$$

$$y(z) = \frac{\mathbf{H} \text{adj}(z\mathbf{I} - \mathbf{F}_{\text{new}}) \mathbf{G}}{z^n} u(z) + \frac{z \mathbf{H} \text{adj}(z\mathbf{I} - \mathbf{F}_{\text{new}})}{z^n} \mathbf{x}_0 = \frac{b(z)}{z^n} u(z) + \frac{c_{x_0}(z)}{z^n}$$

$$\deg(b(z), c_x(z)) \leq n$$



- The deadbeat system is the most stable of all discrete ones. For **continuous** systems **holds the same**: No later than  $n$  steps, from each initial state is at steady state – i.e. all states and output are zero. **even between sampling times!** Continuous control doesn't allow that!
- The deadbeat system responds very quickly: sometimes it is beneficial but sometimes disadvantageous (if there is noise in the system).
- Shortening sampling period, size of input signals increases, and limit of  $h \rightarrow 0$  goes to infinity. Therefore, we do not reduce the period.
- Deadbeat is designed just like any other pole assignment.

- Because  $p_{new}(z) = z^n$ , is Ackermann's formula specially

$$\mathbf{K} = [0 \quad \dots \quad 0 \quad 1] \mathbf{C}^{-1} \mathbf{F}^n$$

- If matrix  $\mathbf{F}$  is invertible, it holds that

$$\mathbf{K} = [0 \quad \dots \quad 0 \quad 1] [\mathbf{F}^{-n} \mathbf{G} \quad \mathbf{F}^{-n+1} \mathbf{G} \quad \dots \quad \mathbf{G}]^{-1}$$



- Even when estimating states by a discrete observer, we can use the deadbeat strategy.
- We choose the characteristic polynomial of the observation dynamics

$$p_{\text{poz}}(z) = z^n, \text{ so } \mathbf{F}^n = \mathbf{0}$$

- For the observation error  $\tilde{\mathbf{x}} = \mathbf{x} - \hat{\mathbf{x}}$ , where  $\tilde{\mathbf{x}}_{k+1} = \mathbf{F}_{\text{poz}} \tilde{\mathbf{x}}_k$  it holds that

$$\tilde{\mathbf{x}}_k = \mathbf{0}, \forall k \geq n, \forall \tilde{\mathbf{x}}_0$$

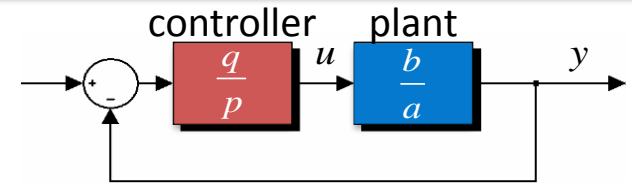
- Thus latest at n-th step, the deviation is zero, and from that moment on, the observer status is exactly equal to the state of the system.
- If the system is continuous, the same applies, but only at samples.
- Such an observer will be proposed, for example, by modifying the Ackermann's formula

$$\mathbf{L} = \mathbf{F}^n \mathbf{O}^{-1} [0 \quad \dots \quad 0 \quad 1]^T$$



# Polynomial Pole Assignment

## Polynomial solution in $z$ – same as continuous



- For given a system  $b(z)/a(z)$   
and given pole positions, given by CL characteristic pol.  $c(z)$
- we solve the equation  $a(z)p(z) + b(z)q(z) = c(z)$

## Polynomial solution in $d$

- Similarly  $b(d)/a(d), c(d) \Rightarrow a(d)p(d) + b(d)q(d) = c(d) \Rightarrow q(d)/p(d)$

## Deadbeat polynomial – special case of pole assignment

- In  $z$  we choose  $c(z) = z^m$ , where  $m \geq (2 \times \text{system order}) - 1$ , solve

$$a(z)p(z) + b(z)q(z) = z^m$$

and choose the solution of minimum degree  $q$

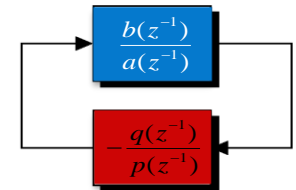
- By solving in  $z^{-1}$  is the solution simpler: We solve the equation

$$a(z^{-1})p(z^{-1}) + b(z^{-1})q(z^{-1}) = 1$$



# Stabilization by discrete controller

- If we work in  $z$ , then the parameterization of all the stabilizing regulators is the same as in cont. time.
- If we work in  $z^{-1}$ , it is even simpler:



All stabilizing controllers are parametrized as follows

$$\frac{q}{p} = \frac{y + at}{x - bt}$$

- Where  $t$  is any fraction of polynomials with a stable denominator and polynomials  $x, y$  fulfill  $\bar{a}x + \bar{b}y = 1$
- where  $a = (a, b)\bar{a}$ ,  $b = (a, b)\bar{b}$

Solvability:

- The system does not have unstable hidden modes and  $\gcd(a, b)$  is stable.

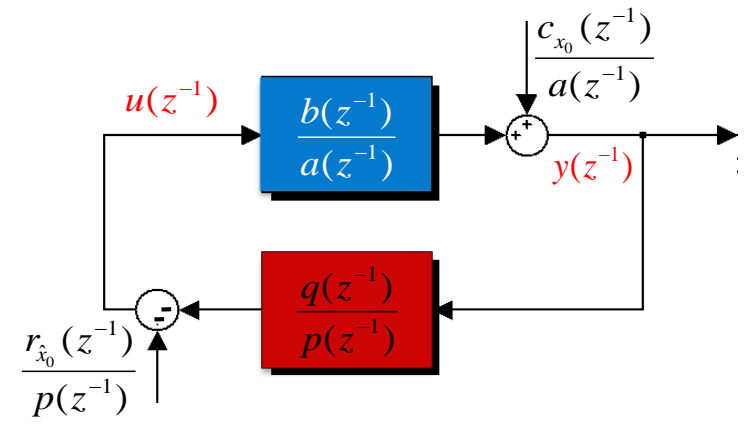


# Deriving a weak and a strong version of deadbeat

- The design of deadbeat regulator in  $z^{-1}$  is very similar to that in  $z$
- So, for the sake of interest, at least choose the opposite derivation:
- We have considered deadbeat to be a special case of poles (When all poles are placed to origin, the behavior will be deadbeat)
- Now, look for a regulator so that the behavior is of the deadbeat type:
- In addition, we will work with polynomials in  $z^{-1}$  (it is more convenient).

## Problem formulation: Deadbeat controller – strong version

- Find the controller so that the input and output sequences have a final length for each i.c. of plant and controll
- Moreover, in order to have the shortest length (the smallest number of steps).







# Derivation: Deadbeat – strong version

- We derive from the equations of the system

$$y = -\frac{b}{ap + bq} r_{\hat{x}_0} + \frac{p}{ap + bq} c_{x_0}$$

$$u = -\frac{a}{ap + bq} r_{\hat{x}_0} - \frac{q}{ap + bq} c_{x_0}$$

- The goal is to find polynomials  $p(z^{-1}), q(z^{-1})$  such, that for all possible polynomials  $r_{\hat{x}_0}(z^{-1}), c_{x_0}(z^{-1})$  the resulting sequences were polynomials (= had a final length).
- Because  $r_{\hat{x}_0}, c_{x_0}$  are not give (they represent different init. cond.), each of the four fractions must be polynomial in itself.
- A polynomial fraction is generally an infinite sequence. The endpoint will be when the denominator divides the numerator without the rest.
- However, we can show that we can not simultaneously denote the denominator  $ap + bq$  with all four numerators  $a, b, p, q$
- Each of the fractions is a polynomial  $\iff a(z^{-1})p(z^{-1}) + b(z^{-1})q(z^{-1}) = 1$



# Derivation: Deadbeat – strong version

- It holds that  $ap + bq = 1$ , then the above relationships are given

$$y(z^{-1}) = b(z^{-1})r_{\hat{x}_0}(z^{-1}) + p(z^{-1})c_{x_0}(z^{-1})$$

$$u(z^{-1}) = a(z^{-1})r_{\hat{x}_0}(z^{-1}) - q(z^{-1})c_{x_0}(z^{-1})$$

- Because everything on the right are polynomials, they are  $y(z^{-1}), u(z^{-1})$  polynomial for all initial condition as well, which had to be ensured.
- The length of sequences is given by the degree of polynomials +1
- The only option for shortening in some i.c.,  $\left[ p(z^{-1}), q(z^{-1}) \right]$  is to choose a solution of the minimum degree.

## Summary of the solution

- To find a strong version of the deadbeat controller, you need to solve the polynomial equation  $a(z^{-1})p(z^{-1}) + b(z^{-1})q(z^{-1}) = 1$  and from possible solutions to select the one with minimum degree  $\left[ p(z^{-1}), q(z^{-1}) \right]$
- Solvability:  $a, b$  relatively prime (controllable and observable system), there must also be no hidden modes.



## Problem formulation: Deadbeat controller – weak version

- Find the controller so that the output sequence  $y$  has a final length, for each initial conditions for both, the system and the controller.
- Moreover, so that it has the shortest length.
- At the same time the CL system must be stable.
- Difference against strong version:  $u$  can be infinitely long but stable.

## Solution

- We are only interested in one formula and then we can cancel against  $b$ .
$$y = -\frac{b}{ap + bq} r_{\hat{x}_0} + \frac{p}{ap + bq} c_{x_0}$$
- Divide  $b$  to stable and unstable factors  $b(z^{-1}) = b^+(z^{-1})b^-(z^{-1})$
- and set  $ap + bq = b^+$



- Equation  $ap + bq = b^+$  can be divided by  $b^+$  (or  $p$  must be divisible) we obtain  $ax + b^-q = 1$ , where  $p = xb^+$
- Then the output relationship switches to  $y = b^- r_{\hat{x}_0} + xc_{x_0}$
- Thus, the output is a polynomial for all initial conditions.
- The shortest march we get by choosing a solution with  $x$  min. degree.
- We'll check the input  $u$ : (it is not finite but it is stable)

$$u = -\frac{a}{ap + bq} r_{\hat{x}_0} - \frac{q}{ap + bq} c_{x_0} = -\frac{a}{b^+} r_{\hat{x}_0} - \frac{q}{b^+} c_{x_0}$$

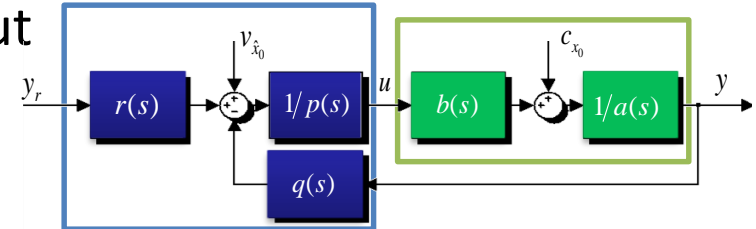
- CL system is stable as well, or  $ap + bq = b^+$

## Solution summary

- Divide  $b(z^{-1}) = b^+(z^{-1})b^-(z^{-1})$
- Solve equation  $a(z^{-1})x(z^{-1}) + b^-(z^{-1})q(z^{-1}) = 1$  for  $x$  min. degree.
- Solvability:  $a, b^-$  relatively prime  $\gcd(a, b^-)$  stable.



- If the control system has a reference input it is natural to use a controller with two degrees of freedom (2DOF)



$$u(z) = -\frac{q(z)}{p(z)} y(z) + \frac{r(z)}{p(z)} y_r(z)$$

$$p(z)u(z) = -q(z)y(z) + r(z)y_r(z)$$

- For causality must be  $\deg p \geq \deg [q(z), r(z)]$
- Classic deviation control (1DOF) is a special case when  $q(z) = r(z)$
- We compute the FB part of the known equation when designing

$$a(z)p(z) + b(z)y(z) = c(z)$$

- Where we suitably choose a CL characteristic polynomial
- From a comparison with a state approach it follows that  $c(z) = c_c(z)c_o(z)$
- Where the factors are  $c_c(z) = \det(zI - F + GK)$ ,  $c_o(z) = \det(zI - F + LH)$




- The resulting system transfer function is

$$y(z) = \frac{b(z)r(z)}{a(p)p(z) + b(z)q(z)} y_r(z)$$
$$= \frac{b(z)r(z)}{c(z)} y_r(z) = \frac{b(z)r(z)}{c_c(z)c_o(z)} y_r(z)$$

- We choose the **direct branch** to cancel the poles of the observer, i.e.  $c_o(z) \mid r(z)$  for example as

$$r(z) = t_0 c_o(z)$$


$$y(z) = \frac{t_0 b(z)}{c_c(z)} y_r(z)$$

- Then the control do not to generate a deviation of the observations
- constant  $t_0$  is chosen to ensure required static gain
- Usually, the static gain should be = 1, therefore we set  $t_0 = c_c(1)/b(1)$



**Asymptotic tracking** is the same as for continuous systems

- Equations are the same

$$ap + bq = m, \quad f^-t + br = m, \quad m \text{ and stable}$$

- Conditions are the same

1)  $\gcd(a, b)$  stable; 2)  $\gcd(f^-, b) = 1$ ; 3)  $f^- \mid a$

- Solution is the same in  $z$  and in  $z^{-1}$ , Except that we still have to choose  $m$  of appropriately high degree when dealing with  $z$

Unlike in a continuous case, there is a deadbeat option, which allows tracking in the final number of steps:

- If we proceed in  $z$ , we choose  $m(z) = z^{n-1}$   
if in  $z^{-1}$ , we choose  $m(z^{-1}) = 1$
- And choose the solution of minimum degrees (coincidence occurs)
- Solution exists, iff  $\gcd(a, b) = 1$   
the other conditions are the same.