

**CHALMERS UNIVERSITY OF TECHNOLOGY**  
**Department of Signals and Systems**  
**Division of Automatic Control, Automation and Mechatronics**

**EXAMINATION IN NONLINEAR AND ADAPTIVE CONTROL**

(ESS076)

Wednesday January 11, 2012

Time and place: 14.00 – 18.00, V-building

Teacher: Claes Breitholtz, phone 3718

**The following items are allowed:**

1. *Control Theory* (Glad, Ljung) or *Applied Nonlinear Control* (Slotline/Li)
2. ESS076 Supplement
3. Mathematical handbooks of tables such as Beta Mathematics Handbook

**Notes, calculator, mobile telephones, laptops or palmtops, are not allowed!**

Reasonable notes in textbook are allowed but no solved problems.

The total points achievable are 30 with the following scales for grading:

Grade 3: at least 12 points  
Grade 4: at least 18 points  
Grade 5: at least 24 points

Incorrect solutions with significant errors, unrealistic results or solutions that are difficult to follow result in 0 points.

Grading results are posted not later than January 27. Review of the grading is offered Monday January 30 at 12.30-13.30. If you cannot attend this occasion, any objections concerning the grading must be filed in written form not later than two weeks after the regular review occasion.

**Good luck!**

1. Consider the following nonlinear system:

$$\frac{d}{dt} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_2 \\ (2 - \cosh x_1) \cdot x_2 - \sinh x_1 \end{bmatrix}$$

Compute the stationary points of the system and formulate the corresponding linearized systems. What conclusion regarding local stability of the zero solution can be drawn from *Lyapunov's linearization method* (section 12.1 in the text-book)?

**3 points**

2. Consider the following nonlinear system:

$$\frac{d}{dt} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -x_2 - x_1^3 \\ 3x_1 - 2x_2^5 \end{bmatrix}$$

Show that the above system is asymptotically stable. Is the stability global or local?

**4 points**

3. The linear system  $y(t) = G(p)u(t)$  is given. The transfer function  $G$  is given by

$$G(p) = \frac{p^3 + 2p^2 + 3p + 8}{(p + 1)^4}$$

Is it possible to write this system as a linear state space model, in such a way that two positively definite matrices,  $P$  and  $Q$ , according to the Kalman-Yakubovich Lemma is found? Motivate Your answer carefully!

**4 points**

4. The linear time invariant system

$$G(s) = \frac{1}{(s+1)^6}$$

is connected to the nonlinear function given by  $output = \text{sign}\{input\}$  in a feedback structure. Investigate the possibility of periodic solutions by use of the describing function method. Also, give estimates of the possible frequency and amplitude of an oscillation.

**6 points**

5. Consider the nonlinear system:

$$\frac{d}{dt} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_2 + 2x_1^2 \\ x_3 + u \\ x_1 - x_3 \end{bmatrix}, y = x_1$$

Design a state feedback control law such that the output  $y$  asymptotically tracks the reference signal  $r(t) = \sin(\omega t)$  (where  $\omega$  is an arbitrary known frequency).

**4 points**

6. A time-varying automotive system is modelled by the law of motion

$$\frac{d}{dt} \left( m(t) \cdot \frac{dy}{dt} \right) = u(t)$$

where  $y$  is the position,  $m$  is the mass of the vehicle and  $u$  is the applied force. In principle, the total mass is monotonously decreasing due to loss of fuel. However, loss of fuel is a rather **slow process** compared to changes in speed and position. Assume that there is a separate estimation system for the continuous accurate update of the mass and its derivative. Can a gain scheduled **proportional controller**  $K_p$ , using the estimates of  $m$  and  $\dot{m}$ , be used in a system where the position  $y$  should track a given reference  $y_{ref}$ ? The resulting phase margin should be 45 degrees. Motivate carefully!

**3 points**

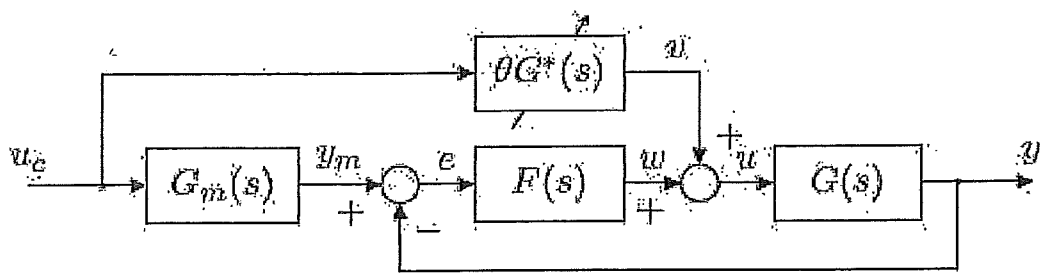
7. In this problem, we shall study an adaptive controller as depicted in the figure below. The controller uses a combination of feedforward and feedback. The process is given by the transfer function

$$G(s) = G_n(s)/\theta_0$$

where  $G_n(s)$  is assumed to be known with  $G_n(0) = 1$  and  $1/\theta_0$  is the unknown steady-state gain of the process.  $F(s)$  is a fixed regulator and  $G_m(s)$  is a reference model for the desired closed-loop system. The feedforward signal  $v(t)$  is given by

$$v(t) = \theta G^*(p)u_c(t) = \theta[G_m(p)/G_n(p)]u_c(t)$$

where  $G^* = G_m/G_n$  is a fixed filter and the scalar  $\theta$  shall be adapted online.



(a) Show that the desired closed-loop transfer function (from  $u_c$  to  $y$ ) is obtained if  $\theta$  has the correct value  $\theta_0$ .

**2 points**

(b) Determine an adaption law for  $\theta$  using the MIT-rule, and using the criterion  $J = w^2$ . (Hint: Use the approximation  $FG/(1+FG) \approx 1$ .)

**2 points**

(c) Derive an error model for  $w$  and show stability of the adaptation rule in (b) under suitable assumptions.

**2 points**

$$1. \quad \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} k_2 \\ (2 - \cosh k_1)x_2 - \sinh k_1 \end{bmatrix}$$

Stationary points:  $k_2 = 0 \Rightarrow \sinh k_1 = 0 \Rightarrow$   
 $k_1 = 0 \Rightarrow (0, 0)$  is the  
 only stat. point.

$$\begin{bmatrix} \Delta \dot{x}_1 \\ \Delta \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -\sinh k_1 \cdot k_2 - \cosh k_1 & 2 - \cosh k_1 \end{bmatrix} =$$

$$= \begin{pmatrix} 0 & 1 \\ -1 & 1 \end{pmatrix}$$

$x_1 = 0$   
 $x_2 = 0$

$$\begin{vmatrix} \lambda & -1 \\ 1 & \lambda - 1 \end{vmatrix} = \lambda^2 - \lambda + 1 = 0 \Rightarrow$$

both eigenvalues are  
 in the RHP  $\Rightarrow$  unstable!

$$2. \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} -x_2 - x_1^3 \\ 3x_1 - 2x_2^5 \end{bmatrix}$$

$$\text{Try } V = ax_1^m + bx_2^n$$

$$\begin{aligned} \dot{V} &= -(x_2 + x_1^3) \frac{\partial V}{\partial x_1} + (3x_1 - 2x_2^5) \frac{\partial V}{\partial x_2} = \\ &= -(x_2 + x_1^3) m a x_1^{m-1} + (3x_1 - 2x_2^5) n b x_2^{n-1} = \\ &= -a m x_1^{m-1} x_2 - a m x_1^{m+2} + \\ &\quad + 3b n x_1 x_2^{n-1} - 2b n x_2^{n+4} \end{aligned}$$

Make  $m-1=1 \Rightarrow m=2$  and

$n-1=1 \Rightarrow n=2$

$$\begin{aligned} \dot{V} &= \underbrace{(6b - 2a)}_{=0} x_1 x_2 - 2a x_1^4 - 4b x_2^6 = \\ &= -\frac{3}{2} x_1^4 - x_2^6 = -\left(\frac{3}{2} x_1^4 + x_2^6\right) < 0 \end{aligned}$$

$\forall x_1$  and  $x_2 \neq 0. \Rightarrow$

$\dot{V}$  is a neg. def. function everywhere  
 $\Rightarrow$  the origin is globally asympt. stable.

3. Given  $G(s)$  a <sup>minimal order</sup> state-space model is (for example)

$$\dot{x} = \begin{bmatrix} -4 & 1 & 0 & 0 \\ -6 & 0 & 1 & 0 \\ -4 & 0 & 0 & 1 \\ -1 & 0 & 0 & 0 \end{bmatrix} x + \begin{bmatrix} 1 \\ 2 \\ 3 \\ 8 \end{bmatrix} u$$

$$y = [1 \ 0 \ 0 \ 0] x \quad (\text{or any other state-space model!})$$

Consider the Lyapunov equation

$$PA + A^T P = -Q$$

At-Y-Lemma states that a necessary condition for finding  $P$  and  $Q$  ( $\text{both} > 0$ ) is that ~~the all eigenvalues in LHP.~~  
 $G(s)$  is a SPR

$P > 0$  and  $Q > 0 \Rightarrow G(s)$  is SPR.

$$G(s) = \frac{P^3 + 2P^2 + 3P + 8}{(s+1)^4}$$

The polynomial  $P^3 + 2P^2 + 3P + 8$  does not have all its zeros in the LHP  $\Rightarrow$

$G(s)$  is non-minimum phase  $\Rightarrow$

$\Rightarrow G(s) = C(sI - A)^{-1} B$  is not SPR  $\Rightarrow$

$\Rightarrow$  You cannot find  $P$  and  $Q > 0$ !

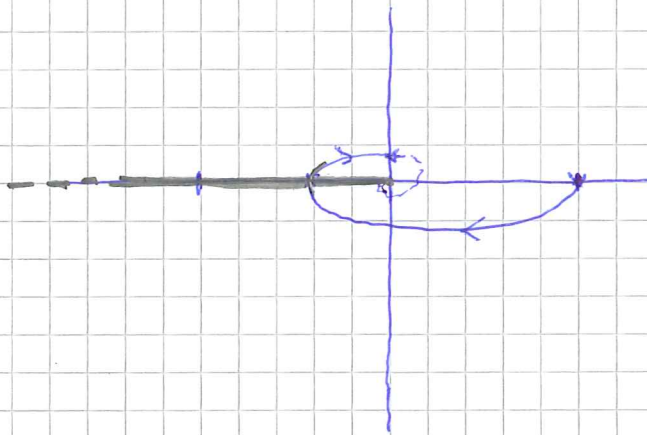
4.

$$G(j\omega) = \frac{1}{(j\omega + 1)^6}$$

$$|G(j\omega)| = \left( \frac{1}{\sqrt{\omega^2 + 1}} \right)^6 = \frac{1}{(\omega^2 + 1)^3}$$

$$\angle G(j\omega) = -6 \arctan(\omega)$$

$\omega$	$ G $	$\angle G$
0	1	0
$1/\sqrt{3}$	$27/64$	$-180^\circ$
1	$1/8$	$-270^\circ$
$\infty$	0	$-540^\circ$



$$Y_P(c) = \frac{4}{\pi c} \Rightarrow$$

$$\Rightarrow -\frac{1}{Y_P(c)} = -\frac{\pi c}{4}$$

One possible cross:  $-\frac{27}{64} = -\frac{\pi}{4} c$

$$\Rightarrow c = \frac{27}{16\pi} (\approx 0.537 \approx 0.54)$$

and  $\omega = \frac{1}{\sqrt{3}} (\approx 0.577 \approx 0.58 \text{ rad/sec})$



$$5. \begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \end{pmatrix} = \begin{pmatrix} x_2 + 2x_1^2 \\ x_3 + u \\ x_1 - x_3 \end{pmatrix}, \quad y = x_1$$

$$\dot{y} = \dot{x}_1 = x_2 + 2x_1^2$$

$$\ddot{y} = \dot{x}_2 + 4\dot{x}_1 x_1 = 4(x_2 + 2x_1^2)x_1 + x_3 + u$$

Minimum phase?  $y(t) \equiv 0 \Rightarrow \dot{x}_3 = -x_3 \Rightarrow$

Minimum phase, as the zero-dynamics are clearly stable! Let  $e = y - y_R$ :

$$\dot{e} = \dot{y} - \dot{y}_R = x_2 + 2x_1^2 - \dot{y}_R$$

$$\ddot{e} = \ddot{y} - \ddot{y}_R = x_3 + u + 4(x_2 + 2x_1^2)x_1 - \ddot{y}_R$$

$$\text{Take } u = -x_3 - 4(x_2 + 2x_1^2)x_1 + \dot{y}_R - k_1 e - k_2 \dot{e} \Rightarrow$$

$$\ddot{e} + k_2 \dot{e} + k_1 e = 0$$

which is a stable system for  $k_1 > 0, k_2 > 0$

$$\Rightarrow e(t) \rightarrow 0 \text{ as } t \rightarrow \infty$$

$$5. \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \end{bmatrix} = \begin{bmatrix} x_2 + 2x_1^2 \\ x_3 + u \\ x_1 - x_3 \end{bmatrix} = \begin{bmatrix} x_2 + 2x_1^2 \\ x_3 \\ x_1 - x_3 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} u$$

$$y = x_1 \Rightarrow \dot{y} = \dot{x}_1 = x_2 + 2x_1^2 \quad (\text{not degree 1!})$$

$$\ddot{y} = \dot{x}_2 + 4x_1 \dot{x}_1 = x_3 + u + 4x_1 x_2 + 8x_1^3$$

$\Rightarrow$  degree 2!

From (17.9) we have

$$u = \frac{r - L_f^2 h}{L_g L_f h} \quad h = x_1 \text{ in this case}$$

$$L_f h = \nabla h \cdot f = f_1 \frac{\partial h}{\partial x_1} + f_2 \frac{\partial h}{\partial x_2} + f_3 \frac{\partial h}{\partial x_3} =$$

$$= x_2 + 2x_1^2$$

$$L_f^2 h = L_f(L_f h) = f_1 \frac{\partial}{\partial x_1}(L_f h) + \dots + f_3 \frac{\partial}{\partial x_3}(L_f h)$$

$$= (x_2 + 2x_1^2) \cdot 4x_1 + \cancel{x_3 + u} \cdot x_3 \cdot 1 + (x_1 - x_3) \cdot 0$$

$$= 4x_1 x_2 + 8x_1^3 + x_3$$

$$L_g L_f h = \underbrace{g_1}_{=0} \frac{\partial}{\partial x_1} L_f h + \underbrace{g_2}_{=1} \frac{\partial}{\partial x_2} L_f h + \underbrace{g_3}_{=0} \frac{\partial}{\partial x_3} L_f h$$

$$u = r - (4x_1 x_2 + 8x_1^3 + x_3) \quad \text{DUH}$$

$$6. \quad \frac{d}{dt} (m \dot{y}) = m \ddot{y} + \dot{m} \dot{y} = u$$

$$\hat{m} = \alpha \quad \text{and} \quad \hat{\dot{m}} = \beta$$

$\alpha$  and  $\beta$  can be regarded as constants compared to  $y$  and  $\dot{y}$  !

$$Y(s) = \frac{U(s)}{\alpha s^2 + \beta s}$$

$$P\text{-Controller} \Rightarrow u = K_p (r - y)$$

$$T(s) = \frac{K_p}{\alpha s^2 + \beta s + K_p}$$

$$\alpha > 0 \quad \text{as} \quad m(t) > 0 \quad \forall t$$

$$K_p > 0 \quad (\text{if chosen so})$$

$$\text{But: } \beta \leq 0 \quad \forall t \quad \text{as}$$

$$m(t) \text{ decreases, then } \dot{m}(t) \leq 0 \quad \forall t$$

$$\text{So if } \beta = 0 \Rightarrow T(s) = \frac{K_p}{\alpha s^2 + K_p}$$

an oscillator!

~~if~~ if  $\beta < 0 \Rightarrow T(s)$  is unstable

A P-controller cannot stabilize the system!

The Hamiltonian is  $H = u^2 + \lambda(bu - cx) = \left(u + \frac{\lambda b}{2}\right)^2 - \frac{\lambda^2 b^2}{4} - \lambda cx$ , which is minimized for

$$u = \begin{cases} 0, & -b\lambda/2 \leq 0 \\ -b\lambda/2, & 0 < -b\lambda/2 \leq u_{\max} \\ u_{\max}, & u_{\max} < -b\lambda/2 \end{cases}$$

We also have the adjoint equation

$$\dot{\lambda} = -H_x^T = c\lambda, \quad \lambda(T) = -a \quad \Leftrightarrow \quad \lambda(t) = -ae^{c(t-T)} < 0.$$

The optimal control strategy thus becomes

$$u = \begin{cases} \frac{ba}{2}e^{c(t-T)}, & \frac{ba}{2}e^{c(t-T)} \leq u_{\max} \\ u_{\max}, & \text{otherwise} \end{cases}$$

or with numerical values,  $u(t) = \min \{44 \cdot e^{0.2(t-7)}, 30\}$ .

5.

- (a) The transfer function is determined by “the direct path divided by [one plus the loop gain]”. This leads to

$$\frac{y}{u_c} = \frac{GFG_m + GG^*}{1 + GF} = \frac{GFG_m + G\theta G_m/G_n}{1 + FG} = \frac{GFG_m + \frac{G_n}{\theta_0}\theta G_m/G_n}{1 + FG} = \frac{GF + 1}{1 + FG}G_m = G_m$$

if  $\theta = \theta_0$ .

- (b) The MIT rule uses an update law moving in the negative direction of the gradient of  $J$  with respect to  $\theta$ . We get  $\frac{\partial J}{\partial \theta} = 2w\frac{\partial w}{\partial \theta}$ . The update law should thus be

$$\dot{\theta} = -\gamma w \frac{\partial w}{\partial \theta}. \quad (1)$$

The only signal that is unclear if we can measure in this update law is  $\frac{\partial w}{\partial \theta}$ . So try to determine it by first writing  $w$  as a function of  $\theta$ :

$$w = \frac{FG_m - FG\theta G_m/G_n}{1 + FG}u_c \Rightarrow \frac{\partial w}{\partial \theta} = \frac{-FGG_m/G_n}{1 + FG}u_c \approx -\frac{G_m}{G_n}u_c$$

if we use the proposed approximation  $\frac{FG}{1+FG} \approx 1$ . So  $\dot{\theta} = \gamma w \frac{G_m}{G_n}u_c$  is an approximation of (1) we settle for with the MIT rule.

- (c) We want to reach an error model  $w = H(s)[\tilde{\theta}^T \varphi]$ , where  $\tilde{\theta} = \theta - \theta_0$ . If  $H(s)$  is SPR, the update law  $\dot{\tilde{\theta}} = \gamma w \varphi$  can be proven to be asymptotically stable with Lyapunov analysis. From (b) we have

$$w = \frac{FG_m - FG\theta G_m/G_n}{1 + FG}u_c = \frac{FG}{1 + FG} \left[ \tilde{\theta} \frac{G_m}{G_n} u_c \right],$$

so if the “closed-loop system”  $FG/(1+FG)$  is SPR, stability of the update law in (b) can be proven.

**Note.** For full points in (c), a brief analysis (or referring to the correct details/equations in the course material) for showing stability under the SPR assumption is needed.