

Student number \_\_\_\_\_

**PhD Written Qualifying Exam 2011**  
**Systems**

## SYSTEM THEORY 2011

1. For the dynamic system given by:

$$\begin{aligned}\dot{x}_1 &= x_2 \\ \dot{x}_2 &= -\sin x_1 - x_2\end{aligned}$$

- (a) Formulate the third Liapunov Theorem on the linear approximations.
- (b) Find all equilibrium points of the system and determine the corresponding linear approximations (“abridged systems”). Classify the stability of these points.
- (c) Sketch a phase portrait of this system. Clearly label equilibrium points, trajectories and directions, and areas of stability (if any).
- (d) Find a suitable Liapunov function for this system and, if possible, find a region of stability using this function.

2. Consider the following initial-value problem.

$$\begin{cases} \ddot{x} + \dot{x} = -\delta(t - 1) \\ x(0) = \dot{x}(0) = 0, \end{cases}$$

where  $\delta$  denotes the Dirac's delta.

- (a) Explain how the presence of the delta functional translates into an interface condition, and solve the problem using elementary calculus
- (b) Define the Laplace transform for the delta distribution and compute it.
- (c) Compute the Laplace transform of the initial-value problem.
- (d) Use the attached Table of Laplace Transforms to compute the inverse Laplace transform and compare it with the solution obtained using the elementary calculus.

**Table of  
Laplace Transforms**

Number	$F(s)$	$f(t), t \geq 0$
1	1	$\delta(t)$
2	$1/s$	$1(t)$
3	$1/s^2$	$t$
4	$2!/s^3$	$t$
5	$3!/s^4$	$t^3$
6	$m!/s^{m+1}$	$t^m$
7	$1/(s + a)$	$e^{-at}$
8	$1/(s + a)^2$	$te^{-at}$
9	$1/(s + a)^3$	$\frac{1}{2!}t^2e^{-at}$
10	$1/(s + a)^m$	$\frac{1}{(m - 1)!}t^{m-1}e^{-at}$
11	$\frac{a}{s(s + a)}$	$1 - e^{-at}$
12	$\frac{a}{s^2(s + a)}$	$\frac{1}{a}(at - 1 + e^{-at})$
13	$\frac{b - a}{(s + a)(s + b)}$	$e^{-at} - e^{-bt}$
14	$\frac{s}{(s + a)^2}$	$(1 - at)e^{-at}$
15	$\frac{a^2}{s(s + a)^2}$	$1 - e^{-at}(1 + at)$
16	$\frac{(b - a)s}{(s + a)(s + b)}$	$be^{-bt} - ae^{-at}$
17	$a/(s^2 + a^2)$	$\sin at$
18	$s/(s^2 + a^2)$	$\cos at$
19	$\frac{s + a}{(s + a)^2 + b^2}$	$e^{-at}\cos bt$
20	$\frac{b}{(s + a)^2 + b^2}$	$e^{-at}\sin bt$
21	$\frac{a^2 + b^2}{s[(s + a)^2 + b^2]}$	$1 - e^{-at}\left(\cos bt + \frac{a}{b}\sin bt\right)$

3. Consider the problem of optimizing the performance index

$$J = xy$$

subject to the inequality constraint

$$x + y \geq 1.$$

Find and identify (max, min) the optimal points. Only consider solutions for which  $x, y \geq 0$ . Use a slack variable and a Lagrange multiplier to solve the problem.

4. Solve the following optimal control problem:

a. Find the control history  $u(t)$  that minimizes the performance index

$$J = x_{3_f}$$

subject to the differential constraints

$$\dot{x}_1 = u$$

$$\dot{x}_2 = x_1$$

$$\dot{x}_3 = u^2/2$$

and the prescribed boundary conditions

$$t_0 = 0, \quad x_{1_0} = 0, \quad x_{2_0} = 0, \quad x_{3_0} = 0$$

$$t_f = 1, \quad x_{1_f} = 0, \quad x_{2_f} = 1$$

b. Apply the Legendre-Clebsch condition and the Weierstrass condition to verify that the solution can be a minimum.

c. What is the conjugate point condition? What does it mean if a conjugate point exists within the interval of integration?

5. (a) Starting from the batch equations:

$$\hat{x}_k = P_k(H_k^T W_k y_k + \bar{P}_k^{-1} \bar{x}_k) \quad P_k = (H_k^T W_k H_k + \bar{P}_k^{-1})^{-1}$$

derive the sequential equations:

$$P_k = (I - K_k H_k) \bar{P}_k \quad K_k = \bar{P}_k H_k^T (H_k \bar{P}_k H_k^T + W_k^{-1})^{-1}$$

$$\hat{x}_k = \bar{x}_k + K_k (y_k - H_k \bar{x}_k)$$

(b) Explain why  $H_k$  can be replaced by  $\tilde{H}_k$  in the equations in (a) to obtain

$$P_k = (I - K_k \tilde{H}_k) \bar{P}_k \quad K_k = \bar{P}_k \tilde{H}_k^T (\tilde{H}_k \bar{P}_k \tilde{H}_k^T + W_k^{-1})^{-1}$$

$$\hat{x}_k = \bar{x}_k + K_k (y_k - \tilde{H}_k \bar{x}_k)$$

(c) For processing a scalar observation using the sequential equations, what are the dimensions of  $K_k$ ?

(d) In the sequential algorithm, the estimate and covariance are propagated to a different time using the state transition matrix, as follows:

$$\bar{x}(t) = \Phi(t, t_k) \hat{x}_k \quad \bar{P}(t) = \Phi(t, t_k) P_k \Phi^T(t, t_k)$$

Starting from these equations, and given that

$$\dot{\Phi}(t, t_k) = A(t) \Phi(t, t_k)$$

derive the following alternate equations for propagating the estimate and covariance:

$$\dot{\bar{x}}(t) = A(t) \bar{x}(t) \quad \dot{\bar{P}}(t) = A(t) \bar{P}(t) + \bar{P}(t) A^T(t)$$

6. Assume you have the linearized data equation

$$y = Hx + \epsilon$$

where  $y$  is  $m \times 1$ ,  $H$  is  $m \times n$ ,  $x$  is  $n \times 1$  and is the true deviation away from the nominal trajectory, and  $\epsilon$  is  $m \times 1$  with

$$E[\epsilon] = 0 \quad E[\epsilon\epsilon^T] = R$$

Assume  $m < n$ .

- (a) Derive the equation for the Minimum Norm estimate of  $x$ .
- (b) What is the expected value of the estimate derived in (a)?
- (c) Is the Minimum Norm estimate an unbiased estimate? Explain.
- (d) Derive the equation for the covariance of this estimate.
- (e) Given the linearized data equation

$$y(t) = x_1 + tx_2 + t^2x_3 + \epsilon(t)$$

and one observation  $y=42$  at  $t=2$ , calculate the Minimum Norm estimate of

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$