



ORBIT DETERMINATION

ASEN 5070

Fall 2011

LECTURE 9

9/12/2011

Supplemental Reading:

Sections 4.1 – 4.4

Concept Test

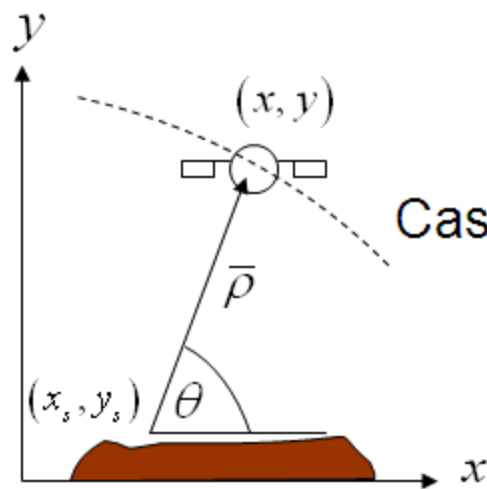


True or False

1) If the relationship between the observations and the state is linear we do not have to iterate the Newton-Raphson equation. _____

2) Given observations of range rate $\dot{\rho}_j$, $j=1\text{---}5$, all elements of the following state vectors can be solved for (indicate T or F for case A and B)

$$\dot{\rho}_j = \frac{1}{\rho_j} \left[\left(\mathbf{x}_0 - \mathbf{x}_s + \dot{\mathbf{x}}_0 t_j \right) \dot{\mathbf{x}}_0 + \left(y_0 - y_s + \dot{y}_0 t_j - \frac{gt_j^2}{2} \right) \left(\dot{y}_0 - g t_j \right) \right]$$



Case A: $X = \begin{bmatrix} x_0 \\ \dot{x}_0 \\ y_0 \\ \dot{y}_0 \\ y_s \end{bmatrix}$ T or F ____.

Case B: $X = \begin{bmatrix} x_0 \\ \dot{x}_0 \\ \dot{y}_0 \\ y_s \\ g \end{bmatrix}$ T or F ____.



Concept Test

- 3) The state vector in case 2.B could be solved for uniquely with one observation each of $\rho, \dot{\rho}, \theta, \dot{\theta}$ at one instant in time. _____
- 4) For problem 2.B we may use any initial guess for the state but it may take many iterations to converge. _____

5) Given:

$$\ddot{\theta} + \omega^2 \theta = 0 \quad X_0 = \begin{bmatrix} \theta_0 \\ \dot{\theta}_0 \end{bmatrix} \quad Y(t) = \begin{bmatrix} \theta_0(t) \\ \dot{\theta}_0(t) \end{bmatrix} + \varepsilon(t)$$

Since the observation-state and state propagation equations are linear we do not have to use a state deviation vector. _____

Concept Test



- 6) The differential equation in each column of $\dot{\Phi} = A\Phi$ is independent of the equations in other columns. _____
- 7) The least squares solution minimizes the sum of the residuals. _____
- 8) For the equation $y = Hx + \varepsilon$ we always have more unknowns than equations. _____
- 9) If the determinant of a symmetric matrix is negative (answer T or F)
- a) It is not positive definite _____
 - b) It's inverse does not exist. _____
 - c) Some eigenvalues are imaginary. _____
- 10) The derivative of a scalar WRT a vector is a scalar. _____



Concept Test

11) The rank of $H = \begin{bmatrix} 2 & 3 & 6 \\ 4 & 6 & 12 \end{bmatrix}$ is 2. _____

Variable
\tilde{H}
ϵ
ρ
y
θ
Y
\hat{X}_k
$G(X, t)$
$\dot{\rho}$
x
$\Phi(t, t_0)$
$\dot{\theta}$
X_k
X^*

Alternate Description of Φ using Taylor Series Expansion



Assume that we can write the solution for $X^*(t)$ based on initial conditions X_0^*

$$X^*(t) = F(X_0^*, t)$$

Expand the true solution about $X^*(t) = F(X_0^*, t)$ and retain 1st order terms

$$X(t) = X^*(t) + \frac{\partial X^*(t)}{\partial X_0} (X(t_0) - X^*(t_0)) + \dots$$

$$X(t) - X^*(t) = \frac{\partial X^*(t)}{\partial X_0} (X(t_0) - X^*(t_0))$$

Define $\chi(t) = X(t) - X^*(t)$, then

$$\chi(t) = \frac{\partial X^*(t)}{\partial X_0} \chi(t_0) = \Phi(t, t_0) \chi_0$$

Alternate Description of Φ using Taylor Series Expansion



For example, assume that

$$\mathbf{X}(t) = \begin{bmatrix} x \\ y \\ \dot{x} \\ \dot{y} \\ \alpha \end{bmatrix}$$

then $\Phi(t, t_0) = \frac{\partial \mathbf{X}(t)}{\partial \mathbf{X}(t_0)} = \frac{\partial \mathbf{X}(t)}{\partial \mathbf{X}_0}$

α is a constant

and

$$\Phi(t, t_0) = \begin{bmatrix} \frac{\partial x(t)}{\partial x_0} & \frac{\partial x(t)}{\partial y_0} & \frac{\partial x(t)}{\partial \dot{x}_0} & \frac{\partial x(t)}{\partial \dot{y}_0} & \frac{\partial x(t)}{\partial \alpha} \\ \frac{\partial y(t)}{\partial x_0} & \frac{\partial y(t)}{\partial y_0} & \dots & \dots & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \frac{\partial \alpha}{\partial x_0} & \frac{\partial \alpha}{\partial y_0} & \dots & \dots & \frac{\partial \alpha}{\partial \alpha} \end{bmatrix}$$

*Note that
the last row
will be

$$[0 \ 0 \ 0 \ 0 \ 1]$$

State Transition Matrix Differential Equation for a General State Vector



Let β represent a $j \times 1$ vector of force model parameters and γ a $k \times 1$ vector of measurement model parameters. β and γ are constants.

$$X = \begin{bmatrix} \bar{r} \\ \bar{v} \\ \bar{\beta} \\ \bar{\gamma} \end{bmatrix} \quad \dot{X} = F(X, t) = \begin{bmatrix} \dot{\bar{r}} \\ \dot{\bar{v}} \\ \dot{\bar{\beta}} = 0 \\ \dot{\bar{\gamma}} = 0 \end{bmatrix}$$

$$A = \frac{\partial F(X, t)}{\partial X} = \begin{bmatrix} \left[\frac{\partial \dot{\bar{r}}}{\partial \bar{r}} \right]_{3 \times 3} = 0 & \left[\frac{\partial \dot{\bar{r}}}{\partial \bar{v}} \right]_{3 \times 3} = I & \left[\frac{\partial \dot{\bar{r}}}{\partial \bar{\beta}} \right]_{3 \times j} = 0 & \left[\frac{\partial \dot{\bar{r}}}{\partial \bar{\gamma}} \right]_{3 \times k} = 0 \\ \left[\frac{\partial \dot{\bar{v}}}{\partial \bar{r}} \right]_{3 \times 3} & \left[\frac{\partial \dot{\bar{v}}}{\partial \bar{v}} \right]_{3 \times 3} & \left[\frac{\partial \dot{\bar{v}}}{\partial \bar{\beta}} \right]_{3 \times j} & \left[\frac{\partial \dot{\bar{v}}}{\partial \bar{\gamma}} \right]_{3 \times k} = 0 \\ \left[\frac{\partial \dot{\bar{\beta}}}{\partial \bar{r}} \right]_{j \times 3} = 0 & \left[\frac{\partial \dot{\bar{\beta}}}{\partial \bar{v}} \right]_{j \times 3} = 0 & \left[\frac{\partial \dot{\bar{\beta}}}{\partial \bar{\beta}} \right]_{j \times j} = 0 & \left[\frac{\partial \dot{\bar{\beta}}}{\partial \bar{\gamma}} \right]_{j \times k} = 0 \\ \left[\frac{\partial \dot{\bar{\gamma}}}{\partial \bar{r}} \right]_{k \times 3} = 0 & \left[\frac{\partial \dot{\bar{\gamma}}}{\partial \bar{v}} \right]_{k \times 3} = 0 & \left[\frac{\partial \dot{\bar{\gamma}}}{\partial \bar{\beta}} \right]_{k \times j} = 0 & \left[\frac{\partial \dot{\bar{\gamma}}}{\partial \bar{\gamma}} \right]_{k \times k} = 0 \end{bmatrix}_{(6+j+k) \times (6+j+k)}$$

State Transition Matrix Differential Equation for a General State Vector



Hence, $A(t)$ may be written as:

$$A = \begin{bmatrix} [0]_{3 \times 3} & [I]_{3 \times 3} & [0]_{3 \times j} & [0]_{3 \times k} \\ \left[\frac{\partial \ddot{\mathbf{r}}}{\partial \mathbf{r}} \right]_{3 \times 3} & \left[\frac{\partial \ddot{\mathbf{r}}}{\partial \mathbf{v}} \right]_{3 \times 3} & \left[\frac{\partial \ddot{\mathbf{r}}}{\partial \bar{\boldsymbol{\beta}}} \right]_{3 \times j} & [0]_{3 \times k} \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & I & 0 & 0 \\ A_{21} & A_{22} & A_{23} & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}_{(6+j+k) \times (6+j+k)}$$

State Transition Matrix Differential Equation for a General State Vector



Then $\dot{\Phi} = A\Phi$ yields

$$\dot{\Phi} = \begin{bmatrix} 0 & I & 0 & 0 \\ A_{21} & A_{22} & A_{23} & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \phi_{11} & \phi_{12} & \phi_{13} & 0 \\ \phi_{21} & \phi_{22} & \phi_{23} & 0 \\ 0 & 0 & I & 0 \\ 0 & 0 & 0 & I \end{bmatrix}_{(6+j+k) \times (6+j+k)}$$

$$= \begin{bmatrix} [\phi_{21}]_{3 \times 3} & [\phi_{22}]_{3 \times 3} & [\phi_{23}]_{3 \times k} & [0]_{3 \times k} \\ [A_{21}\phi_{11} + A_{22}\phi_{21}]_{3 \times 3} & [A_{21}\phi_{12} + A_{22}\phi_{22}]_{3 \times 3} & [A_{21}\phi_{13} + A_{22}\phi_{23} + A_{23}]_{3 \times j} & [0]_{3 \times k} \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

State Transition Matrix Differential Equation for a General State Vector



Hence, we need only to integrate the $6 \times (6 + j)$ matrix of differential equations,

$$\dot{\Phi}_1 = \begin{bmatrix} \phi_{21} & \phi_{22} & \phi_{23} \\ [A_{21}\phi_{11} + A_{22}\phi_{21}] & [A_{21}\phi_{12} + A_{22}\phi_{22}] & [A_{21}\phi_{13} + A_{22}\phi_{23} + A_{23}] \end{bmatrix}_{6 \times (6+j)}$$

within I.C.

$$\Phi_1(t_0, t_0) = \begin{bmatrix} [I]_{6 \times 6} & [0]_{6 \times j} \end{bmatrix}$$

The remaining elements of Φ simply are the elements of an identity matrix.

Symplectic Property of Φ



Under certain conditions on $A(t)$ the state transition matrix may be inverted analytically (Battin, 1987). Under these conditions Φ is referred to as being symplectic.

If the matrix $A(t)$ can be partitioned in the form

$$A(t) = \begin{bmatrix} A_1 & A_2 \\ A_3 & A_4 \end{bmatrix} \quad (4.2.12)$$

where the submatrices have the properties that

$$A_1^T = -A_4, \quad A_2^T = A_2, \text{ and } A_3^T = A_3 \quad (4.2.13)$$

Symplectic Property of Φ



Then $\Phi(t, t_k)$ can be similarly partitioned as

$$\Phi(t, t_k) = \begin{bmatrix} \Phi_1 & \Phi_2 \\ \Phi_3 & \Phi_4 \end{bmatrix}$$

and $\Phi^{-1}(t, t_k)$ may be written as

$$\Phi^{-1}(t, t_k) = \begin{bmatrix} \Phi_4^T & -\Phi_2^T \\ -\Phi_3^T & \Phi_1^T \end{bmatrix} \quad (4.2.14)$$

If $\ddot{\vec{r}} = \nabla U$ then Eq. (4.2.13) is true

Symplectic Property of Φ



In this case (consider a 2-D case for simplicity)

$$\begin{aligned} \ddot{x} &= \frac{\partial U}{\partial x} \\ \ddot{y} &= \frac{\partial U}{\partial y} \end{aligned} \quad X = \begin{bmatrix} x \\ y \\ u \\ v \end{bmatrix} \quad \longrightarrow$$

$$\dot{x} = u \quad \dot{y} = v$$

$$\dot{u} = \frac{\partial U}{\partial x} \quad \dot{v} = \frac{\partial U}{\partial y}$$

$$A = \frac{\partial F}{\partial X} = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ \hline \frac{\partial^2 U}{\partial x^2} & \frac{\partial^2 U}{\partial x \partial y} & 0 & 0 \\ \frac{\partial^2 U}{\partial y \partial x} & \frac{\partial^2 U}{\partial y^2} & 0 & 0 \end{bmatrix}$$

$$\text{and } A_1^T = -A_4,$$

$$A_2^T = A_2, \quad A_3^T = A_3$$

$$\text{Because } \frac{\partial^2 U}{\partial x \partial y} = \frac{\partial^2 U}{\partial y \partial x},$$

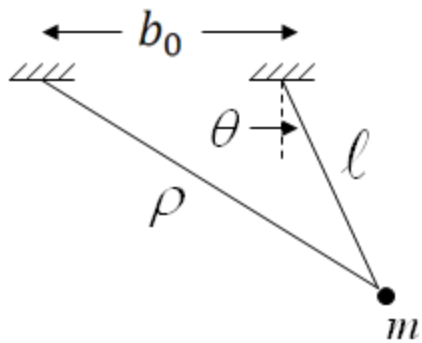
Φ is symplectic

Example, Problem 14, Chapter 4 of Text



14a) Generate the $A(t)$ and \tilde{H} matrix for the pendulum problem. Assume that we wish to estimate θ , $\dot{\theta}$, and b_0 at some epoch time.

Derive the equations of motion from the free body diagram.

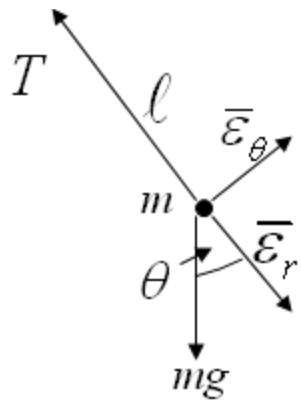


$$\bar{\mathbf{r}} = \ell \bar{\mathbf{e}}_r \quad \dot{\bar{\mathbf{r}}} = \ell \dot{\bar{\mathbf{e}}}_r = \ell \dot{\theta} \bar{\mathbf{e}}_\theta$$

$$\ddot{\bar{\mathbf{r}}} = \ell \ddot{\theta} \bar{\mathbf{e}}_\theta - \ell \dot{\theta}^2 \bar{\mathbf{e}}_r$$

$$m \ddot{\bar{\mathbf{r}}} = \sum \bar{\mathbf{F}}$$

$$m \ddot{\bar{\mathbf{r}}} = mg \cos \theta \bar{\mathbf{e}}_r - T \bar{\mathbf{e}}_r - mg \sin \theta \bar{\mathbf{e}}_\theta$$



$$m(\ell \ddot{\theta} \bar{\mathbf{e}}_\theta - \ell \dot{\theta}^2 \bar{\mathbf{e}}_r) = (mg \cos \theta - T) \bar{\mathbf{e}}_r - mg \sin \theta \bar{\mathbf{e}}_\theta$$

Example, Problem 14, Chapter 4 of Text



In component form: $\bar{\varepsilon}_r \quad -m\ell \dot{\theta}^2 = mg \cos \theta - T \quad (1)$

$$\bar{\varepsilon}_\theta \quad m\ell \ddot{\theta} = -mg \sin \theta \quad (2)$$

Eq. (2) gives us $\theta(t)$, $\dot{\theta}(t)$ and Eq. (1) gives the tension in the cord,

$$T = mg \cos \theta + m\ell \dot{\theta}^2$$

Hence, we need to solve $\ddot{\theta} = \frac{-g}{\ell} \sin \theta$, $\omega \equiv \sqrt{\frac{g}{\ell}}$
 $= -\omega^2 \sin \theta$

writing Eqs. in 1st order form:

$$\begin{aligned} \dot{\theta} &= \alpha \\ \dot{\alpha} &= -\omega^2 \sin \theta \\ \dot{b}_0 &= 0 \end{aligned} \quad \text{Then,} \quad X = \begin{bmatrix} \theta \\ \alpha \\ b_0 \end{bmatrix}, \dot{X} = \begin{bmatrix} \dot{\theta} \\ \dot{\alpha} \\ \dot{b}_0 \end{bmatrix} = \begin{bmatrix} \alpha \\ -\omega^2 \sin \theta \\ 0 \end{bmatrix}$$

Example, Problem 14, Chapter 4 of Text



$$A(t) = \frac{\partial \dot{X}(t)}{\partial X(t)} = \begin{bmatrix} \frac{\partial \alpha}{\partial \theta} & \frac{\partial \alpha}{\partial \alpha} & \frac{\partial \alpha}{\partial b_0} \\ \frac{\partial \dot{\alpha}}{\partial \theta} & \frac{\partial \dot{\alpha}}{\partial \alpha} & \frac{\partial \dot{\alpha}}{\partial b_0} \\ \frac{\partial \dot{b}_0}{\partial \theta} & \frac{\partial \dot{b}_0}{\partial \alpha} & \frac{\partial \dot{b}_0}{\partial b_0} \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ -\omega^2 \cos \theta & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\dot{\Phi}(t, t_o) = A(t) \Phi(t, t_o) = \begin{bmatrix} \dot{\phi}_{11} & \dot{\phi}_{12} \\ \dot{\phi}_{21} & \dot{\phi}_{22} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -\omega^2 \cos \theta & 0 \end{bmatrix}^* \begin{bmatrix} \phi_{11} & \phi_{12} \\ \phi_{21} & \phi_{22} \end{bmatrix}$$

Can we use Laplace Transforms to solve for $\Phi(t, t_o)$?

Where $[\]^*$ indicates that $A(t)$ is evaluated on a reference solution for $\theta(t)$.

Example, Problem 14, Chapter 4 of Text



Choose initial conditions θ_0^* , $\dot{\theta}_0^*$ and generate the reference trajectory while simultaneously integrating $\dot{\Phi} = A(t)\Phi$ i.e.

$$\dot{\theta} = \alpha$$

$$\dot{\phi}_{12} = \phi_{22}$$

$$\dot{\alpha} = -\omega^2 \sin \theta$$

$$\dot{\phi}_{21} = -\omega^2 \cos(\theta) \phi_{11}$$

$$\dot{\phi}_{11} = \phi_{21}$$

$$\dot{\phi}_{22} = -\omega^2 \cos(\theta) \phi_{12}$$

Note that we do not need to integrate equations for b_0 since it is a constant

Example, Problem 14, Chapter 4 of Text



$$\text{IC: } \theta^*(t_0), \dot{\theta}^*(t_0), \Phi(t_0, t_0) = \mathbf{I}$$

$$\text{i.e., } \phi_{11}(t_0, t_0) = \phi_{22}(t_0, t_0) = 1, \phi_{12}(t_0, t_0) = \phi_{21}(t_0, t_0) = 0$$

To do this in Matlab we would use the Reshape command. Which would write a matrix as a vector and vice versa. (see hints under handouts on web – “Matlab help for solving problem 4.10”). The vector derivatives are:

$$\begin{bmatrix} \dot{\theta} \\ \dot{\alpha} \\ \dot{\phi}_{11} \\ \dot{\phi}_{12} \\ \dot{\phi}_{21} \\ \dot{\phi}_{22} \end{bmatrix}$$

Example, Problem 14, Chapter 4 of Text



Compute \tilde{H}

From the law of cosines

$$\begin{aligned}\rho^2 &= b_0^2 + l^2 - 2b_0l\cos(90^\circ + \theta) \\ &= b_0^2 + l^2 + 2b_0l\sin\theta \\ &= \frac{1}{\rho} [b_0l\cos\theta \quad 0 \quad b_0 + l\sin\theta]\end{aligned}$$

$$\text{Hence, } \tilde{H} = \frac{\partial \rho}{\partial x} = \begin{bmatrix} \frac{\partial \rho}{\partial \theta} & \frac{\partial \rho}{\partial \alpha} & \frac{\partial \rho}{\partial b_0} \end{bmatrix}$$

\tilde{H} is evaluated on the reference solution

Example, Problem 14, Chapter 4 of Text



The observations are related to a reference state deviation vector by,

$$y(t_1) = \tilde{H}(t_1)\Phi(t_1, t_0)x_0 + \varepsilon_1$$

$$y(t_2) = \tilde{H}(t_2)\Phi(t_2, t_0)x_0 + \varepsilon_2$$

$$y(t_m) = \tilde{H}(t_m)\Phi(t_m, t_0)x_0 + \varepsilon_m$$

Defining

$$y = \begin{bmatrix} y(t_1) \\ \vdots \\ y(t_m) \end{bmatrix}, \quad H = \begin{bmatrix} \tilde{H}(t_1)\Phi(t_1, t_0) \\ \vdots \\ \tilde{H}(t_m)\Phi(t_m, t_0) \end{bmatrix}, \quad \varepsilon = \begin{bmatrix} \varepsilon_1 \\ \vdots \\ \varepsilon_m \end{bmatrix}$$

Then $y = Hx_0 + \varepsilon$ and $\hat{x}_0 = (H^T H)^{-1} H^T y$

$$= \left(\sum_{i=1}^m H_i^T H_i \right)^{-1} \sum_{i=1}^m H_i^T y_i$$

Example, Problem 14, Chapter 4 of Text



Here

$$x_0 = (X_0 - X_0^*) = \begin{bmatrix} \theta_0 - \theta_0^* \\ \alpha_0 - \alpha_0^* \\ b_0 - b_0^* \end{bmatrix} \text{ at reference time, } t_0$$

and $y(t_i) = \rho(t_i)_{\text{observed}} - \rho(t_i)_{\text{computed}}$

14 b) Assume small oscillations, i.e., $\sin \theta = \theta$, $\cos \theta = 1$. Then the equations of motion become

$$\ddot{\theta} + \frac{g}{l} \theta = 0$$

$$\ddot{\theta} + \omega^2 \theta = 0$$

Define

$$\omega \equiv \sqrt{g/l}$$

Example, Problem 14, Chapter 4 of Text



This is the equation for a harmonic oscillator which has the solution

$$\theta(t) = A \cos \omega t + B \sin \omega t$$

The constants are evaluated by noting that $t = t_0 = 0$, $\theta = \theta_0$, $\dot{\theta} = \dot{\theta}_0$

Hence,

$$\theta(t) = \theta_0 \cos \omega t + \frac{\dot{\theta}_0}{\omega} \sin \omega t$$

$$\dot{\theta}(t) = \alpha(t) = -\theta_0 \omega \sin \omega t + \dot{\theta}_0 \cos \omega t$$

Example, Problem 14, Chapter 4 of Text



We may now write the state transition matrix directly by differentiating the solution for $\theta(t)$ and $\alpha(t)$, i.e.,

$$\Phi(t, t_0 = 0) = \frac{\partial X(t)}{\partial X(t_0)}$$
$$= \begin{bmatrix} \cos \omega t & \frac{1}{\omega} \sin \omega t & 0 \\ -\omega \sin \omega t & \cos \omega t & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Example, Problem 14, Chapter 4 of Text



Alternatively write $\ddot{\theta} = -\omega^2 \theta$ as a 1st order system

$$\begin{aligned}\dot{\theta} &= \alpha \\ \dot{\alpha} &= -\omega^2 \theta\end{aligned}\quad \begin{bmatrix} \dot{\theta} \\ \dot{\alpha} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -\omega^2 & 0 \end{bmatrix} \begin{bmatrix} \theta \\ \alpha \end{bmatrix}$$

$$(SI - A)^{-1} = \begin{bmatrix} s & -1 \\ \omega^2 & s \end{bmatrix}^{-1} = \frac{1}{s^2 + \omega^2} \begin{bmatrix} s & 1 \\ -\omega^2 & s \end{bmatrix}$$

Taking the inverse Laplace Transform gives us

$$= \begin{bmatrix} \cos \omega t & \frac{1}{\omega} \sin \omega t & 0 \\ -\omega \sin \omega t & \cos \omega t & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

14c) The assumption that θ is small restricts this solution to small values of θ . However, if we linearize $\theta(t)$ about a reference solution, we do not require that $\theta(t)$ be small, only that the deviation of $\theta(t)$ from the reference, $\theta^*(t)$, be small.

Least Squares with apriori Information



If an apriori value is available for x_k (call it \bar{x}_k) and an associated symmetric weighting matrix \bar{w}_k , the weighted least squares estimate of \hat{x}_k can be obtained.



Least Squares with apriori Information

Given

$$y = Hx_k + \varepsilon$$

$$\bar{x}_k = x_k + \eta_k$$

Where η_k is the error in \bar{x}_k and its influence on \hat{x}_k is reflected in the weighting matrix \bar{w}_k

and $\begin{bmatrix} y \end{bmatrix}_{m \times 1}$, $\begin{bmatrix} \bar{x}_k \end{bmatrix}_{n \times 1}$

Choose \hat{x}_k to minimize the performance index

$$J(x_k) = \frac{1}{2} \varepsilon^T w \varepsilon + \frac{1}{2} \eta_k \bar{w}_k \eta_k^T$$

Least Squares with apriori Information



Writing $J(x_k)$ explicitly in terms of x_k

$$J(x_k) = \frac{1}{2}(y - Hx_k)^T w (y - Hx_k) + \frac{1}{2}(\bar{x}_k - x_k)^T \bar{w}_k (\bar{x}_k - x_k) \quad (4.3.24)$$

$$\frac{\partial J(x_k)}{\partial x_k} = 0$$

Results in (See Eq B.7.4)

$$\frac{\partial J(x_k)}{\partial x_k} = -(y - Hx_k)^T w H - (\bar{x}_k - x_k)^T w_k = 0$$

$$= -y^T w H + x_k^T H^T w H - \bar{x}_k^T w_k + x_k^T w_k = 0$$

Least Squares with apriori Information



Solving for x_k yields \hat{x}_k

$$x_k^T (H^T w H + w_k) = y^T w H + \bar{x}_k^T w_k$$

$$\hat{x}_k^T = (y^T w H + \bar{x}_k^T w_k) (H^T w H + w_k)^{-1}$$

$$\hat{x}_k = (H^T w H + w_k)^{-1} (H^T w y + w_k \bar{x}_k) \quad (4.3.25)$$

Least Squares with apriori Information



Note that $\left(H^T w H + w_k\right)^{-1}$ is symmetric

also

$$\frac{\partial^2 J(x_k)}{\partial x_k^2} = H^T w H + w_k$$

which will be positive definite if H and/or w_k is full rank.
Hence, \hat{x}_k minimizes $J(x_k)$.