

Given:

$$y = Ax + e$$

- A is known
- e is measurement noise with Gaussian distribution, mean of 0 and covariance R .

$$e \sim N(0, R)$$

Find:

a) $f_{y|x}(y|x)$

b) $\hat{x}_{ML} = \arg \max_x f_{y|x}(y|x)$. Find this in 2 ways

- Computing gradients

- Completing the square

c) Using Bayes' rule, write down $f_{x|y}(x|y)$

d) Show that

$$\hat{x}_{MAP} = \arg \max_x f_{x|y}(x|y) = \arg \max_x f_{y|x}(y|x) f_x(x)$$

e) Perform maximization of \hat{x}_{MAP} using gradients or completing the square

f) Show

$$\hat{x}_{MAP} = x_0 + (\text{some matrix})^{-1} (\text{some vector})$$

Solution:

a)
$$f_{y|x}(y|x) = \frac{1}{(2\pi)^{k/2} |R|^{1/2}} \exp \left[-\frac{1}{2} (y - Ax)^T R^{-1} (y - Ax) \right]$$

b) Gradient

$$\arg \max_x \underbrace{\frac{1}{(2\pi)^{k/2} |R|^{1/2}}} \exp \left[-\frac{1}{2} (y - Ax)^T R^{-1} (y - Ax) \right]$$

Just a constant

$$\arg \min_x (y - Ax)^T R^{-1} (y - Ax)$$

$$\arg \min_x y^T R^{-1} y - y^T R^{-1} A x - x^T A^T y + x^T A^T R^{-1} A x$$

$$\arg \min_x . y^T R^{-1} y - 2x^T A^T R^{-1} y + x^T A^T R^{-1} A x = \tilde{A}$$

$$\frac{\partial \tilde{A}}{\partial x} = 0 - 2A^T R^{-1} y + 2A^T R^{-1} A x$$

Find where gradient derivative is 0

$$A^T R^{-1} y = A^T R^{-1} A x$$

$$x = (A^T R^{-1} A)^{-1} (A^T R^{-1}) y$$



Complete square

Starting with

$$(x - 1/2 A^{-1} b)^T A (x - 1/2 A^{-1} b) + c - 1/4 b^T A^{-1} b$$

We have most of the square values, except b. To find that expand

$$(y - Ax)^T R^{-1} (y - Ax)$$

to get

$$y^T R^{-1} y - 2y^T R^{-1} A x + \underbrace{w^T A^T Q^{-1} A w}_{b^T} \quad \underbrace{w^T A^T Q^{-1} A w}_{A}$$

$$x = \frac{1}{2} (A^T Q^{-1} A)^{-1} (2A^T Q^{-1} y)$$



c)

$$\begin{aligned}
 f_{x|y}(x|y) &= \frac{f(y|x)f(x)}{f(y)} \\
 &= \frac{\exp\left[-\frac{1}{2}(y-Ax)^T Q^{-1}(y-Ax)\right] \exp\left[-\frac{1}{2}(x-x_0)^T Q^{-1}(x-x_0)\right]}{\exp\left[-\frac{1}{2}(y-\mu_y)^T R_y^{-1}(y-\mu_y)\right]}
 \end{aligned}$$

d) When performing an optimization, we want to maximize the numerator to obtain the largest value.
 Therefore it is safe to state

$$\arg \max_x f_{x|y} = \arg \max_x f(y|x) f(x)$$

e)

$$\arg \max_x [-(y-Ax)^T Q^{-1}(y-Ax)] + [(x-x_0)^T Q^{-1}(x-x_0)] = \tilde{A}$$

$$(y^T Q^{-1} y - 2x^T A^T Q^{-1} y + x^T A^T Q^{-1} A x) +$$

$$(x^T Q^{-1} x - 2x_0^T Q^{-1} x + x_0^T Q^{-1} x_0) = \tilde{A}$$

$$\frac{\partial \tilde{A}}{\partial x} = (0 - 2A^T Q^{-1} y + 2A^T Q^{-1} A x) + (-2Q^{-1} x - 2x_0^T Q^{-1} + 0)$$

$$-2A^T Q^{-1} y + 2A^T Q^{-1} Ax + 2Q^{-1} x - 2x_0^T Q^{-1}$$

$$x = (2A^T Q^{-1} A + 2Q^{-1})^{-1} (Qx_0^T Q^{-1} - 2A^T Q^{-1} y) //$$

$$d) (A^T Q^{-1} A + Q^{-1})^{-1} (x_0^T Q^{-1} - A^T Q^{-1} y)$$

$$(A^T Q^{-1} A)^{-1} (A^T Q y^{-1} + x_0^T Q + y^{-1}) + x_0 //$$



2)

Given:

$$\begin{bmatrix} \bar{r}_0 & \bar{r}_1 & \dots & \bar{r}_M & \bar{r}_N \\ \bar{r}_1 & \bar{r}_0 & \dots & \bar{r}_{M-1} & \bar{r}_N \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ \bar{r}_N & \bar{r}_{M-1} & \dots & \bar{r}_1 & \bar{r}_0 \end{bmatrix} \begin{bmatrix} w_0 \\ w_1 \\ \vdots \\ w_N \end{bmatrix} = \begin{bmatrix} \bar{r}_1 \\ \bar{r}_2 \\ \vdots \\ \bar{r}_M \end{bmatrix}$$

- Where the overbar defines the complex conjugate

- The matrix is complex Hermitian symmetric Toeplitz

Find:

- a) Derive the "fast" algorithm for solving the system of equations
 b) Code algorithm in MATLAB and verify it works.

Solution:

a)

To begin, show that given the Hermitian matrix, show that it is persymmetric

Def: A matrix is persymmetric if, for an $n \times n$ matrix

$$a_{ij} = a_{n-i+1, n-j+1} \neq ij$$

To show a matrix is persymmetric

Def: Let $J = \begin{bmatrix} [0] & 1 & & \\ & \ddots & & \\ & & [0] & \\ 1 & & & \end{bmatrix}_{n \times n}$ A matrix T is persymmetric if

$$T = JT^TJ$$

$$T^T = \begin{bmatrix} R_0 & \bar{R}_1 & \bar{R}_2 & \cdots & \bar{R}_{m-1} & \bar{R}_m \\ R_1 & R_0 & \bar{R}_1 & \cdots & \bar{R}_{m-2} & \bar{R}_m \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ R_m & \bar{R}_{m-1} & \bar{R}_m & \cdots & R_0 & R_1 \end{bmatrix}$$

$$JT^T = \begin{bmatrix} R_m & R_{m-1} & R_m & \cdots & R_1 & R_0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ R_1 & R_0 & \bar{R}_1 & \cdots & \bar{R}_{m-2} & \bar{R}_m \\ R_0 & \bar{R}_1 & \bar{R}_0 & \cdots & \bar{R}_{m-1} & \bar{R}_m \end{bmatrix} = T$$

$$\tilde{T}J = \begin{bmatrix} R_0 & R_1 & R_2 & \cdots & R_{m-1} & R_m \\ R_1 & R_0 & R_1 & \cdots & R_{m-2} & R_{m-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ R_m & R_{m-1} & R_m & \cdots & R_1 & R_0 \end{bmatrix} = T$$

//

From the notes we have the fact about persymmetric matrices.

Fact: if T is persymmetric and J is the counter-identity, then

$$JT^{-1} = T^{-T} J$$

//

Now, the moment we have all been waiting for!

Suppose we know the solution to

$$r_n = T_n w_n$$

and we want to find the solution to

$$T_{n+1} \underline{w}_{n+1} = \underline{r}_{n+1}$$

Let $\underline{w}_n = \begin{bmatrix} \underline{z}_n \\ \alpha_n \end{bmatrix}$ where \underline{z}_n is a vector and α_n is a scalar

$$T_{n+1} = \begin{bmatrix} T_n & \underline{r}_n \\ \vdots & \vdots \\ \underline{r}_n & \underline{r}_n \dots \underline{r}_1 & r_0 \end{bmatrix} = \begin{bmatrix} [T_n] & [J_n \underline{s}_0] \\ [\underline{r}_n^T J] & r_0 \end{bmatrix}$$

$$\begin{bmatrix} [T_n] & [J_n \underline{s}_0] \\ [\underline{r}_n^T J] & r_0 \end{bmatrix} \begin{bmatrix} \underline{z}_n \\ \alpha_n \end{bmatrix} = \begin{bmatrix} \underline{r}_n \\ \underline{r}_{n+2} \end{bmatrix}$$

$$T_n \underline{z}_n + J_n \underline{s}_0 \alpha_n = \underline{r}_n \quad (1)$$

$$\underline{r}_n^T J \underline{z}_n + r_0 \alpha_n = \underline{r}_{n+2} \quad (2)$$

$$(1) \text{ Solve for } \underline{z}_n : \underline{z}_n = T_n^{-1} (\underline{r}_n - J_n \underline{r}_n \alpha_n)$$

Aside

Because T is
PERSYMMETRIC

$$T^{-1} J = J T^{-1}$$

$$= T_n^{-1} \underline{r}_n - T_n^{-1} J_n \underline{r}_n \alpha_n$$

$$= \underline{w}_n - \alpha_n J_n \underline{w}_n$$

Substitute back into (2) and solve
for α_n

$$\underline{r}_n^T J_n (\underline{w}_n - \alpha_n J_n \underline{w}_n) + r_0 \alpha_n = \underline{r}_{n+2}$$

$$\Rightarrow \alpha_n = \frac{\underline{r}_{n+2} - \underline{r}_n^T J_n \underline{w}_n}{r_0 - \underline{r}_n^T \underline{w}_n} //$$

14-13)

Given,

$$H(z) = \frac{1-5z^{-1}-6z^{-2}}{1-1.5z^{-1}+0.56z^{-2}}$$

Find:

Write in controller canonical form and draw a realization
 Solution:

Begin with Partial Fraction expansion

$$H(z) = \frac{120.71}{z-0.7} - \frac{117.0}{z-0.6}$$

To write $H(z)$ in CCF we write

$$u(k) = 0.56g(k+2) - 1.5g(k+1) + g(k)$$

$$\begin{aligned} g(k+2) &= (1.5g(k+1) - g(k) + u(k)) / 0.56 \\ &= 2.68g(k+1) - 1.79g(k) + 1.79u(k) \end{aligned}$$

$$y(k) = -6g(k-2) - 5g(k-1) + g(k),$$

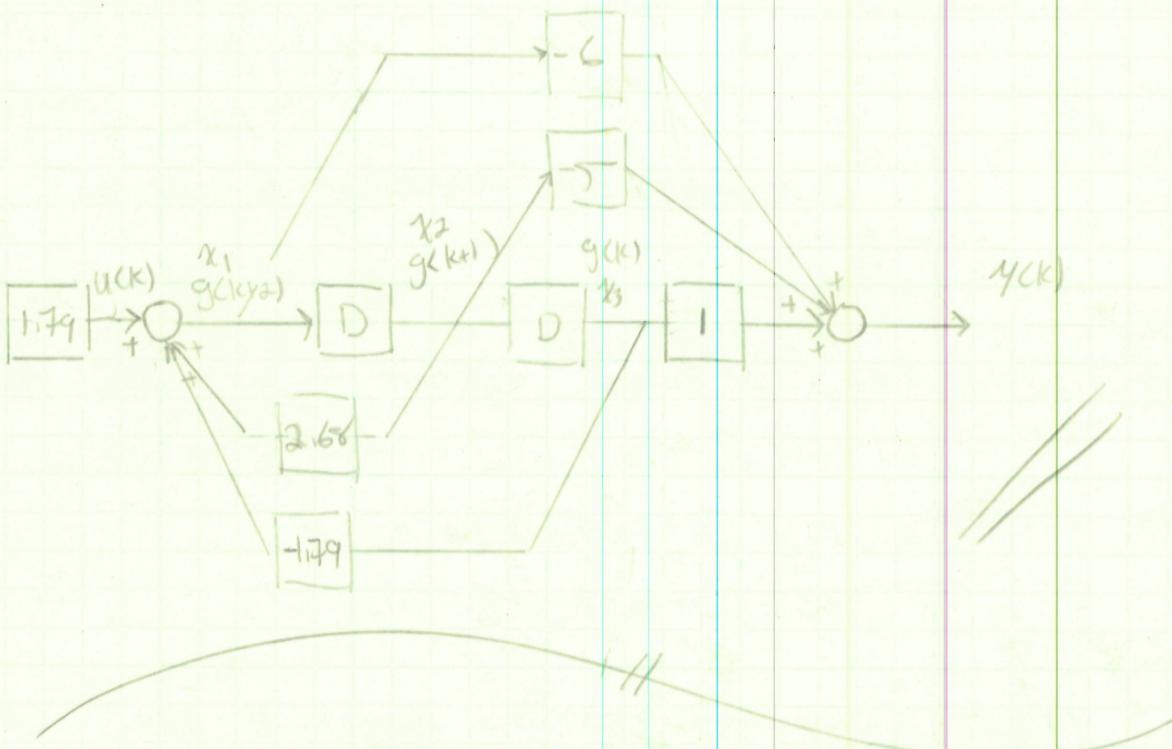
Now to define

$$x(k+1) = Ax(k) + bu(k); y(k) = cx(k)$$

$$\text{Let } x_1[k] = g(k+1) \quad x_2(k+1) = x_1(k) \quad x_3[k+1] = x_2(k)$$

$$x(k+1) = \begin{bmatrix} x_3(k+1) \\ x_2(k+1) \\ x_1(k+1) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & -1.79 & 2.68 \end{bmatrix} \begin{bmatrix} x_3(k) \\ x_2(k) \\ x_1(k) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 1.79 \end{bmatrix} u(k)$$

$$y(k) = [1 \ -5 \ -6] x(k)$$



1.4-14)

Given:

Introduction to observer canonical form

Find:

- a) Show the z -transform implied by (1.2) can be written as

$$Y(z) = b_0 F(z) + [b_1 F(z) - \bar{a}_1 Y(z)] z^{-1} + \\ [\bar{b}_2 F(z) - \bar{a}_2 Y(z)] z^{-2} + \dots \quad (1.81)$$

$$[\bar{b}_p F(z) - \bar{a}_p Y(z)] z^{-p}$$

- b) Draw a block diagram representing (1.81)

- c) Label the outputs of the delay elements from right to left as x_1, x_2, \dots, x_p . Show that the system can be written as

$$A = \begin{bmatrix} -\bar{a}_1 & 1 & 0 & \dots & 0 \\ -\bar{a}_2 & 0 & 1 & \dots & 0 \\ \vdots & & & & \vdots \\ -\bar{a}_{p-1} & 0 & 0 & \dots & 0 \\ -\bar{a}_p & 0 & 0 & \dots & 0 \end{bmatrix}$$

$$b = \begin{bmatrix} b_1 - \bar{a}_1 b_0 \\ b_0 - \bar{a}_2 b_0 \\ \vdots \\ b_p - \bar{a}_p b_0 \end{bmatrix}$$

$$c = \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$d = \bar{b}_0$$

- d) Draw the block diagram in observer canonical form for

$$H(z) = \frac{2+3z^{-1}+4z^{-2}}{1+z^{-1}-6z^{-2}-7z^{-3}}$$

Solution:

$$a) \sum_{k=0}^p \bar{a}_k y[t-k] = \sum_{k=0}^p \bar{b}_k f[t-k] \quad \text{and } \bar{a}_0 = 1$$

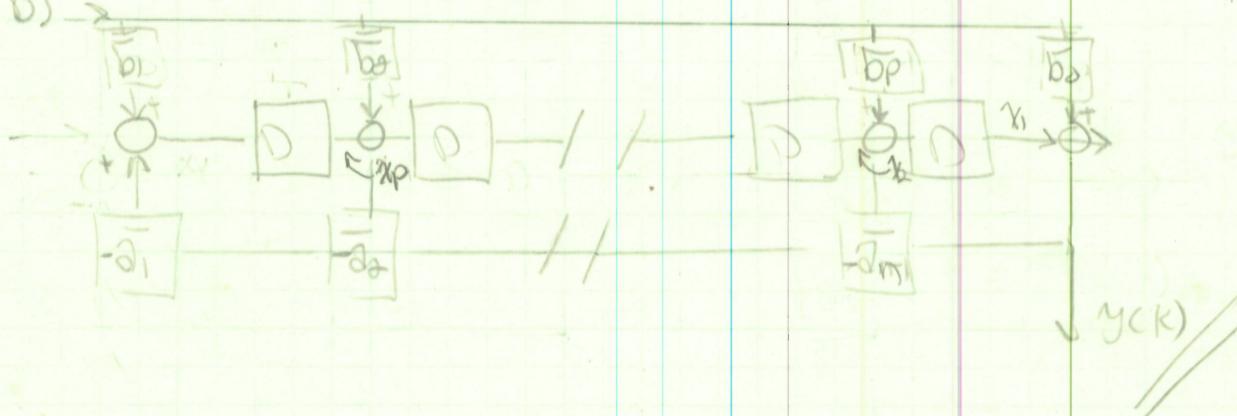
$$y[t-k] + \sum_{k=1}^p \bar{a}_k y[t-k] = \bar{b}_p f[t]^p + \sum_{k=1}^p \bar{b}_k f[t-k]$$

Take \bar{z} transformation

$$Y(z) + \sum_{k=1}^p \bar{a}_k Y(z) z^{-p} = b_p f(z)^p + \sum_{k=1}^p \bar{b}_k f(z) z^{-p}$$

$$\Rightarrow Y(z) = b_p f(z)^p + \sum_{k=1}^{p-1} (\bar{b}_{k+1} f(z) - \bar{a}_k Y(z)) z^{-p}$$

b)



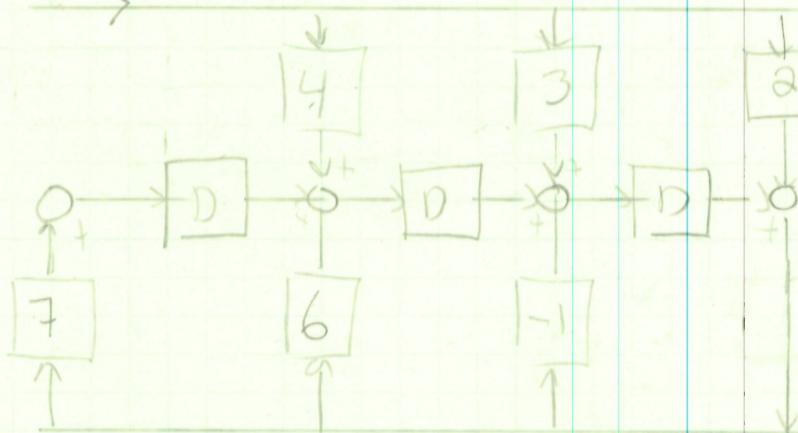
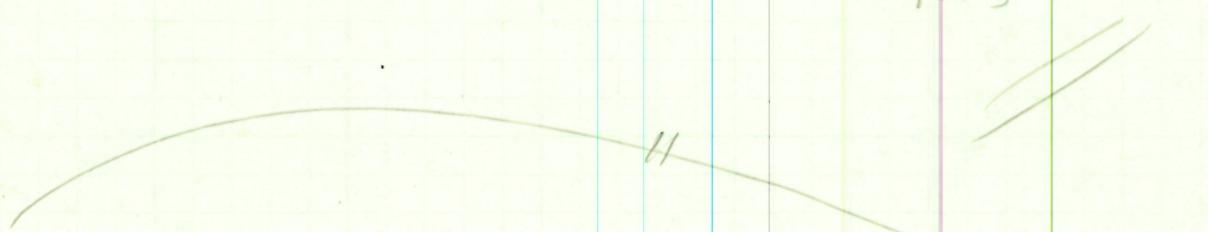
c) Follow diagram to get

$$x_p[k+1] = -\bar{a}_p x_p[k] + \bar{b}_1 u[k] - \bar{a}_p \bar{b}_0 [k]$$

$$\Rightarrow x[k] = [x_1 \ x_2 \ x_3 \ \dots \ x_p]$$

so and

so forth //

d) u  $y[k]$ 

1.4-16)

Given:

$$x[t+1] = Ax[t] + bf[t]$$

$$y[t] = Cx[t] + df[t]$$

(1.21)

$$\bar{x}[t+1] = \bar{A}\bar{x}[t] + \bar{b}f[t]$$

$$y[t] = \bar{C}\bar{x}[t] + \bar{d}f[t]$$

(1.22)

$$\bar{A} = T^{-1}AT \quad \bar{b} = T^{-1}b \quad \bar{C} = T^T C \quad \bar{d} = d \quad \bar{x} = T^{-1}x$$

Find:

$$x = T\bar{x}$$

Show (1.21) and (1.22) have the same transfer function
Solution

$$\hat{G}(z) = C(zI - A)^{-1}B + D$$

where $\hat{G}(z)$ is the transfer function

$$\hat{G}(z) = C(zI - A)^{-1}B + d$$

Now

$$\bar{\hat{G}}(z) = \bar{C}(zI - \bar{A})^{-1}\bar{B} + \bar{d}$$

$$= CT(zI - T^{-1}AT)^{-1}T^{-1}B + d$$

$$= C(T^T zIT^T - T^T T^{-1}AT)^{-1}T^T B + d$$

$$= C((zI + T^T A T^{-1}) - I A I)^{-1}B + d$$

$$= C(zI - A)^{-1}B + d$$

//



1.4 - 18)

Given: θ systems (A_1, b_1, C_1^T) and (A_2, b_2, C_2^T)

Find:

(A, B, C^T) by connecting the systems in

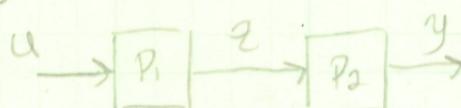
a) Series

b) Parallel

c) A feed forward configuration with (A_1, b_1, C_1^T) and
feedback configuration with (A_2, b_2, C_2^T)

Solution:

a)



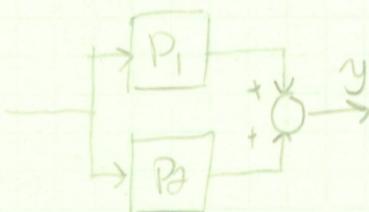
$$\dot{x}_1 = A_1 x_1 + b_1 u \quad \dot{x}_2 = A_2 x_2 + b_2 z$$

$$z = C_1^T x$$

$$y = C_2^T x_2$$

$$\dot{x} = \begin{bmatrix} A_1 & 0 \\ B_2 C_1 & A_2 \end{bmatrix} x + \begin{bmatrix} B_1 \\ B_2 \end{bmatrix} u \quad y = [C_1 \ C_2] x //$$

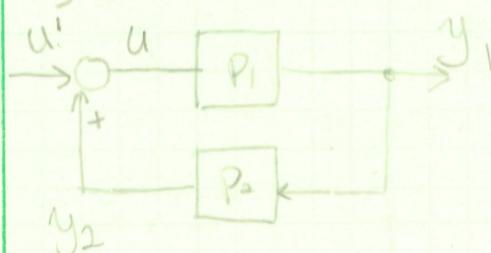
b)



$$\dot{x} = \begin{bmatrix} A_1 & 0 \\ 0 & A_2 \end{bmatrix} x + \begin{bmatrix} B_1 \\ B_2 \end{bmatrix} u$$

$$y = [C_1 \ C_2] x$$

c)



$$u' = u + y_2 \Rightarrow u = u' - y_2$$

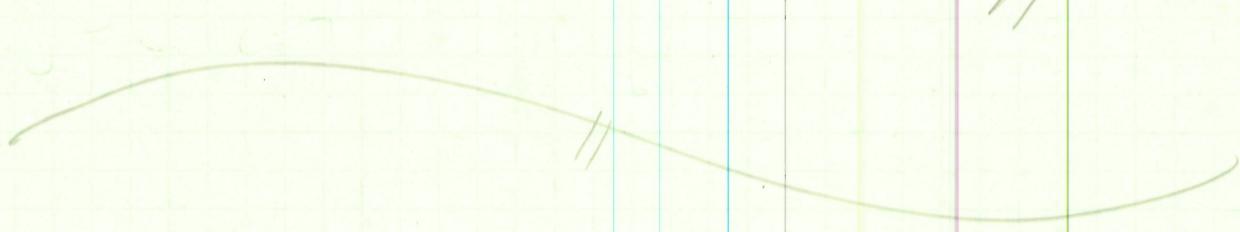
$$\dot{x}_1 = A_1 x_1 + B_1 u = A_1 x_1 + B_1 (u' - y_2) //$$

$$y_1 = C_1^T x_1 //$$

$$\dot{x}_2 = A_2 x_2 + B_2 y_1 = P_2 x_2 + B_2 (C_1^T x_1) // \quad y_2 = C_2^T x_2 //$$

$$\dot{x} = \begin{bmatrix} A_1 & -C_1^T \\ C_1^T & A_2 \end{bmatrix} x + \begin{bmatrix} B_1 \\ 0 \end{bmatrix} u$$

$$y_1 = C_1^T x_1$$



1.4-19)

Given:

$$(1) \begin{bmatrix} A & A_1 \\ 0 & A_0 \end{bmatrix} \begin{bmatrix} b \\ 0 \end{bmatrix} \begin{bmatrix} C & D \\ 0 & 0 \end{bmatrix}$$

Both have same transfer function

$$(2) \begin{bmatrix} A & 0 \\ A_1 & A_2 \end{bmatrix} \begin{bmatrix} b \\ q \end{bmatrix} \begin{bmatrix} C & 0 \end{bmatrix}$$

Find:

Show that realizations can have different state numbers

Solution

Transfer Function from a SS realization B

$$G(z) = C(zI - A)^{-1}B + D$$

Plug values for (1) in

$$G_1(z) = \frac{bc}{-A+x} //$$

Similarly for (2)

$$G_2(z) = \frac{bc}{-A+x} //$$

1.4-21)

Given:

(A, b, C^T, d) and $d \neq 0$, describes a system $H(s)$ state space form.

Find:

Show that

$$(A - bC^T/d, b/d, -C^T/d, 1/d)$$

describes $H(s)$

Solution:

$G(s)G(s)^T$ has to be 1 if $G(s) = G(s)^{-1}$

$$[C^T(SI - A)^{-1}B + d] \left[-\frac{C^T}{d}(SI - A + bC^T)^{-1} \frac{b}{d} + \frac{1}{d} \right]$$

$$\left(\begin{matrix} BC \\ -A + s \\ D \end{matrix} \right) \left(-\frac{BC}{D^2(-A + \frac{BC}{D} + s)} + \frac{1}{D} \right)$$

Multiplying out and simplifying gives $G(s)G(s)^T = 1$

1.4-26)

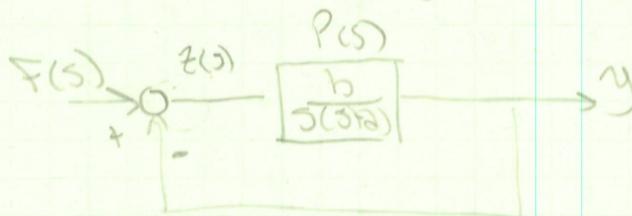
Given

Assume a system has an open-loop transfer function

$$H_0(s) = \frac{b}{s(s+a)}$$

Find:

a) Show that a system with the configuration



has the transfer function

$$H_C(s) = \frac{Y(s)}{F(s)} = \frac{1}{1 + (\frac{b}{f_0})s + (\frac{t_0}{f_0})s^2}$$

b) Show

$$\frac{1}{H_C(j\omega)} = A(j\omega) \angle \phi(j\omega)$$

where

$$A(j\omega) = \frac{1}{b} \sqrt{(b - \omega^2)^2 + (0\omega)^2} \quad \text{and} \quad \tan \phi = \frac{\omega}{b - \omega^2}$$

$A(j\omega)$ and $\phi(j\omega)$ correspond to the reciprocal amplitude and the phase difference between input and output

c)

Show that if amplitude/phase measurements are taken at n frequencies ω and ϕ can be solved by

$$\left[\begin{array}{cc|c} A(j\omega_1) & -\omega_1 \sqrt{1 + \tan^2 \phi(j\omega_1)} & 0 \\ \tan \phi(j\omega_1) & -\omega_1 & w_1^2 \tan \phi(j\omega_1) \\ \vdots & \vdots & \vdots \\ A(j\omega_n) & -\omega_n \sqrt{1 + 1/\tan^2 \phi(j\omega_n)} & 0 \\ \tan \phi(j\omega_n) & -\omega_n & w_n^2 \tan \phi(j\omega_n) \end{array} \right] = \left[\begin{array}{c} b \\ 0 \\ \vdots \\ 0 \end{array} \right]$$

Solution

a)

$$Y(s) = Z(s) P(s)$$

$$Z(s) = F(s) - Y(s)$$

$$F(s) = Z(s) + Y(s)$$

$$= Z(s) + Z(s) P(s)$$

$$\frac{Y(s)}{F(s)} = \frac{Z(s) P(s)}{Z(s) + P(s) Z(s)}$$

$$= \frac{P(s)}{1 + P(s)} = \frac{\frac{b}{s(s+a)}}{1 + \frac{b}{s(s+a)}} \left(\frac{\frac{s(s+a)}{b}}{\frac{s(s+a)}{b}} \right)$$

$$= \frac{1}{s(s+a) + b}$$

b)

$A(j\omega)$ is amplitude, so rewrite

$$\frac{1}{b} (b + 2j\omega - \omega^2)$$

take magnitude of real and imaginary to get

$$\frac{1}{b} \sqrt{(b - \omega^2)^2 + (\alpha\omega)^2}$$

Similarly phase is $\tan^{-1} \left(\frac{\text{Imaginary}}{\text{Real}} \right)$

$$\phi = \tan^{-1} \left(\frac{\alpha\omega}{b - \omega^2} \right)$$

c)

At this point there are only 2 unknowns let

$$A_n = A(j\omega_n) ; \tan\phi(j\omega_n) = \phi_n ; \xi_n = b\omega_n \sqrt{1 + \tan^2 \phi(j\omega)}$$

and $a_n = A_n$

Write augmented matrix

$$\left[\begin{array}{ccc|c} a_1 & \xi_1 & 0 \\ d_1 & -\omega_1 & \omega_1^2 \phi_1 \\ a_2 & \xi_2 & 0 \\ d_2 & -\omega_2 & \omega_2^2 \phi_2 \\ \vdots & \vdots & \vdots \\ a_n & \xi_n & 0 \\ \phi_n & \phi_n & \omega_n^2 \phi_n \end{array} \right]$$

for the case of 2 you get

$$\left[\begin{array}{ccc|c} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{array} \right]$$

which can be solved
because its full rank

1.4-30)

Given:

The following data is measured from a third order system

$$\mathbf{y} = [0.32, 0.25, 0.1, -0.022, 0.0006, -0.0012, \\ 0.0005, -0.0001]$$

Assume first time index is 0, so $y[0] = 0.32$

Find

a) Determine the modes in the system and then plot them in the complex plane

b) The data can be written as

$$y[t] = c_1(p_1)^t + c_2(p_2)^t + c_3(p_3)^t \quad t \geq 0$$

Find c_1, c_2, c_3

c) Add Gaussian random noise to the system with $\sigma^2=0.01$ then find modes of noisy data.

Do this several times and comment.

Solution:

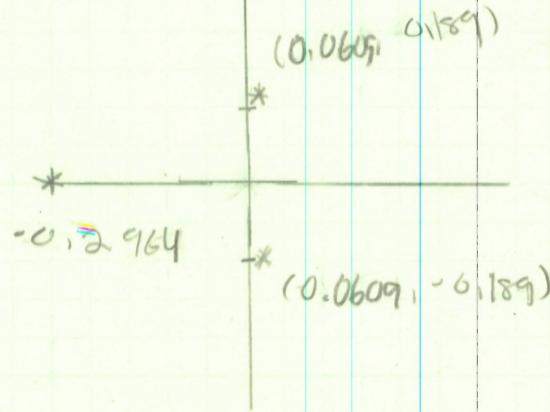
a)

$$\begin{bmatrix} -y_0 & -y_1 & -y_0 \\ -y_3 & -y_2 & -y_1 \\ -y_4 & -y_3 & -y_2 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} = \begin{bmatrix} y_3 \\ y_4 \\ y_5 \end{bmatrix}$$

$$a_1 = 1.75 \cdot 10^{-1}, \quad a_2 = 3.24 \cdot 10^{-3} \quad 1.166 \cdot 10^{-2}$$

$$z^3 + 1.75 \cdot 10^{-1} z^2 + 3.24 \cdot 10^{-3} z + 1.166 \cdot 10^{-2}$$

$$z = -0.2964 \quad z = 0.0609 \pm 0.189j$$



b)

$$z^2 + a_1 z_1 + a_2 = (z - p_1)(z - p_2)$$

$y[t] = C_1(p_1)^t + C_2(p_2)^t$ p_1 and p_2 are based
on initial conditions

$$y[t] = C_1 e^{0.0609t} \cos(0.189t) + C_2 e^{0.0609t} \sin(0.189t) + C_3 e^{-0.2964t}$$

$$\begin{bmatrix} y[0] \\ y[1] \\ y[2] \end{bmatrix} = \begin{bmatrix} p_{10}, p_{20}, p_{30} \\ p_{11}, p_{21}, p_{31} \\ p_{12}, p_{22}, p_{32} \end{bmatrix} \begin{bmatrix} C_1 \\ C_2 \\ C_3 \end{bmatrix} = \begin{bmatrix} 1.0439 & 0.1997 & 0.7448 \\ 0.0499 & 0.4169 & 0.5555 \\ 1.0126 & 0.6448 & 0.413 \end{bmatrix} \begin{bmatrix} C_1 \\ C_2 \\ C_3 \end{bmatrix}$$

$$C_1 = 0.9263, C_2 = -0.8972, C_3 = -0.6281$$

(1)

$$\sigma^2 = 0.01$$

$$\text{modes} = [-0.4211, 6.0345 \pm 0.37j]$$

$$\sigma^2 = 0.05$$

$$\text{modes} = [-0.3466, 0.1927 \pm 0.162j]$$

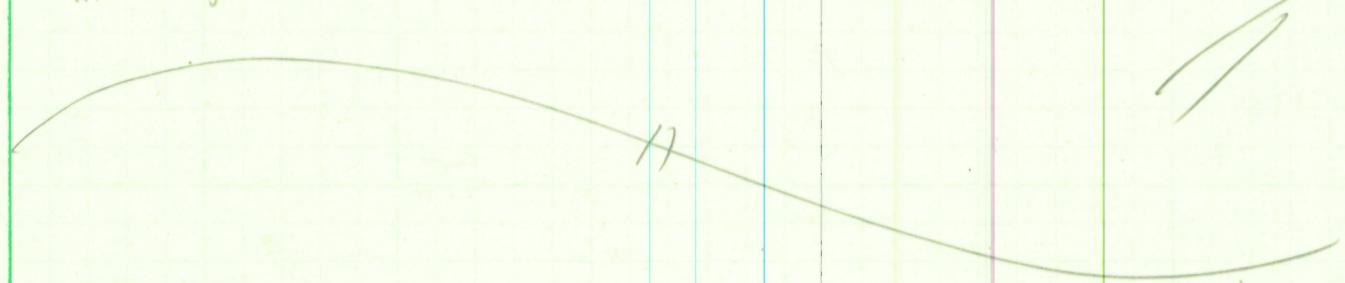
$$\sigma^2 = 0.1$$

$$\text{modes} = -0.371, 0.063 \pm 0.181j$$

$$\sigma^2 = 0.0001$$

$$\text{modes} = [-0.609, 0.0537 \pm 0.49j] -$$

The smaller the variance, the larger the modes seem to be in magnitude.



1.4 - 31

11

Given:

$$y[t] = A \cos(\omega_1 t + \theta_1) + B \cos(\omega_2 t + \theta_2)$$

And the frequencies are known

Find:

Determine the means of computing the amplitudes and phases at t_0, t_1, t_2, t_3 .

Solution:

$$y[0] = A \cos(\omega_1 t_0 + \theta_1) + B \cos(\omega_2 t_0 + \theta_2)$$

$$y[1] = A \cos(\omega_1 t_1 + \theta_1) + B \cos(\omega_2 t_1 + \theta_2)$$

$$y[2] = A \cos(\omega_1 t_2 + \theta_1) + B \cos(\omega_2 t_2 + \theta_2)$$

$$y[3] = A \cos(\omega_1 t_3 + \theta_1) + B \cos(\omega_2 t_3 + \theta_2)$$

Now we have 4 eq's and 4 unknowns



B.1-1)

Given:

The characteristic function of a random variable X is the (conjugate of) Fourier transform of its density

$$\Phi_X(\omega) = \int f_X(x) e^{j\omega x} dx$$

Find:

a) Show that a Gaussian density with

$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left[-\frac{(x-\mu)^2}{2\sigma^2}\right]$$

has the characteristic function

$$\Phi_X(\omega) = \exp\left[j\mu\omega - \frac{1}{2}\omega^2\sigma^2\right]$$

b) Show that the n^{th} moment of X can be obtained from its characteristic function by

$$E[X^n] = \frac{1}{j^n} \left. \frac{d^n \Phi_X(\omega)}{d\omega^n} \right|_{\omega=0}$$

Solution:

Look at exponent

$$E = \frac{1}{\sigma^2} (x^2 + \mu^2 - 2\mu x + 2\sigma^2\omega x)$$

$$= \frac{1}{\sigma^2} (x^2 - 2(\mu - \sigma^2\omega)x + \mu^2 + (\mu - \sigma^2\omega)^2 - (\mu - \sigma^2\omega)^2)$$

$$= \frac{1}{\sigma^2} (x - (\mu - \sigma^2\omega))^2 - (\mu - \sigma^2\omega)^2 + \mu^2$$

$$\begin{aligned} \mathcal{F}(f_x(x)) &= \frac{1}{\sqrt{2\pi}\sigma} \exp\left[-\frac{1}{2\sigma^2}((x-\mu)^2 + \sigma^2\omega^2)\right] + \cancel{\int \exp\left[x - \frac{(x-\mu)^2 + \sigma^2\omega^2}{2\sigma^2}\right]} \\ &= \exp\left[\frac{1}{2\sigma^2}((x-\mu)^2 + \sigma^2\omega^2)\right] \\ &= \exp\left[-\frac{\sigma^2\omega^2}{2} + j\mu\omega\right] \end{aligned}$$

b)

$$\begin{aligned} \phi_x(\omega) &= E[e^{j\omega x}] \quad \text{Expand using Taylor series} \\ &= E[1 + j\omega x + \frac{(j\omega x)^2}{1!} + \frac{(j\omega x)^3}{3!}] \\ &= E[1] + E\left[\frac{j\omega x}{1!}\right] + E\left[\frac{(j\omega x)^2}{2!}\right] + \dots \end{aligned}$$

If $n=0$ then no derivatives are taken. Then the expected value is

$$E[1]$$

taking one derivative leaves

$$0 + E\left[\frac{jx}{1!}\right] + E\left[\frac{0(j\omega x)}{2!}\right] + \dots$$

$$\omega=0$$

$$= E\left[\frac{jx}{1!}\right]$$

Taking subsequent derivatives then removes the previous values. The next term loses all ω 's.



B.1-2)

Given:

$$\frac{1}{\sqrt{2\pi(1-\rho^2)\sigma_x}} \exp \left[-\frac{1}{2\sigma_x^2(1-\rho^2)} \left(x - \left(\mu_x + \frac{\partial x}{\partial y} \rho (y - \mu_y) \right) \right)^2 \right]$$

Show:

That the above eq is correct

Solution

$$f(x|y) = \frac{1}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \exp \left[-\frac{1}{2(1-\rho^2)} \left(\frac{(x-\mu_x)^2}{\sigma_x^2} + \frac{(y-\mu_y)^2}{\sigma_y^2} - \frac{2\rho}{\sigma_x\sigma_y} (x-\mu_x)(y-\mu_y) \right) \right]$$



Let's look at this

$$-\frac{1}{2(1-\rho^2)\sigma_x^2} \left((x-\mu_x)^2 + \frac{(y-\mu_y)^2}{\sigma_y^2} \sigma_x^2 - \frac{2\rho\sigma_x}{\sigma_y} (x-\mu_x)(y-\mu_y) \right)$$

Forget
this
term

$$-\frac{1}{2(1-\rho^2)\sigma_x^2} \left(x^2 - 2\mu_x x + \mu_x^2 - \frac{2\rho\sigma_x}{\sigma_y} (xy - x\mu_y - y\mu_x + \mu_x\mu_y) \right)$$

$$-\frac{1}{2(1-\rho^2)\sigma_x^2} \left(x^2 - 2 \left[\mu_x + \frac{\rho\sigma_x}{\sigma_y} (y - \mu_y) \right] x + \mu_x^2 - \frac{2\rho\sigma_x}{\sigma_y} (-y\mu_x + \mu_x\mu_y) \right)$$

$$-\frac{1}{2(1-\rho^2)\sigma_x^2} \left((x - \left[\mu_x + \frac{\sigma_x}{\sigma_y} (y - \mu_y) \right])^2 + \frac{\sigma_x^2(y - \mu_y)^2}{\sigma_y^2} \right)$$

$$f(x|y) = \frac{1}{2\pi\sigma_x\sigma_y} \exp \left(-\frac{1}{2(1-\rho^2)\sigma_x^2} \left(x - \left(\mu_x + \frac{\sigma_x}{\sigma_y} \rho (y - \mu_y) \right) \right)^2 \right) +$$

$$\exp \left(\frac{(y - \mu_y)^2 \sigma_x^2}{\sigma_y^2} \right)$$

$$\frac{1}{\sqrt{2\pi}\sigma_y} \exp \left[-\frac{1}{2\sigma_y^2} (y - \mu_y)^2 \right]$$

$$f(x|y) = \frac{1}{\sqrt{2\pi(1-\rho^2)\sigma_x^2}} \exp\left[-\frac{1}{2\sigma_x^2(1-\rho^2)}(x - (\mu_x + \frac{\partial_x\rho}{\partial_y}(y - \mu_y)))^2\right]$$



B.2-3)

Given

$$(x+z)^T A(x+z) + d ; z = \frac{1}{2} A^{-1}y ; d = c - \frac{1}{2} y^T z$$

Find:

$f(x|y)$ when

$$Y = X + N$$

and X and N are independent Gaussian random vectors w/

$$X \sim \mathcal{N}(M, R_x) \quad N \sim \mathcal{N}(0, R_n)$$

Identify the mean of $X|Y$ and variance $X|Y$

Solution

$$\begin{aligned} E &= \frac{1}{2} [(y-x)^T R_n^{-1} (y-x) + (x-Mx)^T R_x^{-1} (x-Mx)] \\ &= -\frac{1}{2} \left[x^T \underbrace{\left((R_n^{-1} + R_x^{-1}) \right)}_{A} x - 2 \underbrace{(R_n^{-1} y + R_x^{-1} Mx)}_{y} \right] + c \end{aligned}$$

$$(x+z)^T A(x+z) + d$$

$$z = \frac{1}{2} (R_n + R_x) (R_n^{-1} y + R_x^{-1}) Mx \quad A = (R_n^{-1} + R_x^{-1})$$

$$d = c_1 - \frac{1}{2} y^T z$$

$$M_{X|Y} = -z$$

$$R_{X|Y} = R_n + R_x$$

1.8-41)

Show that there are an infinite number of primes

Hint: Use proof by contradiction.

Solution

Suppose that there are a finite set of prime numbers

$$P = \{2, 3, 5, 7, \dots, p_n\}$$

where p_n is the last prime in the set.

$$\text{Let } q = 1 + \prod_{p_i \in P} p_i$$

Stating the fact that every natural number, $n \geq 1$, is divisible by some prime number.

Now going back to q . This new number we created has to either be prime or not prime. Using our set q and the fact stated above suppose we divide q by every $p_i \in P$.

$$\forall p_i \in P \quad \frac{q}{p_i}$$

They would all have a remainder which violates the fact we stated. Therefore, there must be another prime that our number q is divisible by.



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1.8-43)

Given:

$$\sum_{i=0}^n 2^i$$

Find:

- By trial and error, find a formula for what is above.
- Prove by induction that your formula is correct.

Solution:

a)

Suppose the formula of the form

$$f(n) = 2^{n+1} - 1 \quad \checkmark$$

Base checks

$$1: 2^0 + 2^1 = 3 \quad f(1) = 2^2 - 1 = 3$$

$$2: 2^0 + 2^1 + 2^2 = 7 \quad f(2) = 2^3 - 1 = 7$$

b)

Base case

$$0: 2^0 = 1$$

$$2^0 - 1 = 1 \quad \checkmark$$

Suppose $\sum_{i=0}^n 2^i = 2^{n+1} - 1$

Then add 2^{n+1} to each side

$$\sum_{i=0}^{n+1} 2^i = 2^{n+2} - 1$$

Which concludes the proof

(1.8-45)

Given:

$$\sum_{i=1}^{n+1} \frac{1}{i^2+i}$$

Find:

a) Formula for the above written summation

b) Prove the formula by induction

Solution:

a)

Suppose the formula of the form

$$f(n) = 1 - \frac{1}{n+1} = \frac{n}{n+1}$$

Basic checks

$$1: \frac{1}{1^2+1} = \frac{1}{2}$$

$$f(1) = 1 - \frac{1}{2} = \frac{1}{2}$$

$$\begin{aligned} 2: & \frac{1}{2^2+2} + \frac{1}{2} \\ &= \frac{1}{6} + \frac{1}{2} = \frac{2}{3} \end{aligned}$$

$$f(2) = 1 - \frac{1}{3} = \frac{2}{3}$$

b)

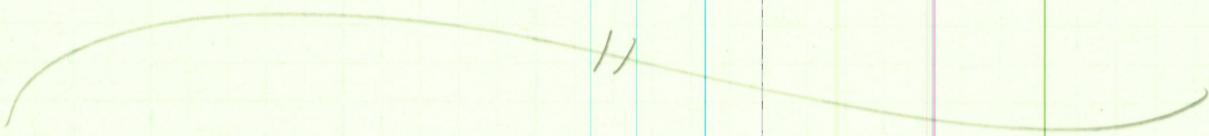
Suppose

$$\sum_{i=1}^n \frac{1}{i^2+i} = 1 - \frac{1}{n+1}$$

then

$$\sum_{i=1}^{n+1} \frac{1}{i^2+i} = \sum_{i=1}^n \frac{1}{i^2+i} + \frac{1}{(n+1)^2+(n+1)}$$

$$\begin{aligned} \sum_{i=1}^{n+1} \frac{1}{i^2+i} &= \sum_{i=1}^{n+1} \frac{1}{i^2+i} + \frac{1}{(n+1)^2(n+1)} = \frac{n}{n+1} + \frac{1}{(n+1)^2(n+1)} \\ &= \frac{n}{n+1} + \frac{1}{(n+1)(n+2)} = \frac{n(n+2) + 1}{(n+1)(n+2)} = \frac{n^2+2n+1}{(n+1)(n+2)} \\ &= \frac{(n+1)^2}{(n+1)(n+2)} = \frac{n+1}{n+2} \end{aligned}$$



1.8-4b)

Given: $n^3 - n$

Show: $\nexists \mathbb{N} \mid n > 0, n \in \mathbb{N}$ such that $n^3 - n$ is divisible by 3.

Solution

Base case

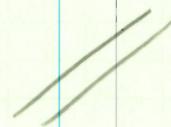
$$0, 1: 1^3 - 1 = 1 - \frac{1}{3} = 0$$

$$2: 2^3 - 2 = 6 \quad ; \quad \frac{6}{3} = 2$$

Suppose $3k = n^3 - n$

$$\begin{aligned} \text{Then } 3k &= (n+1)^3 - (n+1) \\ &= (n^3 - n) + 3n^2 + 3n \\ &= 3(k + n^2 + n) \end{aligned}$$

which must be divisible by 3.



1.8-47)

Given:

$$\binom{n}{k} = \frac{n!}{k!(n-k)!} \quad \text{where } n \geq k$$

Find:

Show by induction for $1 \leq k \leq n$

$$\binom{n+1}{k} = \binom{n}{k} + \binom{n}{k-1}$$

Solution

If $\binom{n+1}{k} = \binom{n}{k} + \binom{n}{k-1}$ then show base case

$$k=1, n=1$$

$$\binom{2}{1} = \binom{1}{1} + \binom{1}{0}$$

$$\frac{2!}{1!(2-1)!} = \frac{1!}{1!(0)!} + \frac{1!}{0!1!}$$

$$\frac{2}{1} = 1 + 1 = 2 \checkmark$$

Let $\binom{n+1}{k} = \binom{n}{k} + \binom{n}{k-1}$ be true

Suppose then we start with

$$\binom{n}{k-1} = \binom{n}{k-1} \quad \binom{n}{k-2}$$

Then by induction

$$\binom{n+1}{k} = \frac{n!}{(n-k)! k!} + \frac{n!}{(n-(k-1))! (k-1)!}$$

$$= \frac{n!}{(n-k)! k!} + \frac{n!}{(n-k+1)! (k-1)!}$$

$$= \frac{(n-k+1)n!}{(n-k+1)(n-k)! k!} + \frac{n!k}{(n-k+1)! (k-1)!}$$

$$= \frac{n \cdot n! - kn! + n! + n!k}{(n-k+1)! k!}$$

$$= \frac{n \cdot n! + n!}{(n-k+1)! k!} = \frac{n! (n+1)}{(n-k+1)! k!} = \frac{(n+1)!}{((n+1)-k)! k!} = \binom{n+1}{k}$$

H

