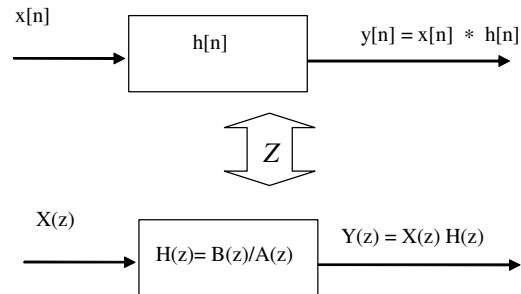


Student Name: _____

Part 1**Approximating a system response**

The Prony method follows the same general idea as the Pade method, except that the derivation of $H(z)$ incorporates more than the first few points in $g[n]$.

As for Pade method, a system is represented by its impulse response $h[n]$ in the time domain, or by its Z-transform $H(z) = B(z)/A(z)$ in the z-domain. As indicated above, the z-transform may be taken across the entire block diagram.

Suppose that we want to approximate a signal $g[n]$ as the output of the above system for some input signal $x[n]$. For example, we may wish to do this to find some more compact representation of $g[n]$ that may require less memory for storage or less bits for some data transmission.

In most cases, the input signal is presumed to be the unit sample, $x[n] = \delta[n]$. Therefore, the focus is on finding some compact polynomial $H(z) = B(z)/A(z)$ that represents $g[n]$ when $x[n] = \delta[n]$.

The trick is then to solve for polynomials $A(z)$ and $B(z)$, given some signal $g[n]$ to be approximated, and given that $x[n] = \delta[n]$. In the Prony method, a least squares approach is used to minimize the error $e[n] = \sum (g[n] - h[n])^2$

Finally, note that:

1. Prony method does not guarantee a stable solution for $H(z)$
2. Prony does not necessarily exactly match the points of $g[n]$,
3. Prony uses more than the first $n_a + n_b + 1$ points of $g[n]$, where n_a and n_b are the order of the polynomials $A(z)$ and $B(z)$... a least squares error measure is used.
4. Lower order models are used to approximate a large number of points in $g[n]$
5. Using more points from $g[n]$, i.e., using more data, should result in a better approximation $h[n]$.

As for the Pade method, to solve for $H(z)$, first rearrange the equations:

$$H(z) = B(z)/A(z)$$

To

$$H(z)A(z) = B(z)$$

But $H(z) = Y(z)/X(z)$ and we want $Y(z) = G(z)$ and so,

$$H(z)A(z) = Y(z)A(z)/X(z) = G(z)A(z)/X(z) = B(z)$$

and since $x[n] = \delta[n]$, and so $X(z) = 1$, then

$$G(z)A(z) = B(z)$$

Finally, taking the inverse z-transform

$$g[n] * a[n] = b[n]$$

And using the definition of convolution

$$\sum_{p=0}^{Na} g[p] a[n-p] = b[n]$$

Which can be written in matrix form as follows, with the main difference between Prony and Pade that the number of rows in the g -matrix are greater than the $Na + Nb + 1$ rows used in Pade method. Hence, the system of equations are overdetermined, and will most commonly not have an exact solution. Therefore a least-squares solution will be needed. The order of $A(z)$ is Na , and the order of $B(z)$ is Nb .

Prony form of problem shown below.
(The number of rows $> Na + Nb + 1$)

$$\begin{pmatrix} g_0 & 0 & 0 \\ g_1 & g_0 & 0 \\ g_2 & g_1 & g_0 \\ g_3 & g_2 & g_1 \\ g_4 & g_3 & g_2 \\ g_5 & g_4 & g_3 \\ g_6 & g_5 & g_4 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ a_1 \\ a_2 \end{pmatrix} = \begin{pmatrix} b_0 \\ b_1 \\ b_2 \\ 0 \\ 0 \end{pmatrix}$$

Pade form of problem shown below.
(The number of rows = $Na + Nb + 1$)

$$\begin{pmatrix} g_0 & 0 & 0 \\ g_1 & g_0 & 0 \\ g_2 & g_1 & g_0 \\ g_3 & g_2 & g_1 \\ g_4 & g_3 & g_2 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ a_1 \\ a_2 \end{pmatrix} = \begin{pmatrix} b_0 \\ b_1 \\ b_2 \\ 0 \\ 0 \end{pmatrix}$$

Q1. Is the Prony approximation IIR?

Part 2

Least squares and left psuedo-inverse

A commonly encountered problem is the solution of an overdetermined system of equations, where the number of equations exceeds the number of variables. Since an exact solution is not usually possible, a common approach is to minimize the error.

In matrix form, let the matrix A have m rows and n columns, where $m > n$ to be overdetermined, and let the rank of A be n (meaning that the columns of A are independent). Let vector x represent the variables to be solved for, and y represent measured output data, or equivalently, let $-y$ represent the vector which is to be approximated by Ax . Then, the least squares problem is written as

$$Ax + y = 0$$

(An alternative form of the problem would be $Ax = y$, and only differs by a change of sign in y .) Since A is not a square matrix, it is not generally invertible. So define an error vector ϵ

$$\epsilon = Ax + y$$

where ϵ is a measure of how much $Ax + y$ differs from the ideal value of 0. The least squares error measure is the L2 norm of ϵ :

$$E = \epsilon^H \epsilon$$

where e^H is the hermitian transpose, i.e., the complex conjugate transpose operator. It can be shown that E is minimized when the solution x_s is given as

$$x = x_s = - \{ A^H A \}^{-1} A^H y$$

where $\{ A^H A \}^{-1} A^H$ is called the left psuedo-inverse of A .

Two important properties of the least squares solution are:

1. The error e is orthogonal to the column space of A , and hence lies completely outside the column space of A . Mathematically, $A^H \epsilon = 0$.
2. The vector Ax_s (i.e., the approximation of $-y$) lies entirely within the column space of A , and hence, Ax_s is orthogonal to e . $(Ax_s)^H \epsilon = x_s^H A^H \epsilon = 0$.

As a simple least squares example, consider the following overdetermined problem:

$$A := \begin{pmatrix} 2 & 1 \\ 1 & 2 \\ 2 & 2 \end{pmatrix} \quad y := \begin{pmatrix} -11 \\ -11 \\ -11 \end{pmatrix}$$

Then, create an appropriately sized x-vector:

$$N_x := \text{cols}(A) - 1 \quad n_x := 0..N_x \quad x_{n_x} := 0 \quad x = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Then, make sure the rank of A equals the number of columns: $\text{rank}(A) = 2$

Finally, get the solution:

$$x_s := -\left[\left((\overline{A})^T \cdot A \right)^{-1} \cdot (\overline{A})^T \cdot y \right] \quad x_s = \begin{pmatrix} 3.235 \\ 3.235 \end{pmatrix}$$

Note: in Mathcad, the bar over the A indicates **complex conjugate** and is entered using the double quote key when the cursor is just behind the A.

Also, note that psuedoinverse acts as an inverse:

$$\left[\left((\overline{A})^T \cdot A \right)^{-1} \cdot (\overline{A})^T \right] \cdot A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

And the error vector e is

$$\varepsilon := A \cdot x_s + y \quad \varepsilon = \begin{pmatrix} -1.294 \\ -1.294 \\ 1.941 \end{pmatrix}$$

And the error is

$$E := (\overline{\varepsilon})^T \cdot \varepsilon \quad E = (7.118)$$

and compare the approximation Ax_s with -y

$$A \cdot x_s = \begin{pmatrix} 9.706 \\ 9.706 \\ 12.941 \end{pmatrix} \quad -y = \begin{pmatrix} 11 \\ 11 \\ 11 \end{pmatrix}$$

And check that ε is orthogonal to the column space of A, and that ε is orthogonal to the approximation Ax_s of -y.

$$(\overline{A})^T \cdot \varepsilon = \begin{pmatrix} 1.599 \times 10^{-14} \\ 1.599 \times 10^{-14} \end{pmatrix} \quad (\overline{A \cdot x_s})^T \cdot \varepsilon = (1.034 \times 10^{-13})$$

Q2. Change the above least-squares problem so that the last row of A is [2, 1], and explain why the error E becomes zero.

Part 3

Using left psuedoinverse to solve for Prony H(z)

Next, we will use the left psuedo-inverse solve for the coefficients a1, a2, b0, b1, b2 as follows. As before, set the order of numerator as Nb, order of denominator as Na where a0, a1, ... aNa, bo, b1, b2, ... bNb

$$Na := 2 \qquad Nb := 2$$

First define the vectors representing the numberator coefficients as b, and denominator coefficients as a. Also, load these vectors with initial values (i.e., 0) for the coefficients and define the desired output response g[n]

$$\begin{aligned} na &:= 0..Na & nb &:= 0..Nb \\ a_{na} &:= 0 & b_{nb} &:= 0 \end{aligned}$$

Then, define the desired output response g[n]

$$g := \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 3 \\ 2 \end{pmatrix}$$

For the Pade method, the number of rows in the matrix was rows = Na + Nb +1, for the Prony method the system of equations must be overdetermined, so the Prony method requires rows > Na + Nb +1.

So define Ncmin as the minimum number of rows required in the matrix G.

$$\text{rows_in_G_min} := Na + Nb + 2 \qquad \text{rows_in_G_min} = 6$$

Since the number of rows in vector g will equal the number of rows in matrix G, set vector g equal to zero to force an error in the rest of the program if the number of rows does not satisfy the requirement rows > Na + Nb +1

$$g := \text{if}(\text{rows}(g) < \text{rows_in_G_min}, 0, g)$$

$$g = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 3 \\ 2 \end{pmatrix}$$

Then set up the matrix G that corresponds to the convolution $g[n] * a[n]$ when G is multiplied by column vector a.

$$\begin{aligned} \text{grows} &:= \text{rows}(g) & \text{gcols} &:= \text{rows}(a) \\ \text{gr} &:= 0.. \text{grows} - 1 & \text{gc} &:= 0.. \text{gcols} - 1 \end{aligned}$$

$$G_{\text{gr}, \text{gc}} := \text{if} \left[(\text{gr} - \text{gc} \geq 0) \wedge (\text{gr} - \text{gc} \leq \text{rows}(g) - 1), g_{\text{gr}-\text{gc}}, 0 \right]$$

$$G = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \\ 4 & 3 & 2 \\ 3 & 4 & 3 \\ 2 & 3 & 4 \end{pmatrix}$$

The Prony method using split G-matrix and matrix inverse is:

Step 1. Solve for a_1, a_2 using lower portion of matrix where there are zeroes in the vector on the right hand side of the equation (below the b_2 term).

This is done using the least-squares method, i.e., left psuedo-inverse.

Step 2. Solve for b_0, b_1, b_2 , using the top of the matrix and the solutions for a_1, a_2 .

Prony Step 1 is to first solve for a_1, a_2 values.

First get the lower right G submatrix, G_{br} , and lower left column, $g_{\text{left_col}}$

$$a_{\text{order}} := \text{rows}(a) - 1 \quad b_{\text{order}} := \text{rows}(b) - 1$$

$$G_{\text{br}} := \text{submatrix}(G, b_{\text{order}} + 1, \text{rows}(G) - 1, 1, \text{cols}(G) - 1)$$

$$g_{\text{left_col}} := \text{submatrix}(G, b_{\text{order}} + 1, \text{rows}(G) - 1, 0, 0)$$

$$G_{\text{br}} = \begin{pmatrix} 3 & 2 \\ 4 & 3 \\ 3 & 4 \end{pmatrix} \quad g_{\text{left_col}} = \begin{pmatrix} 4 \\ 3 \\ 2 \end{pmatrix}$$

then for this example we have to solve the following overdetermined system for vector a

$$G_{\text{br}} \cdot \begin{pmatrix} a_1 \\ a_2 \end{pmatrix} = -g_{\text{left_col}} \quad \text{or} \quad \begin{pmatrix} 3 & 2 \\ 4 & 3 \\ 5 & 4 \end{pmatrix} \cdot \begin{pmatrix} a_1 \\ a_2 \end{pmatrix} = -\begin{pmatrix} 4 \\ 5 \\ 0 \end{pmatrix}$$

The least-squares solution (if G_{br} has full rank) using the left pseudo-inverse is

$$\text{rank}(G_{br}) = 2$$

$$as := \left[\left(\overline{(G_{br})^T} \cdot G_{br} \right)^{-1} \cdot \overline{(G_{br})^T} \cdot g_{left_col} \right] \quad as = \begin{pmatrix} -1.395 \\ 0.581 \end{pmatrix}$$

Note: in Mathcad, the bar over the A indicates complex conjugate and is entered using the double quote key when the cursor is just behind the A.

Checking the pseudoinverse:

$$\left(\overline{(G_{br})^T} \cdot G_{br} \right)^{-1} \cdot \overline{(G_{br})^T} \cdot G_{br} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

Next, load the solution into vector a containing the coefficients of H(z):

$$a_0 := 1$$

$$cnt := 1 .. \text{rows}(a) - 1$$

$$a_{cnt} := as_{cnt-1}$$

And the error vector ε is

$$\varepsilon := G_{br} \cdot as + g_{left_col} \quad \varepsilon = \begin{pmatrix} 0.977 \\ -0.837 \\ 0.14 \end{pmatrix}$$

And the error is

$$E := \left(\overline{\varepsilon} \right)^T \cdot \varepsilon \quad E = (1.674) \quad \left(\overline{g_{left_col}} \right)^T \cdot \varepsilon = (1.674)$$

and compare the approximation Ax_s with -y

$$G_{br} \cdot as = \begin{pmatrix} -3.023 \\ -3.837 \\ -1.86 \end{pmatrix} \quad \underline{g}_{left_col} = \begin{pmatrix} -4 \\ -3 \\ -2 \end{pmatrix}$$

Q3. Is G_{br} in the above example full rank?

And check that e is orthogonal to the column space of A , and that ε is orthogonal to the approximation Ax_s of $-y$.

$$\left(\overline{(G_{br})}\right)^T \cdot \varepsilon = \begin{pmatrix} -1.132 \times 10^{-14} \\ -9.77 \times 10^{-15} \end{pmatrix} \quad \left(\overline{(G_{br} \cdot as)}\right)^T \cdot \varepsilon = \begin{pmatrix} 1.034 \times 10^{-14} \end{pmatrix}$$

Summarizing:

We have now used least-squares method to solve for the coefficients of the denominator, $A(z)$

$$a = \begin{pmatrix} 1 \\ -1.395 \\ 0.581 \end{pmatrix}$$

with error E being

$$E = (1.674)$$

Prony Step 2 is to solve for b_0, b_1, b_2 values.

First get the top G submatrix, G_t , and solve for the upper right side of the equations

$$G_t := \text{submatrix}(G, 0, b_order, 0, \text{cols}(G) - 1)$$

$$G_t = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \end{pmatrix}$$

Then the solution for the vector b containing the coefficients of $B(z)$ is

$$b := G_t \cdot a$$

$$b = \begin{pmatrix} 1 \\ 0.605 \\ 0.791 \end{pmatrix}$$

Summarizing steps 1 and 2, the Prony solution is:

$$a = \begin{pmatrix} 1 \\ -1.395 \\ 0.581 \end{pmatrix} \quad b = \begin{pmatrix} 1 \\ 0.605 \\ 0.791 \end{pmatrix} \quad E = (1.674) \quad \epsilon = \begin{pmatrix} 0.977 \\ -0.837 \\ 0.14 \end{pmatrix}$$

Where the desired function t to be matched was vector g :

$$g = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 3 \\ 2 \end{pmatrix} \quad G = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \\ 4 & 3 & 2 \\ 3 & 4 & 3 \\ 2 & 3 & 4 \end{pmatrix}$$

And the system response is $H(z)=N(z)/D(z)=\text{Num}(z)/\text{Den}(z)$ where

$$\text{Num}(b, z) := \sum_{nn=0}^{\text{rows}(b)-1} b_{nn} \cdot z^{-nn}$$

$$\text{Den}(a, z) := \sum_{nn=0}^{\text{rows}(a)-1} a_{nn} \cdot z^{-nn}$$

or

$$H(a, b, z) := \frac{\sum_{nn=0}^{\text{rows}(b)-1} b_{nn} \cdot z^{-nn}}{\sum_{nn=0}^{\text{rows}(a)-1} a_{nn} \cdot z^{-nn}}$$

Note also, that the Prony method has a nonzero error in the approximation of $g[n]$ by the impulse response of the system, $y[n] = h[n]$.

This is because the Prony method minimizes the mean square error of an overdetermined system of equations, it does not guarantee zero. However, note that the overdetermined system resulted from more data, i.e., more samples of $g[n]$. Hence, it should result in a better model since the Prony method uses more data in estimating the model.

Part 4

Checking the impulse response of the Prony solution

Compute the first N_p points of $y[n]=h[n]$:

$$N_p := \text{rows}(g) \quad a = \begin{pmatrix} 1 \\ -1.395 \\ 0.581 \end{pmatrix} \quad b = \begin{pmatrix} 1 \\ 0.605 \\ 0.791 \end{pmatrix}$$

$$k := 0..N_p - 1 \quad a_1 := a_1 \quad a_2 := a_2 \quad b_0 := b_0 \quad b_1 := b_1 \quad b_2 := b_2$$

$$y_k := 0$$

$$y_k := \text{if}(k = 0, b_0, 0)$$

$$y_k := \text{if}(k = 1, b_1 - a_1 \cdot y_{k-1}, y_k)$$

$$y_k := \text{if}(k = 2, b_2 - a_1 \cdot y_{k-1} - a_2 \cdot y_{k-2}, y_k)$$

$$y_k := \text{if}(k > 2, -a_1 \cdot y_{k-1} - a_2 \cdot y_{k-2}, y_k)$$

So, our model $y[n]$ is compared to the desired response $g[n]$:

$$y = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 3.023256 \\ 2.47431 \\ 1.694819 \end{pmatrix} \quad g = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 3 \\ 2 \end{pmatrix} \quad g - y = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0.977 \\ 0.526 \\ 0.305 \end{pmatrix}$$

With error

$$\left(\overline{(g-y)}\right)^T \cdot (g-y) = (1.324)$$

Note that the error here, $(g-y)^H(g-y)$ is not the same error measure $e^H e$ that was used in the least-squares algorithm. This is because the least-squares algorithm used an "indirect error measure" since it was based on an "indirect method" for finding $y[n]=h[n]$ to model $g[n]$.

- Q4.** Is the Prony approximation $y[n]$ nearly equal to $g[n]$ for $n=0$ to $n=5$ when when $g[n] = \{ 1, 2, 3, 4, 3, 2 \}$?
- Q5.** Add a new plot right below this question, showing the first 20 points of the prony approximation (i.e., the impulse response of $H(z)$) for for $g[n] = \{ 1, 2, 3, 4, 3, 2 \}$.

Part 5

Checking the stability response of the Prony solution

$$a = \begin{pmatrix} 1 \\ -1.395 \\ 0.581 \end{pmatrix} \quad b = \begin{pmatrix} 1 \\ 0.605 \\ 0.791 \end{pmatrix}$$

Find poles and zeroes

$$a_roots := \text{polyroots}(a) \quad a_roots = \begin{pmatrix} 1.2 - 0.529i \\ 1.2 + 0.529i \end{pmatrix}$$

$$Hpoles := \frac{1}{a_roots} \quad Hpoles = \begin{pmatrix} 0.698 + 0.308i \\ 0.698 - 0.308i \end{pmatrix} \quad \overrightarrow{|Hpoles|} = \begin{pmatrix} 0.762 \\ 0.762 \end{pmatrix}$$

$$b_roots := \text{polyroots}(b) \quad b_roots = \begin{pmatrix} -0.382 + 1.058i \\ -0.382 - 1.058i \end{pmatrix}$$

$$Hzeroes := \frac{1}{b_roots} \quad Hzeroes = \begin{pmatrix} -0.302 - 0.836i \\ -0.302 + 0.836i \end{pmatrix}$$

Lets plot poles and Zeroes in z-plane

$$\text{ImagZeroes} := \text{Im}(\text{Hzeroes})$$

$$\text{RealZeroes} := \text{Re}(\text{Hzeroes})$$

$$\text{ImagPoles} := \text{Im}(\text{Hpoles})$$

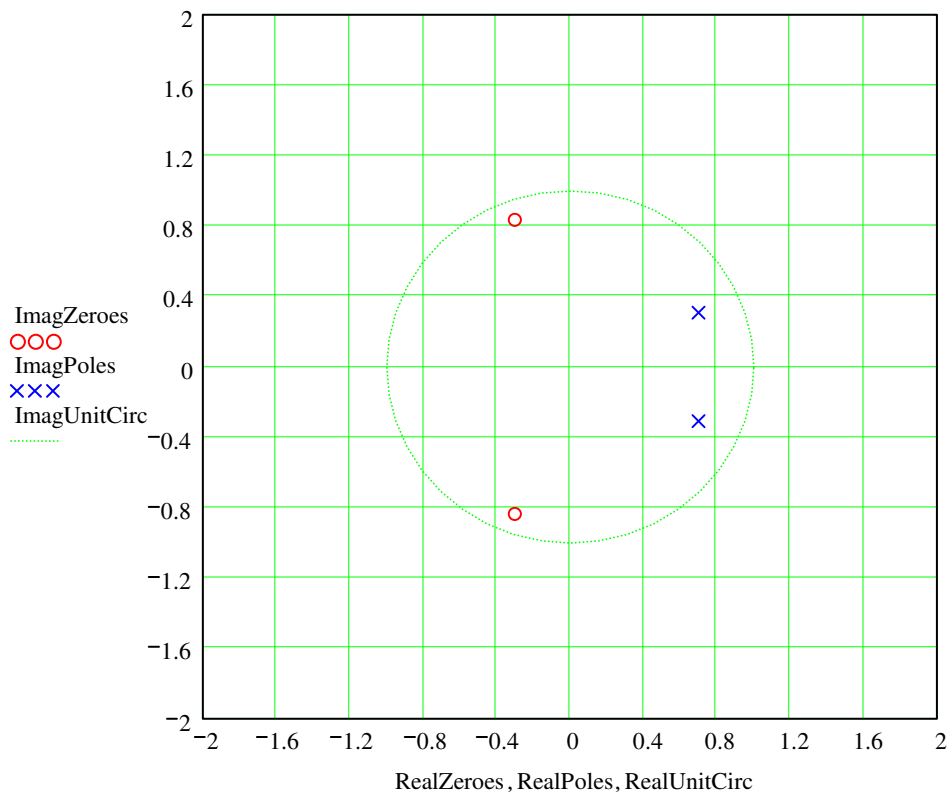
$$\text{RealPoles} := \text{Re}(\text{Hpoles})$$

Lets plot a unit circle too

$$\text{un} := 0..40$$

$$\text{ImagUnitCirc}_{\text{un}} := \sin\left(2 \cdot \pi \cdot \frac{\text{un}}{40}\right) \quad \text{RealUnitCirc}_{\text{un}} := \cos\left(2 \cdot \pi \cdot \frac{\text{un}}{40}\right)$$

$$\text{pmax} := \text{round}(|\text{Hpoles}| + 1)$$



Q6. Is the Prony approximation stable **and** minimum phase for $g[n] = \{ 1, 2, 3, 4, 3, 2 \}$?

Part 6 Frequency response of the Prony solution

Finally, we have $H_{\text{bilin}}(z) = B(z)/N(z)$

$$H_{\text{prony}}(\text{num}, \text{den}, z) := \frac{\sum_{k=0}^{\text{rows}(\text{num})-1} \text{num}_k \cdot z^{-k}}{\sum_{k=0}^{\text{rows}(\text{den})-1} \text{den}_k \cdot z^{-k}}$$

So check a few important points:

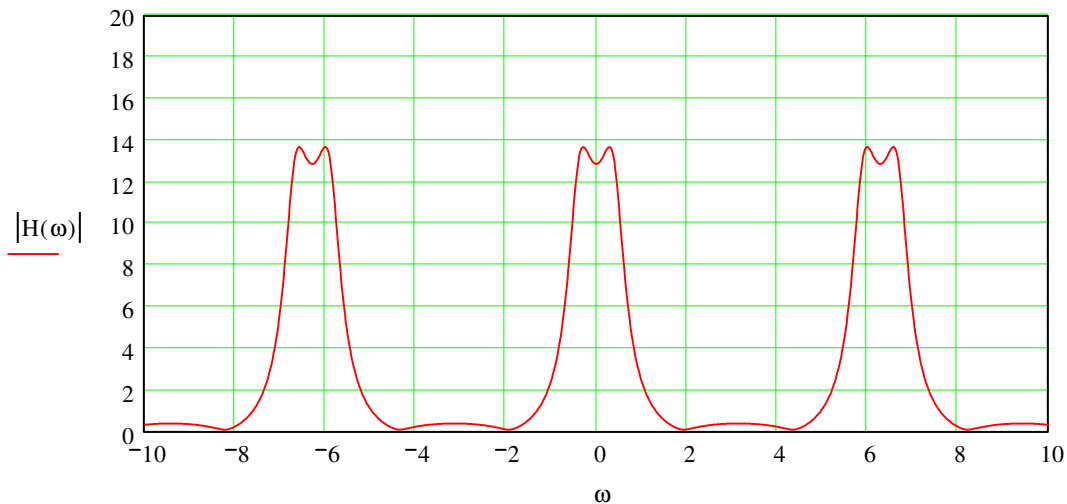
$$H_{\text{prony}}(b, a, e^{i \cdot 0.25}) = 11.052 - 7.984i$$

$$H_{\text{prony}}(b, a, -1) = 0.398$$

Plot the frequency response of the bilinear transform filter

$$H(\omega) := H_{\text{prony}}(b, a, e^{i \cdot \omega}) \quad a = \begin{pmatrix} 1 \\ -1.395 \\ 0.581 \end{pmatrix} \quad b = \begin{pmatrix} 1 \\ 0.605 \\ 0.791 \end{pmatrix} \quad E = (1.674)$$

$$H(0) = 12.875$$



Q7. Is the Prony approximation lowpass, highpass, or bandpass for $g[n] = \{ 1, 2, 3, 4, 3, 2 \}$?

See the last page for additional questions.

Appendix

Using Mathcad Minerr Block to solve for Prony H(z)

THIS MAY GIVE ERRONEOUS RESULTS AS SHOWN BELOW

Mathcad should be able to directly solve for the coefficients a_1, a_2, b_0, b_1, b_2 as follows using a solve block. Set the order of numerator as N_b , order of denominator as N_a where $a_0, a_1, \dots, a_{N_a}, b_0, b_1, b_2, \dots, b_{N_b}$

$$N_a := 2 \qquad N_b := 2$$

First define the vectors representing the numerator coefficients as b , and denominator coefficients as a . Also, load these vectors with initial values (i.e., 0) for the coefficients and define the desired output response $g[n]$

$$\begin{aligned} n_a &:= 0..N_a & n_b &:= 0..N_b \\ a_{n_a} &:= 0 & b_{n_b} &:= 0 \end{aligned}$$

Then, define the desired output response $g[n]$

$$g := \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 0 \end{pmatrix}$$

For the Pade method, the number of rows in the matrix was $\text{rows} = N_a + N_b + 1$, for the Prony method the system of equations must be overdetermined, so the Prony method requires $\text{rows} > N_a + N_b + 1$.

So define N_{cmin} as the minimum number of rows required in the matrix G .

$$\text{rows_in_G_min} := N_a + N_b + 2 \qquad \text{rows_in_G_min} = 6$$

Since the number of rows in vector g will equal the number of rows in matrix G , set vector g equal to zero to force an error in the rest of the program if the number of rows does not satisfy the requirement $\text{rows} > N_a + N_b + 1$

$$g := \text{if}(\text{rows}(g) < \text{rows_in_G_min}, 0, g)$$
$$g = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 0 \end{pmatrix}$$

Then set up the matrix G that corresponds to the convolution $g[n] * a[n]$ when G is multiplied by column vector a.

$$\begin{aligned} \text{grows} &:= \text{rows}(g) & \text{gcols} &:= \text{rows}(a) \\ \text{gr} &:= 0.. \text{grows} - 1 & \text{gc} &:= 0.. \text{gcols} - 1 \\ G_{\text{gr}, \text{gc}} &:= \text{if} \left[(\text{gr} - \text{gc} \geq 0) \wedge (\text{gr} - \text{gc} \leq \text{rows}(g) - 1), g_{\text{gr}-\text{gc}}, 0 \right] \end{aligned}$$

$$G = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \\ 4 & 3 & 2 \\ 5 & 4 & 3 \\ 0 & 5 & 4 \end{pmatrix}$$

Next set up the equations and use the Mathcad solve block to find the solutionn (Given is a keyword to start a solve block)

$$a1 := 0 \quad a2 := 0 \quad b0 := 0 \quad b1 := 0 \quad b2 := 0$$

Given

$$G \cdot \begin{pmatrix} 1 \\ a1 \\ a2 \end{pmatrix} = \begin{pmatrix} b0 \\ b1 \\ b2 \\ 0 \\ 0 \\ 0 \end{pmatrix} \quad \text{where} \quad G = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \\ 4 & 3 & 2 \\ 5 & 4 & 3 \\ 0 & 5 & 4 \end{pmatrix}$$

Note: if no solution exists using Find() below, see the Minerr() help page in the manual.

$$\text{prony_solution} := \text{Minerr}(a1, a2, b0, b1, b2)$$

In some editions of Matlab, this prony solution may be wrong, depending on how the Minerr algorithm is working.

$$\text{prony_solution} = \begin{pmatrix} -2 \\ 1 \\ 1 \\ 0 \\ 0 \end{pmatrix}$$

And place the solutions back in vectors a and b

$$a_{na} := \text{if}(na = 0, 1, \text{prony_solution}_{na-1})$$

$$a = \begin{pmatrix} 1 \\ -2 \\ 1 \end{pmatrix}$$

$$a_2 = 1 \times 10^0$$

$$b_{nb} := \text{prony_solution}_{nb+\text{rows}(a)-1}$$

$$b = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

and check the approximation

$$G \cdot a = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ -6 \end{pmatrix} \quad \begin{pmatrix} b_0 \\ b_1 \\ b_2 \\ 0 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

and the error vector e

$$ne := 0.. \text{rows}(G) - \text{rows}(b) - 1$$

$$\varepsilon_{ne} := \left[G \cdot a - \begin{pmatrix} b_0 \\ b_1 \\ b_2 \\ 0 \\ 0 \\ 0 \end{pmatrix} \right]_{ne+\text{rows}(b)} \quad \varepsilon = \begin{pmatrix} 0 \\ 0 \\ -6 \end{pmatrix}$$

In some editions of Matlab, this error is higher than the pseudo-inverse method solution

and the error E

$$E := (\bar{\varepsilon})^T \cdot \varepsilon$$

$$E = (36)$$

Q8-13. Do not change the results above. In the following pages, let $g[n] = \{ 1, 2, 3, 4, 3, 2, 1 \}$, repeat the Prony approximation for this new $g[n]$ below this page, and let $N_a = 3$ and $N_b = 2$. Do not change the foregoing system that was solved on the previous pages above this point.

Q8. Solve for the new vectors a and b determining $A(z)$ and $B(z)$

Q9. Find the new error E and error vector ε

Q10. Find the first 7 points of the impulse response and compare them to $g[n] = \{ 1, 2, 3, 4, 3, 2, 1 \}$ and compute the error from $g[n]$

Q11. Find and plot the poles and zeroes of the new Prony approximation

Q12. Find and plot $|H(\omega)|$ of the new Prony approximation

Q13. Is this new design stable and minimum phase?

$$g := \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 3 \\ 2 \\ 1 \end{pmatrix}$$