

Lecture 13.

Pole and Zero Analysis

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Outline

- **Computation of transfer function coefficients**
- **Discrete Fourier Transform**
- **Polynomial root finding algorithms**
 - **Muller's Method**
- **Implementation issues**
 - **Polynomial deflation**

AC Analysis & Pole / Zero

- **Pole-zero (PZ) analysis is based on small-signal.**
- **A DC operating point must be simulated first.**
- **A stamp formulation is similar to AC analysis, except for 's' is not replaced by 'j ω '.**
 - **That is, values of s other than 'j ω ' will be used in PZ analysis.**

AC Analysis

By AC stamping, the following equation is obtained:

$$A(s) X = bU(s)$$

Let the LU factorization of $A(s)$ be:

$$A(s) = L U;$$

Assuming the diagonal elements of L are all '1'.

$$\det A(s) = u_{11} \cdot u_{22} \cdots u_{nn}$$

The determinant of $A(s)$ will be used for PZ analysis.

Transfer Function

Define an output $Y(s) = c X(s) = c A^{-1}(s) b U(s)$.

$$Y(s) = cA^{-1}(s)bU(s) = \frac{c \cdot \text{adj}(A(s))b}{\det(A(s))} U(s) = \frac{N(s)}{D(s)} U(s)$$

Adj(A) is the adjoint matrix of **A**.

D(s) and **N(s)** are polynomials of **s**.

$$D(s) = \det A(s)$$

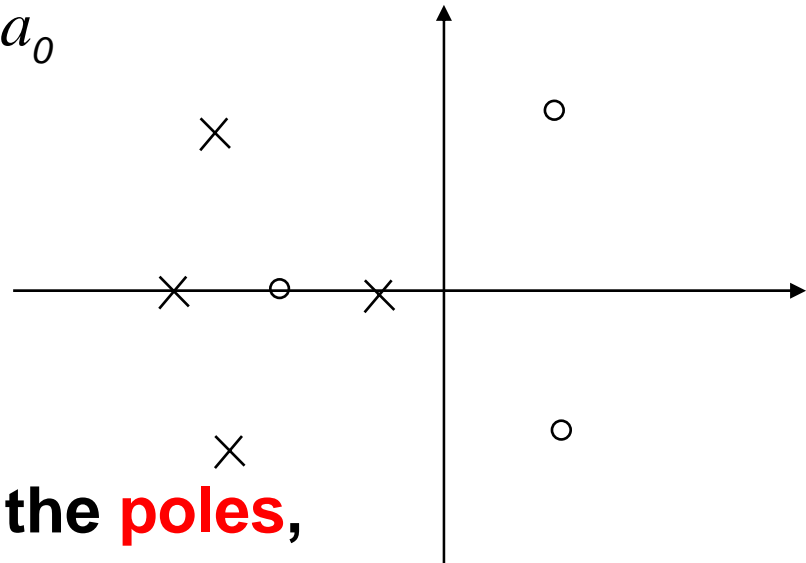
The transfer function is

$$H(s) = \frac{N(s)}{D(s)} = \frac{b_m s^m + \dots + b_1 s + b_0}{a_n s^n + \dots + a_1 s + a_0}$$

Factorized Form

- Writing the polynomials in factorized form

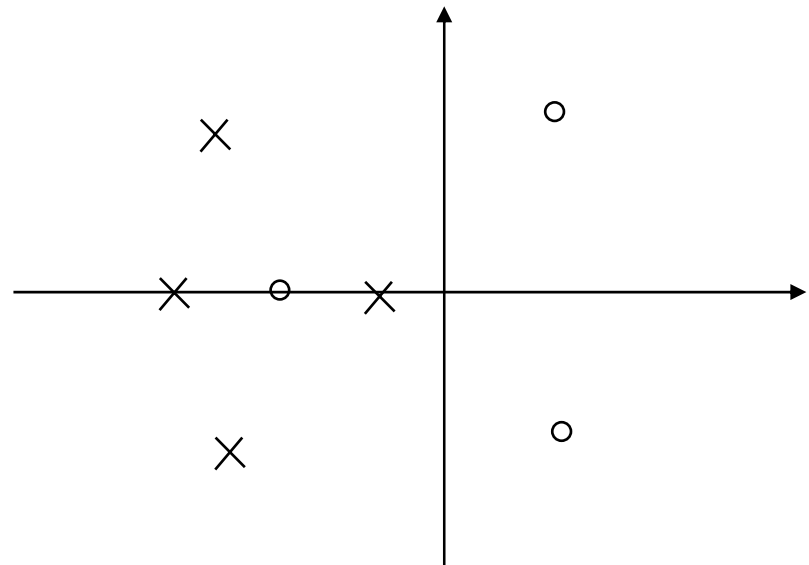
$$H(s) = \frac{N(s)}{D(s)} = \frac{b_m s^m + \dots + b_1 s + b_0}{a_n s^n + \dots + a_1 s + a_0}$$
$$= \left(\frac{b_m}{a_n} \right) \frac{(s - z_1) \cdots (s - z_m)}{(s - p_1) \cdots (s - p_n)}$$



z_i 's are the **zeros** and p_j 's are the **poles**,
suppose no z_i and p_j are canceled.

Conjugate Poles and Zeros

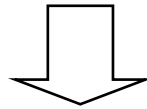
- A real circuit has a transfer function with real coefficients.
- Hence, the complex poles and zeros must exist in conjugate form.



Computation Flow

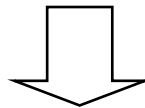
$$\mathbf{A}(s) \mathbf{X} = \mathbf{b} \mathbf{U}(s)$$

Step 1



$$H(s) = \frac{b_m s^m + \dots + b_1 s + b_0}{a_n s^n + \dots + a_1 s + a_0}$$

Step 2



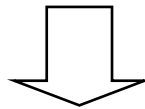
$$H(s) = \left(\frac{b_m}{a_n} \right) \frac{(s - z_1) \cdots (s - z_m)}{(s - p_1) \cdots (s - p_n)}$$

Computation Flow (Step 1)

- In Step 1: We compute the coefficients b_i and a_i .

$$A(s) X = bU(s)$$

Step 1



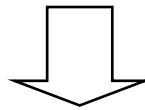
$$H(s) = \frac{b_m s^m + \cdots + b_1 s + b_0}{a_n s^n + \cdots + a_1 s + a_0}$$

Computation Flow (Step 2)

- In Step 2: We compute the poles, zeros, and the dc gain.

$$H(s) = \frac{b_m s^m + \dots + b_1 s + b_0}{a_n s^n + \dots + a_1 s + a_0}$$

Step 2



$$H(s) = \left(\frac{b_m}{a_n} \right) \frac{(s - z_1) \cdots (s - z_m)}{(s - p_1) \cdots (s - p_n)}$$

Computation of Step 1

- We first compute all \mathbf{a}_k .
- Then compute all \mathbf{b}_i .
- The method is by “*sampling*” and “*fitting*”.

$$H(s) = \frac{N(s)}{D(s)} = \frac{b_m s^m + \dots + b_1 s + b_0}{a_n s^n + \dots + a_1 s + a_0}$$

$$D(s) = \det A(s)$$

$$D(s) = a_n s^n + \dots + a_1 s + a_0 = \det A(s)$$

$$N(s) = b_m s^m + \dots + b_1 s + b_0 = D(s)H(s) \quad H(s) = \left[cA^{-1}(s)b \right]$$

Sampling 's'

- The following eqn holds for all s.
- Hence, we can take some special samples of 's' to solve all a_k 's.

$$D(s) = a_n s^n + \dots + a_1 s + a_0 = \det A(s)$$

Take (n+1) samples of s: $s_1, s_2, \dots, s_k, s_{k+1}$.

$$\begin{bmatrix} 1 & s_1 & s_1^2 & \dots & s_1^n \\ 1 & s_2 & s_2^2 & \dots & s_2^n \\ 1 & s_3 & s_3^2 & \dots & s_3^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & s_{n+1} & s_{n+1}^2 & \dots & s_{n+1}^n \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix} = \begin{bmatrix} d_1 \\ d_2 \\ d_3 \\ \vdots \\ d_{d_1} \end{bmatrix} \quad d_i = D(s_i)$$

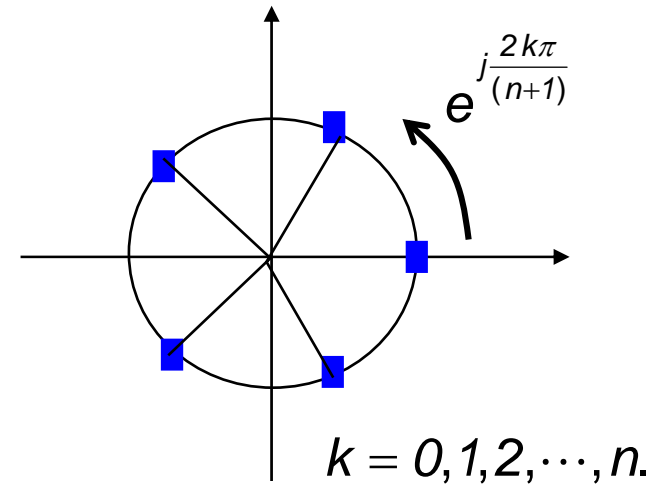
Numerical Stability

- We have to be careful on the samples.
- The coefficient matrix might be ill conditioned if s is not sampled properly.
- We'll **sample s** from the “**unit circle**” (best).

$$\begin{bmatrix} 1 & s_1 & s_1^2 & \cdots & s_1^n \\ 1 & s_2 & s_2^2 & \cdots & s_2^n \\ 1 & s_3 & s_3^2 & \cdots & s_3^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & s_{n+1} & s_{n+1}^2 & \cdots & s_{n+1}^n \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix} = \begin{bmatrix} d_1 \\ d_2 \\ d_3 \\ \vdots \\ d_{d_1} \end{bmatrix}$$

Unit-Circle Sampling

- Sampling on the unit circle keep the norm $|s_i| = 1$.
- Uniformly spaced samples s_i^k make a **Discrete Fourier Transform** of the coefficients a_k .



DFT of (a_0, \dots, a_n)

$$a_0 + a_1 \left(e^{j\frac{2\pi}{n+1}} \right) + a_2 \left(e^{j\frac{4\pi}{n+1}} \right) + \dots + a_n \left(e^{j\frac{2n\pi}{n+1}} \right) = d_k$$

Inverse DFT

- The coefficients a_k can be obtained by Inverse DFT.

$$a_0 + a_1 \left(e^{j\frac{2\pi}{n+1}} \right) + a_2 \left(e^{j\frac{4\pi}{n+1}} \right) + \dots + a_n \left(e^{j\frac{2n\pi}{n+1}} \right) = d_k$$

$$a_k = \frac{1}{n+1} \sum_{i=0}^n d_i e^{-j\frac{2\pi ik}{n+1}}$$

(IDFT)

This is the best numerical algorithm with the least numerical error.

Order 'n'

- The polynomial order 'n' can be determined from the circuit: $\#L + \#C$
 - Count the C's and L's in circuit.
- n is usually less than the order of the MNA matrix.
- **An overestimate of 'n' also is OK.**
 - The IDFT algorithm will calculate very small coefficient values for a_i , $i > n$, which can be removed.

Computing the b_k 's

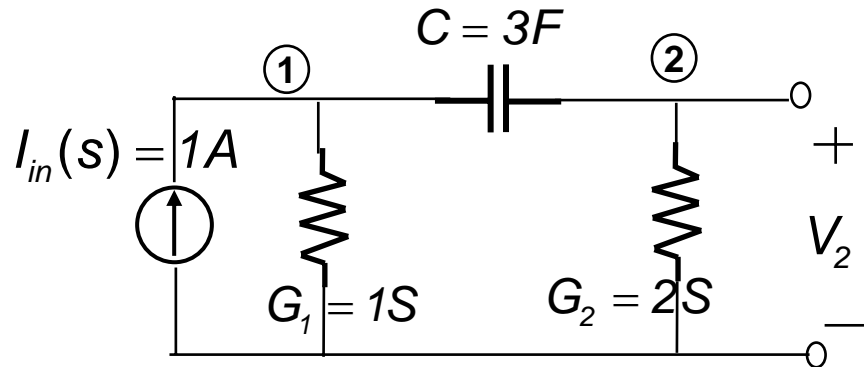
$$N(s_k) = b_m s_k^n + \cdots + b_1 s_k + b_0 = D(s_k)H(s_k)$$

- We have to **solve the output at $s = s_k$** , which gives $H(s_k)$
 - Assuming the input is a unit impulse: $U(s) = 1$.
- Then apply the same **IDFT** to compute b_k 's.

Example 1

AC formulation:

$$\begin{bmatrix} G_1 + Cs & -Cs \\ -Cs & G_2 + Cs \end{bmatrix} \begin{bmatrix} V_1(s) \\ V_2(s) \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$



There is only one C, we know the degree $n = 1$.

We'll determine the coefficients of $H(s)$.

$$H(s) = \frac{b_0 + b_1s}{a_0 + a_1s}$$

$$\begin{bmatrix} 1 + 3s & -3s \\ -3s & 2 + 3s \end{bmatrix} \begin{bmatrix} V_1(s) \\ V_2(s) \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

Example 1 (cont'd)

s will be sampled at (+1) and (-1).

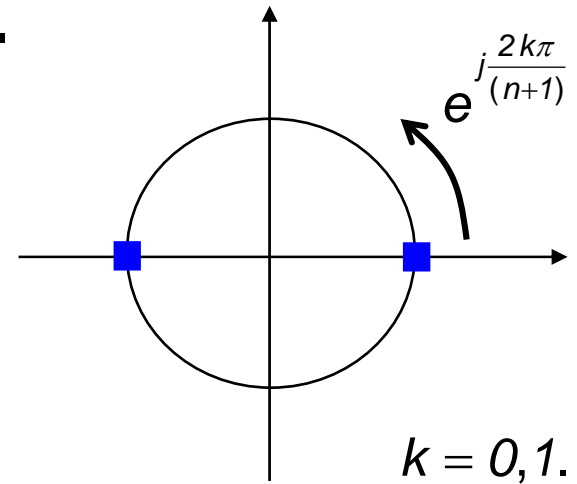
$$s_0 = 1; \quad s_1 = -1.$$

LU factorize for s_0 and s_1 :

$$\begin{bmatrix} 1+3s & -3s \\ -3s & 2+3s \end{bmatrix} \begin{bmatrix} V_1(s) \\ V_2(s) \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$s_0 = 1 \quad \Rightarrow \quad \begin{bmatrix} 4 & -3 \\ -3 & 5 \end{bmatrix} \begin{bmatrix} V_1(s) \\ V_2(s) \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$s_1 = -1 \quad \Rightarrow \quad \begin{bmatrix} -2 & 3 \\ 3 & -1 \end{bmatrix} \begin{bmatrix} V_1(s) \\ V_2(s) \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$



Example 1 (cont'd)

LU factorizations:

V_2 is output.

$$s_0 = 1 \Rightarrow \begin{bmatrix} 4 & -3 \\ -3 & 5 \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 0 \\ -\frac{3}{4} & 1 \end{bmatrix} \begin{bmatrix} 4 & -3 \\ 0 & \frac{11}{4} \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\Rightarrow D(1) = 11; V_2(1) = \frac{3}{11}; \Rightarrow N(1) = D(1)V_2(1) = 11 \times \frac{3}{11} = 3$$

$$s_1 = -1 \Rightarrow \begin{bmatrix} -2 & 3 \\ 3 & -1 \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 0 \\ -\frac{3}{2} & 1 \end{bmatrix} \begin{bmatrix} -2 & 3 \\ 0 & \frac{7}{2} \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\Rightarrow D(-1) = -7; V_2(-1) = \frac{3}{7}; \Rightarrow N(-1) = D(-1)V_2(-1) = (-7) \times \frac{3}{7} = -3$$

Example 1 (cont'd)

$$s_0 = 1$$

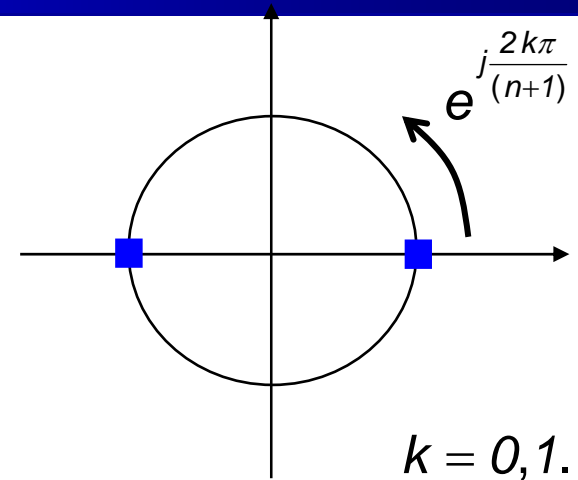
$$d_0 = D(s_0) = 11;$$

$$n_0 = N(s_0) = 3;$$

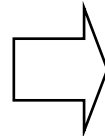
$$s_1 = -1$$

$$d_1 = D(s_1) = -7;$$

$$n_1 = N(s_1) = -3;$$

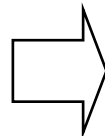


$$\text{DFT}(\mathbf{a}_0, \mathbf{a}_1) = (\mathbf{d}_0, \mathbf{d}_1)$$



$$(\mathbf{a}_0, \mathbf{a}_1) = \text{IDFT}(\mathbf{d}_0, \mathbf{d}_1)$$

$$\text{DFT}(\mathbf{b}_0, \mathbf{b}_1) = (\mathbf{n}_0, \mathbf{n}_1)$$



$$(\mathbf{b}_0, \mathbf{b}_1) = \text{IDFT}(\mathbf{n}_0, \mathbf{n}_1)$$

Example 1 (cont'd)

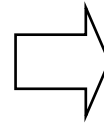
$$a_k = \frac{1}{2} \left(d_0 e^{-j0} + d_1 e^{-j\frac{2\pi k}{2}} \right); \quad k = 0, 1$$

(IDFT)

$$H(s) = \frac{3s}{2 + 9s}$$

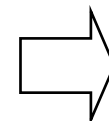
$$d_0 = D(s_0) = 11; \quad d_1 = D(s_1) = -7;$$

$$(a_0, a_1) = \text{IDFT}(d_0, d_1)$$



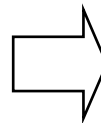
$$\begin{cases} a_0 = \frac{1}{2}(11 - 7) = 2 \\ a_1 = \frac{1}{2}(11 + 7) = 9 \end{cases}$$

$$n_0 = N(s_0) = 3; \quad n_1 = N(s_1) = -3;$$

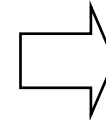


$$D(s) = 2 + 9s;$$

$$(b_0, b_1) = \text{IDFT}(n_0, n_1)$$

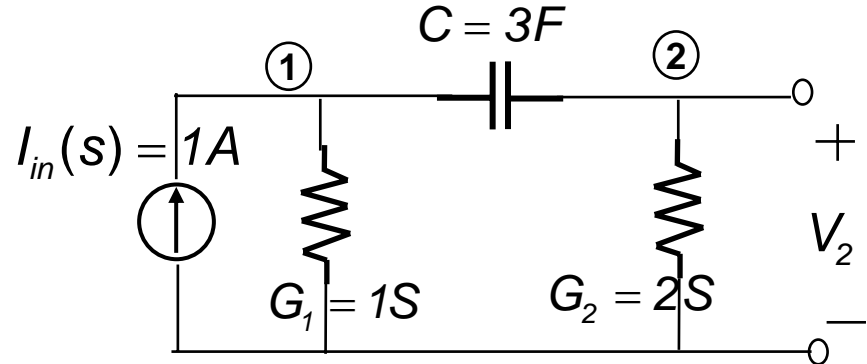


$$\begin{cases} b_0 = \frac{1}{2}(3 - 3) = 0 \\ b_1 = \frac{1}{2}(3 + 3) = 3 \end{cases}$$



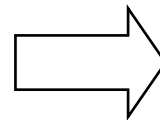
$$N(s) = 3s;$$

Check



The symbolic transfer function is

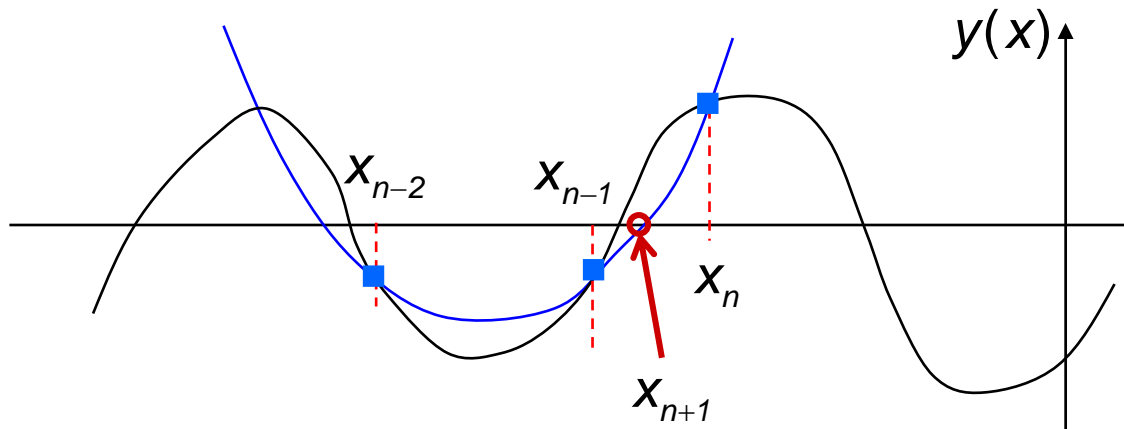
$$H(s) = \frac{Cs}{G_1G_2 + (G_1 + G_2)Cs}$$



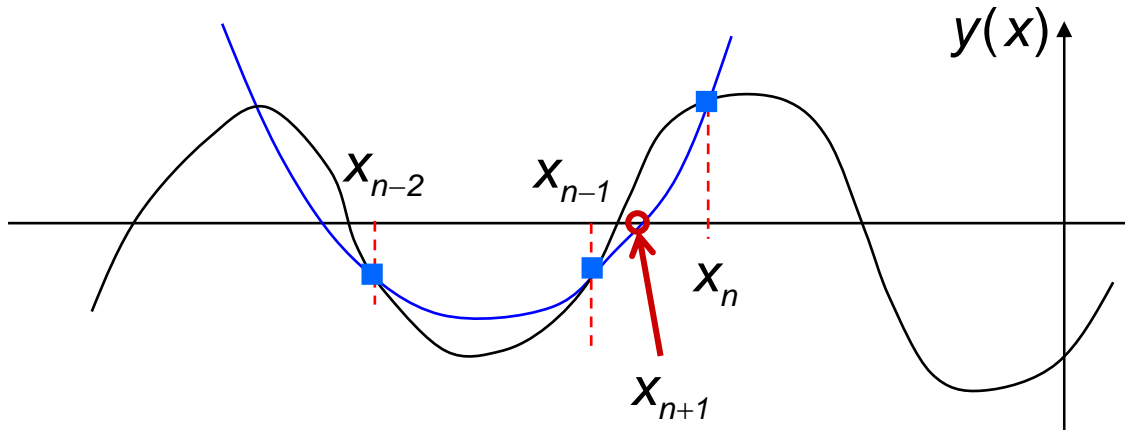
$$H(s) = \frac{3s}{2 + 9s}$$

Muller's Method

- Muller's method is an iteration algorithm using quadratic polynomial interpolation.
- Used in Berkeley Spice3f5 and HSPICE.



Muller's Method



Given three points on the polynomial $y(x) = 0$:

(x_k, y_k) , $k = n-2, n-1, n$.

Find a parabola $h(x)$ passing the three points.

Find a root of $h(x) = 0$ as the next root estimate.

Muller's Method


- Given three previous guesses for the root x_{n-2} , x_{n-1} , x_n , and their polynomial values y_{n-2} , y_{n-1} , y_n , the next approximation x_{n+1} is produced by the following formulas:

$$q = \frac{x_i - x_{i-1}}{x_{i-1} - x_{i-2}}$$

$$a = qy_n - q(q + 1)y_{n-1} + q^2y_{n-2};$$

$$b = (2q + 1)y_n - (q + 1)^2y_{n-1} + q^2y_{n-2};$$

$$c = (q + 1)y_n;$$

$$x_{n+1} = x_n - (x_n - x_{n-1}) \frac{2c}{b \pm \sqrt{b^2 - 4ac}}$$


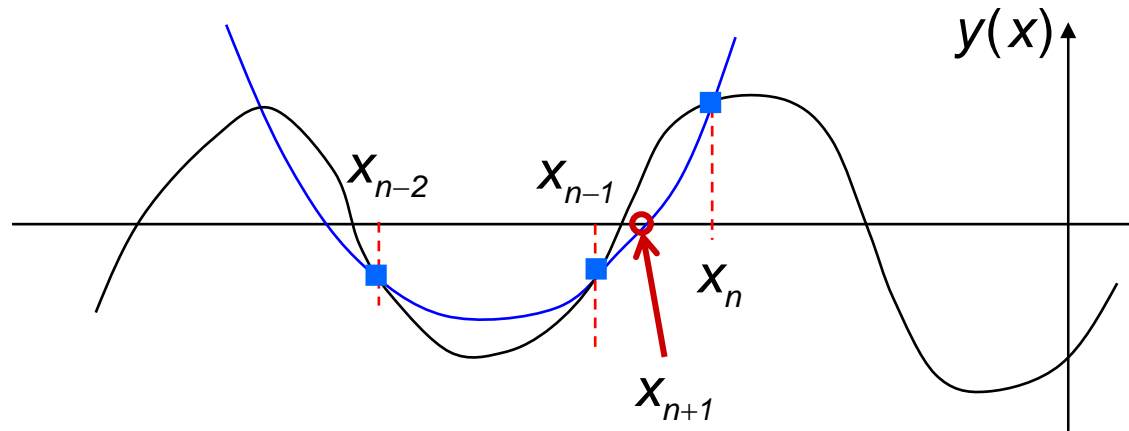
Choose +/- to make the modulus larger.

$$\left| b \pm \sqrt{b^2 - 4ac} \right|$$

Advantages of Muller's Method

- Muller's method solves roots of a quadratic function, it can find **complex pairs of roots**.
- You may start the iteration with any three values of x , e.g., three equally spaced values on the x axis.
- The denominator could possibly be a complex number, hence your implementation must use complex number arithmetic.

Derivation of Muller's Method



The Lagrange interpolation polynomial is

$$L(x) = y_n \frac{(x - x_{n-2})(x - x_{n-1})}{(x_n - x_{n-2})(x_n - x_{n-1})} + y_{n-1} \frac{(x - x_{n-2})(x - x_n)}{(x_{n-1} - x_{n-2})(x_{n-1} - x_n)} + y_{n-2} \frac{(x - x_{n-1})(x - x_n)}{(x_{n-2} - x_{n-1})(x_{n-2} - x_n)}$$

This polynomial is a parabola passing the three points.

Derivation (cont'd)

$$L(x) = y_n \frac{(x - x_{n-2})(x - x_{n-1})}{(x_n - x_{n-2})(x_n - x_{n-1})} + y_{n-1} \frac{(x - x_{n-2})(x - x_n)}{(x_{n-1} - x_{n-2})(x_{n-1} - x_n)} + y_{n-2} \frac{(x - x_{n-1})(x - x_n)}{(x_{n-2} - x_{n-1})(x_{n-2} - x_n)}$$

$$L(x) = C + B(x - x_n) + A(x - x_n)^2 \quad \text{another form of } L(x)$$

Let's determine A, B, C.

$$L(x_n) = C = y_n$$

$$L'(x_n) = B = y_n \frac{(x_n - x_{n-2}) + (x_n - x_{n-1})}{(x_n - x_{n-2})(x_n - x_{n-1})} + y_{n-1} \frac{(x_n - x_{n-2})}{(x_{n-1} - x_{n-2})(x_{n-1} - x_n)} + y_{n-2} \frac{(x_n - x_{n-1})}{(x_{n-2} - x_{n-1})(x_{n-2} - x_n)}$$

$$\frac{1}{2}L''(x) = A = y_n \frac{1}{(x_n - x_{n-2})(x_n - x_{n-1})} + y_{n-1} \frac{1}{(x_{n-1} - x_{n-2})(x_{n-1} - x_n)} + y_{n-2} \frac{1}{(x_{n-2} - x_{n-1})(x_{n-2} - x_n)}$$

Derivation (cont'd)

Define

$$q = \frac{x_n - x_{n-1}}{x_{n-1} - x_{n-2}} \quad \Rightarrow \quad q+1 = \frac{x_n - x_{n-2}}{x_{n-1} - x_{n-2}}$$

$$\begin{aligned}
 B &= y_n \frac{(x_n - x_{n-2}) + (x_n - x_{n-1})}{(x_n - x_{n-2})(x_n - x_{n-1})} + y_{n-1} \frac{(x_n - x_{n-2})}{(x_{n-1} - x_{n-2})(x_{n-1} - x_n)} + y_{n-2} \frac{(x_n - x_{n-1})}{(x_{n-2} - x_{n-1})(x_{n-2} - x_n)} \\
 &= \left[\underbrace{y_n \frac{(x_n - x_{n-2}) + (x_n - x_{n-1})}{(x_n - x_{n-2})}}_{1 + \frac{q}{q+1} = \frac{2q+1}{q+1}} - \underbrace{y_{n-1} \frac{(x_n - x_{n-2})}{(x_{n-1} - x_{n-2})}}_{-(q+1)} + \underbrace{y_{n-2} \frac{(x_n - x_{n-1})^2}{(x_{n-2} - x_{n-1})(x_{n-2} - x_n)}}_{\frac{q^2}{q+1}} \right] \frac{1}{(x_n - x_{n-1})}
 \end{aligned}$$

$$B = \left[y_n \frac{2q+1}{q+1} - y_{n-1} \frac{(q+1)^2}{(q+1)} + y_{n-2} \frac{q^2}{(q+1)} \right] \frac{1}{(x_n - x_{n-1})}$$

Derivation (cont'd)

$$\begin{aligned} A &= y_n \frac{1}{(x_n - x_{n-2})(x_n - x_{n-1})} + y_{n-1} \frac{1}{(x_{n-1} - x_{n-2})(x_{n-1} - x_n)} + y_{n-2} \frac{1}{(x_{n-2} - x_{n-1})(x_{n-2} - x_n)} \\ &= \left[\underbrace{y_n \frac{(x_n - x_{n-1})}{(x_n - x_{n-2})}}_{\frac{q}{q+1}} - \underbrace{y_{n-1} \frac{(x_n - x_{n-1})}{(x_{n-1} - x_{n-2})}}_{-q} + \underbrace{y_{n-2} \frac{(x_n - x_{n-1})^2}{(x_{n-2} - x_{n-1})(x_{n-2} - x_n)}}_{\frac{q^2}{q+1}} \right] \frac{1}{(x_n - x_{n-1})^2} \end{aligned}$$

$$A = \left[y_n \frac{q}{(q+1)} - y_{n-1} \frac{q(q+1)}{(q+1)} + y_{n-2} \frac{q^2}{(q+1)} \right] \frac{1}{(x_n - x_{n-1})^2}$$

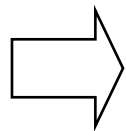
Derivation (cont'd)

$$C = y_n = \frac{c}{(q+1)}$$

$$B = \left[y_n \frac{2q+1}{q+1} - y_{n-1} \frac{(q+1)^2}{(q+1)} + y_{n-2} \frac{q^2}{(q+1)} \right] \frac{1}{(x_n - x_{n-1})} = \frac{b}{(q+1)(x_n - x_{n-1})}$$

$$A = \left[y_n \frac{q}{(q+1)} - y_{n-1} \frac{q(q+1)}{(q+1)} + y_{n-2} \frac{q^2}{(q+1)} \right] \frac{1}{(x_n - x_{n-1})^2} = \frac{a}{(q+1)(x_n - x_{n-1})^2}$$

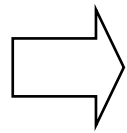
$$L(x) = C + B(x - x_n) + A(x - x_n)^2 = 0$$



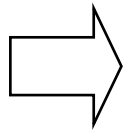
$$c + b \frac{(x - x_n)}{(x_n - x_{n-1})} + a \frac{(x - x_n)^2}{(x_n - x_{n-1})^2} = 0$$

Derivation (cont'd)

$$c + b \frac{(x - x_n)}{(x_n - x_{n-1})} + a \frac{(x - x_n)^2}{(x_n - x_{n-1})^2} = 0$$



$$\frac{(x_{n+1} - x_n)}{(x_n - x_{n-1})} = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a} = -\frac{2c}{b \pm \sqrt{b^2 - 4ac}}$$



$$x_{n+1} = x_n - (x_n - x_{n-1}) \frac{2c}{b \pm \sqrt{b^2 - 4ac}}$$

(Done)

Deflation of Polynomial

- When a root r is found, the polynomial $P(s)$ can be factorized as $P(s) = (s-r)Q(s)$.
- $Q(s)$ is a reduced-order polynomial.
- It avoids finding again the same root already found.
- **Deflation** is just **polynomial division**.
- In the case of conjugate roots $(a \pm jb)$, we can deflate by a quadratic factor:

$$s^2 - 2as + (a^2 + b^2)$$

More on Deflation

- Deflation must be used with care.
- Each new root is known with finite accuracy, error accumulates.
- The **coefficients** of the deflated polynomial $Q(x)$ are **not accurate**.
- The roots can become increasingly inaccurate.

Forward / Backward Deflation

$$P(s) = a_n s^n + a_{n-1} s^{n-1} + \cdots + a_1 s + a_0$$

$$P(s) = (s - p)Q(s)$$

$$Q(s) = q_{n-1} s^{n-1} + q_{n-2} s^{n-2} + \cdots + q_1 s + q_0$$

Stable deflation strategies:

- **Forward Deflation** – Compute from q_{n-1} down to q_0 if the **smallest $|p|$** is divided out at each stage.
- **Backward Deflation** – Compute from q_0 up to q_{n-1} if the **largest $|p|$** is divided out at each stage.

Stable Deflation

$$\begin{array}{r}
 P(s) = a_n s^n + \boxed{a_{n-1} s^{n-1}} + \cdots + \boxed{a_2 s^2} + \boxed{a_1 s} + a_0 \\
 \hline
 sQ(s) = q_{n-1} s^n + q_{n-2} s^{n-1} + \cdots + \boxed{q_1 s^2} + \boxed{q_0 s} \\
 pQ(s) = \boxed{pq_{n-1} s^{n-1}} + \cdots + \boxed{pq_2 s^2} + \boxed{pq_1 s} + pq_0
 \end{array}$$

$$P(s) = (s - p)Q(s)$$

Forward deflation

(|p| smallest)



$$a_n = q_{n-1}$$

$$a_{n-1} = q_{n-2} - pq_{n-1}$$

⋮

$$a_1 = q_0 - pq_1$$

$$a_0 = -pq_0$$

Backward deflation

(|p| largest)



Quadratic Deflation

$$\begin{array}{r}
 P(s) = a_n s^n + \boxed{a_{n-1} s^{n-1} + a_{n-2} s^{n-2}} + \cdots + a_2 s^2 + \boxed{a_1 s} + a_0 \\
 \hline
 s^2 Q(s) = q_{n-2} s^n + q_{n-3} s^{n-1} + q_{n-4} s^{n-2} + \cdots + q_0 s^2 \\
 \alpha s Q(s) = \boxed{\alpha q_{n-2} s^{n-1} + \alpha q_{n-3} s^{n-2}} + \cdots + \alpha q_1 s^2 + \alpha q_0 s \\
 \beta Q(s) = \boxed{\beta q_{n-2} s^{n-2}} + \cdots + \beta q_2 s^2 + \boxed{\beta q_1 s} + \beta q_0
 \end{array}$$

$$P(s) = (s^2 + \alpha s + \beta)Q(s)$$

$$Q(s) = q_{n-2} s^{n-2} + \cdots + q_1 s + q_0$$

Quadratic Deflation

$$a_n = q_{n-2}$$

$$a_{n-1} = q_{n-3} + \alpha q_{n-2}$$

$$a_{n-2} = q_{n-4} + \alpha q_{n-3} + \beta q_{n-2}$$

⋮

$$a_2 = q_0 + \alpha q_1 + \beta q_2$$

$$a_1 = \alpha q_0 + \beta q_1$$

$$a_0 = \beta q_0$$

Forward deflation

($|p|$ smallest)

Backward deflation

($|p|$ largest)

Reference

- **J. Vlach and K. Singhal, Computer Methods for Circuit Analysis and Design, Van Nostrand Reinhold Company, 1983. (Chapter 7)**