

Chapter 3 Data Acquisition and Manipulation

In this chapter we introduce z - *transform*, or the discrete Laplace Transform, to solve linear recursions.

- Section 3.1 z-transform

- Given a data stream $x = \{x_0, x_1, x_2, \dots\}$, let

$$X = Z(x) = \sum_{k=0}^{\infty} \frac{x_k}{z^k}$$

- $x \mapsto X$ is called a z - *transform*

- Example: $x = \{a, a, a, \dots\}$, $X = az(z - 1)^{-1}$

- Example: $x = \{1, a, a^2, a^3, \dots\}$, $X = z(z - a)^{-1}$

- Data x_i could be a complex number.

- For instance, the Sine wave $x(t) = \sin \omega t$ is sampled every T seconds to yield the signals

$$x_k = \sin(\omega T k) = \text{Im}(e^{k\omega T i}), \quad k = 0, 1, 2, \dots$$

So by the Euler formula

$$\begin{aligned}
 X &= \sum_{k=0}^{\infty} \frac{\sin(\omega T k)}{z^k} = \sum_{k=0}^{\infty} \frac{e^{k\omega T i} - e^{-k\omega T i}}{2i z^k} \\
 &= \frac{1}{2i} \sum_{k=0}^{\infty} \left(\frac{e^{\omega T i}}{z}\right)^k - \frac{1}{2i} \sum_{k=0}^{\infty} \left(\frac{e^{-\omega T i}}{z}\right)^k \\
 &= \frac{1}{2i} \left(1 - \frac{e^{\omega T i}}{z}\right)^{-1} - \frac{1}{2i} \left(1 - \frac{e^{-\omega T i}}{z}\right)^{-1} \\
 &= \frac{1}{2i} \frac{z}{z - e^{\omega T i}} - \frac{1}{2i} \frac{z}{z - e^{-\omega T i}} \\
 &= \frac{z \sin \omega T}{z^2 - 2z \cos \omega T + 1}
 \end{aligned}$$

– Linearity Property:

$$Z(ax + bx) = aZ(x) + bZ(x)$$

– Delay of x : $x_{(-1)} = \{0, x_0, x_1, \dots\}$

$$X_{-1} = Z(x_{(-1)}) = \sum_{k=1}^{\infty} \frac{x_{k-1}}{z^k} = \frac{X}{z}$$

– Delay of $x_{(-1)} : x_{(-2)} = \{0, 0, x_0, x_1, \dots\} = (x_{(-1)})_{(-1)}$

$$X_{-2} = Z(x_{(-2)}) = \frac{1}{z} Z(x_{(-1)}) = \frac{X}{z^2}$$

– In general

$$Z(x_{(-j)}) = \frac{X}{z^j} \quad (1)$$

– Forward of $x : x_{(+1)} = \{x_1, x_2, \dots\}$

$$X_{+1} = Z(x_{(+1)}) = \sum_{k=0}^{\infty} \frac{x_{k+1}}{z^k} = z \sum_{k=0}^{\infty} \frac{x_{k+1}}{z^{k+1}} = z \sum_{k=0}^{\infty} \frac{x_k}{z^k} - x_0 z = z(X - x_0) = zX - x_0 z$$

In general for k th forward of $x : x_{(+k)}$, its Z - transform $X_{+k} = Z(x_{(+k)})$

$$X_{+k} = z^k X - x_0 z^k - x_1 z^{k-1} - x_2 z^{k-2} - \dots - x_{k-1} z$$

– For any sequence $x = \{x_k\}_{k=0}$, for convenience we may write

$$Z(x_k) = Z(x).$$

– For a shift $\{x_{k-1}\}_{k=0} = \{x_{-1}, x_0, x_1, x_2, \dots\}$ of x with given initial value x_{-1} , we write

$$\begin{aligned} X_{k-1} &= Z(x_{k-1}) = Z(\{x_{k-1}\}_{k=0}) \\ &= x_{-1} + \frac{x_0}{z} + \frac{x_1}{z^2} + \dots \\ &= x_{-1} + \frac{1}{z} \left(x_0 + \frac{x_1}{z} + \dots \right) \\ &= x_{-1} + \frac{Z(x)}{z} \end{aligned}$$

– For a p th shift $\{x_{k-p}\}_{k=0} = \{x_{-p}, x_{-p+1}, \dots, x_{-1}, x_0, x_1, x_2, \dots\}$ with given initial values $x_{-p}, x_{-p+1}, \dots, x_{-1}$, we have accordingly

$$\begin{aligned} X_{k-p} &= Z(x_{k-p}) = Z(\{x_{k-p}\}_{k=0}) \\ &= x_{-p} + \frac{x_{-p+1}}{z} + \dots + \frac{x_{-1}}{z^{p-1}} + \frac{x_0}{z^p} + \frac{x_1}{z^{p+1}} + \dots \\ &= x_{-p} + \frac{x_{-p+1}}{z} + \dots + \frac{x_{-1}}{z^{p-1}} + \frac{1}{z^p} \left(x_0 + \frac{x_1}{z} + \dots \right) \\ &= x_{-p} + \frac{x_{-p+1}}{z} + \dots + \frac{x_{-1}}{z^{p-1}} + \frac{1}{z^p} Z(x) \end{aligned} \tag{2}$$

– For instance, If $y_k = ax_{k-1} + bx_{k-2}$, and

$$\begin{aligned}
 Y &= Z(y_k) = Z(ax_{k-1} + bx_{k-2}) = aZ(x_{k-1}) + bZ(x_{k-2}) \\
 &= a\left(x_{-1} + \frac{Z(x_k)}{z}\right) + b\left(x_{-2} + \frac{x_{-1}}{z} + \frac{Z(x_k)}{z^2}\right) \\
 &= ax_{-1} + bx_{-2} + \frac{bx_{-1}}{z} + \left(\frac{a}{z} + \frac{b}{z^2}\right)Z(x)
 \end{aligned} \tag{3}$$

- Section 3.2 Linear Recursions

– Consider the equation for all $k = n, n + 1, \dots$

$$x_k = a_1x_{k-1} + a_2x_{k-2} + \dots + a_nx_{k-n}$$

where a_k are fixed constant. $x_{-1}, x_{-2}, \dots, x_{-n}$ are given initially, and are called initial data.

– We can use z – transform to solve x by using the above calculation:

(i) Apply z – transform to both sides.

(ii) note that $Z(x_{k-p})$ is given by (2)

(iii) Solve $X = Z(x)$

(iv) Recover x_k

– Example (page 34) Solve

$$x_k = x_{k-1} + 2x_{k-2}, \quad x_{-1} = 1/2, \quad x_{-2} = -1/4$$

Solution: By (3) with $a = 1, b = 2$, we have

$$\begin{aligned} Z(x_k) &= ax_{-1} + bx_{-2} + \frac{bx_{-1}}{z} + \left(\frac{a}{z} + \frac{b}{z^2}\right) Z(x_k) \\ &= \frac{1}{z} + \left(\frac{1}{z} + \frac{2}{z^2}\right) Z(x_k) \end{aligned}$$

Set $w = 1/z$, then

$$\begin{aligned} Z(x_k) &= \frac{1}{z} \left(1 - \frac{1}{z} - \frac{2}{z^2}\right)^{-1} = \frac{w}{1 - w - 2w^2} = \frac{w}{(1 - 2w)(1 + w)} \\ &= \frac{1}{3(1 - 2w)} - \frac{1}{3(1 + w)} \\ &= \frac{1}{3} \sum_{k=0}^{\infty} (2w)^k - \frac{1}{3} \sum_{k=0}^{\infty} (-w)^k = \frac{1}{3} \sum_{k=0}^{\infty} [2^k - (-1)^k] w^k \\ &= \sum_{k=0}^{\infty} \frac{2^k - (-1)^k}{3} \frac{1}{z^k} \end{aligned}$$

Ans:

$$x_k = \frac{2^k - (-1)^k}{3} \text{ for } k > 0$$

– Problem in page 35: $x_k = ax_{k-1} + b$, for $k = 1, 2, \dots$, $x_0 = 0.7\%$. Find x_k . There are two ways to solve it.

* (1) Set $y_k = x_{k+1}$. So $y_{-1} = x_0 = 0.7\%$ is given, and $y_k = ay_{k-1} + b$ for $k = 0, 1, \dots$. Use the relation we derived earlier,

$$Y = aY_{-1} + Z(b) = a(z^{-1}Y + y_{-1}) + \frac{bz}{z-1}$$

$$(1 - az^{-1})Y = ay_{-1} + \frac{bz}{z-1}$$

$$\begin{aligned} Y &= \frac{ay_{-1}}{1 - az^{-1}} + \frac{bz}{(z-1)(1 - az^{-1})} = \frac{(b + ay_{-1})z^2 - ay_{-1}z}{(z-1)(z-a)} \\ &= (b + ax_0) + \frac{A}{(z-1)} + \frac{B}{(z-a)} \end{aligned}$$

* (2) From $x_0 = ax_{-1} + b$, we solve $x_{-1} = (x_0 - b)/z$. Then, we proceed to solve

$$\begin{aligned} X &= \frac{(b + ax_{-1})z^2 - ax_{-1}z}{(z-1)(z-a)} = \frac{x_0z^2 - ax_{-1}z}{(z-1)(z-a)} \\ &= x_0 + \frac{C}{(z-1)} + \frac{D}{(z-a)} \end{aligned}$$

- Convolution: For $x = \{x_k\}$, $h = \{h_k\}$, the discrete convolution product $y = h * x$ is defined as follows

$$y_k = h_0x_k + h_1x_{k-1} + \dots + h_kx_0 = \sum_{j=0}^k h_jx_{k-j} = \sum_{j=0}^k x_jh_{k-j}$$

- In the frequency domain, $Z(h * u) = Z(h)Z(u)$, or $Y = HU$. This is because the above expression indicate power series product. In fact

$$\begin{aligned} & \left(\sum_{k=0}^{\infty} a_k z^k \right) \left(\sum_{k=0}^{\infty} b_k z^k \right) \\ &= (a_0 + a_1z + a_2z^2 + \dots) (b_0 + b_1z + b_2z^2 + \dots) \\ &= a_0b_0 + (a_0b_1 + a_1b_0)z + (a_0b_2 + a_1b_1 + a_2b_0)z^2 + \dots \end{aligned}$$

- Examples: (1) for $\delta^{(0)} = \{1, 0, 0, 0, \dots\}$, $u * \delta^{(0)} = u$, (2) $\delta^{(1)} = \{0, 1, 0, 0, 0, \dots\} = \delta_{(-1)}$, $u * \delta^{(1)} = u_{(-1)}$, (3) $u * \delta^{(j)} = u_{(-j)}$

• Section 3.3 Filters

- A filter F is a device or algorithm that turns one stream of signals to another more useful one. We assume following three properties
 - F is linear: $F(au + bv) = aF(u) + bF(v)$
 - F is causal, i.e., the output depends on past and current inputs but not future inputs. In other

words, if $y = F(u)$, then

$$u_k = 0 \text{ for all } k < k_0 \text{ implies } y_k = 0 \text{ for all } k < k_0.$$

(iii) F is time invariant: if $y = F(u)$, then the same is true for delay shift, i.e.,

$$F \{u_{(-k_0)}\} = \{y_{(-k_0)}\}$$

In other words, if $F : u \rightarrow y$, then

$$F : \{u_k, u_{k+1}, \dots\} \rightarrow \{y_k, y_{k+1}, \dots\}$$

– Let $\delta = \{1, 0, 0, 0, \dots\}$ be the unit impulse, and $h = F(\delta) = \{h_0, h_1, \dots\}$. Then for any u

$$y = F(u) = h * u$$

where $*$ stands for discrete convolution product defined as follows

$$y_k = h_0 u_k + h_1 u_{k-1} + \dots + h_k u_0 = \sum_{j=0}^k h_j u_{k-j} = \sum_{j=0}^k u_j h_{k-j}$$

– Proof: Let $\delta^{(j)} = \delta_{(-j)} = \{0, \dots, 0, 1, 0, \dots\}$ be a j – delay, where 1 is in j – th position. Then by

property #3, $F(\delta^{(j)}) = h_{(-j)} = \{0, \dots, 0, h_0, h_1, \dots\}$ Now

$$u = \sum_{j=0}^{\infty} u_j \delta_{(-j)}$$

$$y = F(u) = \sum_{j=0}^{\infty} u_j F(\delta_{(-j)}) = \sum_{j=0}^{\infty} u_j h_{(-j)}$$

. Note that

$$(h_{(-j)})_k = 0 \text{ if } k < j, \quad (h_{(-j)})_k = h_{k-j} \text{ if } k \geq j$$

so for any k

$$y_k = \sum_{j=0}^{\infty} u_j (h_{(-j)})_k = \sum_{j=0}^{\infty} u_j h_{k-j} = \sum_{j=0}^k u_j h_{k-j} = (h * u)_k$$

– $H(z) = Z(h)$ is called transfer function of F with impulse response h

– Examples: (1) $u * \delta^{(j)} = u_{(-j)}$, (2) $Z(u_{(-j)}) = Z(u * \delta^{(j)}) = UZ(\delta^{(j)}) = Uz^{-j}$

– Example (page 38) Consider the filter with the finite impulse response (FIR)

$$h = \{1/2, 1/2, 0, 0, \dots\} = \frac{1}{2}\delta^{(0)} + \frac{1}{2}\delta^{(1)}$$

For any signal $u = \{u_0, u_1, \dots\}$,

$$y = u * h = \frac{1}{2}u * \delta^{(0)} + \frac{1}{2}u * \delta^{(1)} = \frac{1}{2}u + \frac{1}{2}h_{(-1)}.$$

This can also be seen directly

$$y_k = \sum_{j=0}^k u_j h_{k-j} = u_k h_0 + u_{k-1} h_1 = \frac{u_k + u_{k-1}}{2}, \quad y_0 = u_0 h_0 = \frac{u_0}{2}$$

So

$$y = \frac{u + u_{(-1)}}{2}, \quad Y = \frac{1}{2}U + \frac{1}{2z}U = \left(\frac{1 + z^{-1}}{2}\right)U$$

The transfer function is

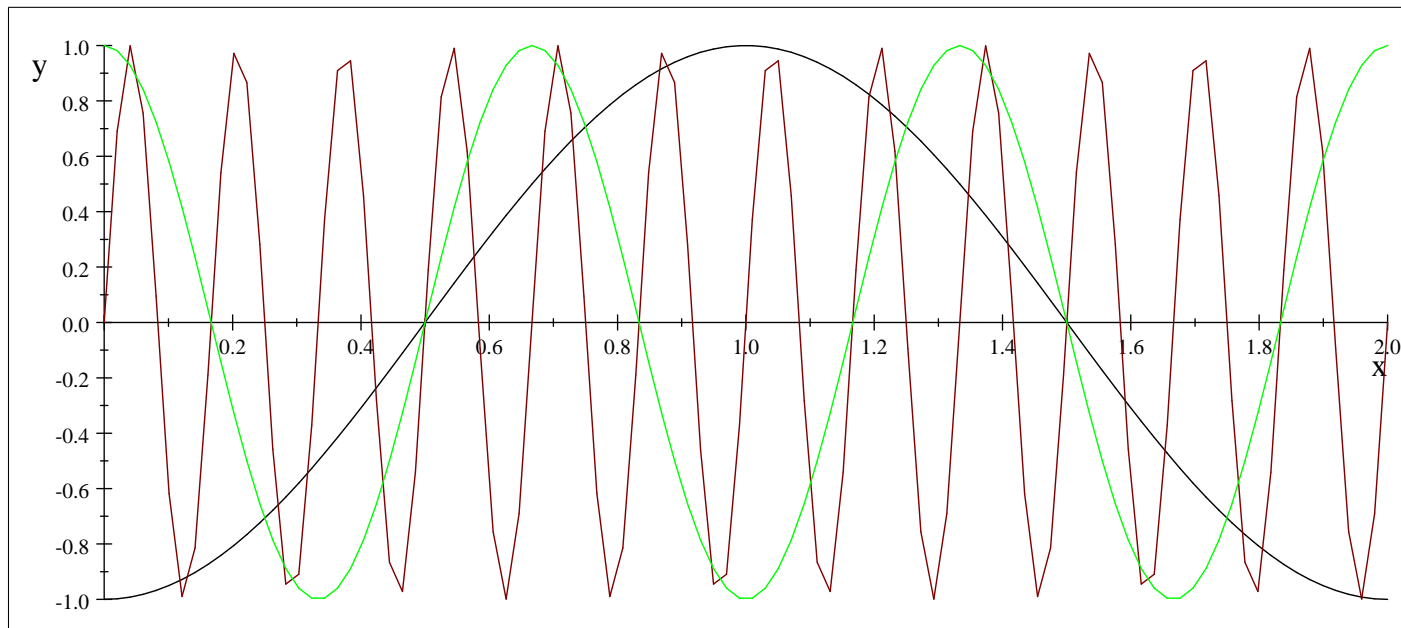
$$H = \frac{1 + z^{-1}}{2}$$

– Problem (page 38): Suppose that in a room with various sounds. One want to eliminate frequency 60 Hz sound wave. In other words, the component $A \sin(120\pi t)$ needs to be eliminated. In theory,

to do so, we need to add

$$-A \sin(120\pi t) = A \sin(120\pi t - \pi) = A \sin\left(120\pi\left(t - \frac{1}{120}\right)\right)$$

The black curve is sine wave of frequency 60 Hz with period $1/60$ the red has frequency 720 Hz with period $1/720$. If we sample using 720 Hz , i.e., pick up a signal every $T = 1/720$ second, it will meet the max of the black wave at $t = 1/120 = 6T$, $3/120 = 18T$, $5/120 = 30T$, $42T, \dots$, i.e., every $12T$.



$$scale = 1 : \frac{1}{120} = 6T$$

Let $u_k = signal$ picked up at time kT . So $u_6, u_{18}, u_{30}, \dots$ will be 60 Hz signal. To cancel it, we use the min point: at the second max point u_{18} , we use the previous min point u_{12} to cancel it. Thus the filter

$$y_k = u_k + u_{k-6}, \text{ or } Y = (1 + z^{-6}) U$$

- – Downside: Green wave has frequency 180 Hz. This filter could also cancel this frequency sound.

- Section 3.4 Stability

- We say a filter with impulse response h is stable if bounded inputs u yield bounded outputs y .

- Theorem: A filter is stable iff the transfer function $H(z)$ absolutely converges on the unit circle $|z| = 1$, i.e.,

$$\sum_{j=0}^{\infty} |h_j| < \infty$$

- If H has a pole at $z = z_0$ of order k , i.e.,

$$H(z) = \frac{G(z)}{(z - z_0)^k}, \quad G(z) \text{ is bounded near } z_0$$

then it is stable only if $|z_0| < 1$. This is because for all $|z| > 1$,

$$|z - z_0| \geq |z| - |z_0| > 1 - |z_0| > 0.$$

Thus

$$|H(z)| = \frac{|G(z)|}{|z - z_0|^k} \leq \frac{|G(z)|}{(1 - |z_0|)^k} \text{ is bounded for all } z.$$

– Example in page 40.

- Section 3.5 Polar and Bode Plots
- We shall introduce two graphing methods to exam efficacy of filters
- Theorem. Let F be a stable filter with real impulse response $h = \{h_0, h_1, \dots\}$ and transfer function $H(z) = Z(h)$. Then after transients have died away, the response to the sinusoidal signal $u_k = \sin \omega kT$ is also a sinusoid $y_k = r \sin(\omega kT + \phi)$ of the same frequency but different amplitude and phase angle

$$r = |H(e^{i\omega T})|, \quad \phi = \arg H(e^{i\omega T}).$$

- Proof: Recall that $y = F(u) = h * u$. In particular

$$y_k = \sum_{j=0}^k h_j u_{k-j}$$

For $u_k = \xi^k$,

$$\begin{aligned} y_k &= \sum_{j=0}^k h_j \xi^{k-j} = \xi^k \sum_{j=0}^k h_j \xi^{-j} \\ &= \xi^k \sum_{j=0}^{\infty} h_j \xi^{-j} - \xi^k \sum_{j=k+1}^{\infty} h_j \xi^{-j} = \xi^k H(\xi) + o(1), \end{aligned}$$

where $o(1) \rightarrow 0$ as $k \rightarrow \infty$ (by assumption of after dying away)

Set

$$H(\xi) = r e^{i\phi}, \quad \xi = e^{i\omega T}$$

then, the response to $u_k = \xi^k = e^{i\omega k T}$ is

$$\begin{aligned} y_k &= \xi^k H(\xi) + o(1) = r e^{i\phi} \xi^k + o(1) = r e^{i(\omega k T + \phi)} + o(1) \\ &= r [\cos(\omega k T + \phi) + i \sin(\omega k T + \phi)] + o(1) \end{aligned}$$

Since h_k are real, the response to the imaginary parts of $e^{i\omega k T}$, i.e., $\text{Im}(u_k) = \sin \omega k T$, should be the imaginary parts of the response to u_k

$$\text{Im}(y_k) = r \sin(\omega k T + \phi) + o(1).$$

- The graph $(r(\omega), \phi(\omega))$ in polar coordinate is called Polar plots. $T = \text{smampling period}$, $1/T$ sampling

rate.

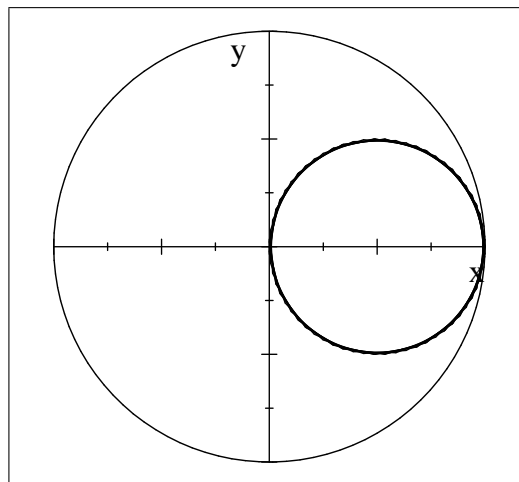
- Two separate graphs, $r = r(\omega)$, and $\phi = \phi(\omega)$ are called *Bode* plots (Bode-ee).
- Bode plot for r is often scaled in decibels (dB) of powers $20 \log_{10} r$ (see section 3.8)
- Example in page 41: $h = \{1/2, 1/2, 0, 0, 0, \dots\}$, $y_k = (u_k + u_{k-1})/2$, $H(z) = (1 + z^{-1})/2$. So

$$H(e^{i\omega T}) = \frac{1 + e^{-i\omega T}}{2} = \frac{1 + \cos \omega T - i \sin \omega T}{2}$$

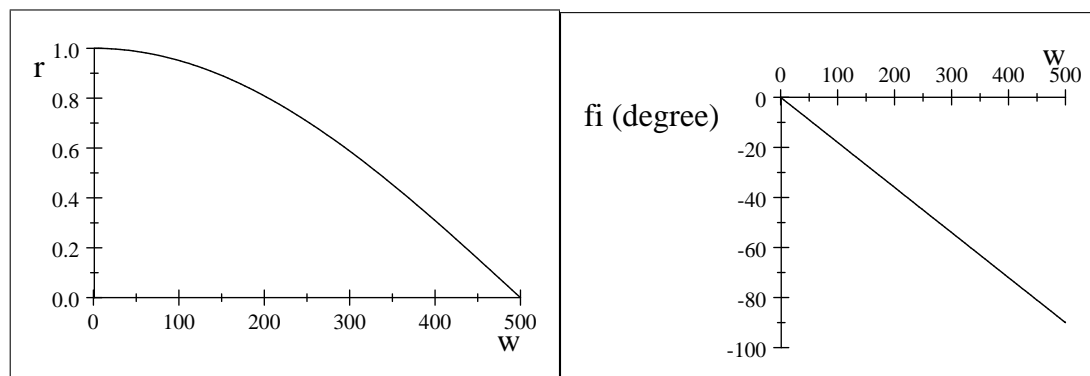
So

$$r = \sqrt{\frac{1 + \cos \omega T}{2}}, \quad \phi = -\arctan \frac{\sin \omega T}{1 + \cos \omega T}$$

– Polar plot



– Bode plots



● Section 3.7 Closing the loop

– See chart in page 47. We add a feedback filter H to improve stability

$$(U - HY)P = Y \implies Y = \frac{P}{1 + HP}U$$

– For any plant

$$P = \frac{1}{1 - az^{-1}} = \frac{z}{z - a}$$

that has a pole at $z = a$. So for $a > 1$, it is unstable. If we add a constant filter $H = k$. Then

$$Y = \frac{P}{1 + kP}U = \frac{z}{(1 + k)z - a}U$$

that has a pole at

$$z = \frac{a}{1 + k} < 1 \quad \text{if } k > a - 1$$

– Example (at bottom of page 48): For the plant $P = (1 - 2z^{-1})^{-1}$, $a = 2$. We add filter $k = 3/2 > 1$ will make it stable.

- Homework: 3.1, 3.2, 3.7, 3.8, 3.18, 3.20, 3.21
- Find explicit formula for Fibonacci sequence: $F_0 = F_1 = 1$, $F_{k+2} = F_k + F_{k+1}$ for $k = 0, 1, 2, \dots$. Then

write a Matlab routine to varify your answer. Would answer be different if we start with

$$F_0 = 1, F_1 = 2, F_{k+2} = F_k + F_{k+1} \text{ for } k = 0, 1, 2, \dots?$$

Or

$$F_0 = 2, F_1 = 3, F_{k+2} = F_k + F_{k+1} \text{ for } k = 0, 1, 2, \dots?$$

- Project: 3.10 (optional)