

6.3100 Lecture 2 Notes – Spring 2023

General solutions to first-order DT system, stability and convergence

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Outline:

1. Proportional control for first order discrete time system
2. Solutions to first order discrete time systems
3. Choosing K_p for a first order system: stability, steady-state error, and convergence

1. Proportional control for first order discrete time system

In the previous lecture, we introduced a simple first order discrete time (DT) system and the proportional controller. As a reminder, the controller and the first order system equations are given by:

$$\text{Proportional controller: } u[n] = K_p(T_d[n] - T_m[n])$$

$$\text{Plant: } \frac{T_m[n] - T_m[n-1]}{\Delta T} = \gamma u[n - 1]$$

We can substitute the first equation into the second equation:

$$\frac{T_m[n] - T_m[n - 1]}{\Delta T} = \gamma K_p(T_d[n - 1] - T_m[n - 1])$$

Simplifying this equation and collecting terms, we obtain the expression:

$$T_m[n] = (1 - \gamma K_p \Delta T) T_m[n - 1] + \gamma \Delta T K_p T_d[n - 1]$$

This equation has the form of a 1st-order DT system. We can write the general form as:

$$y[n] = \lambda y[n - 1] + b x[n - 1] \quad (\#1)$$

Here $y[n]$ is the variable we aim to solve, $x[n]$ is the input (driving) function we set, λ is the natural frequency (we will explain why later), and b is a multiplicative constant. In the next section, we will study the solution and property of equation 1 in detail.

2. Solutions to first order discrete time systems

We are going to solve equation (1) for several cases.

Case 1: $x[n]=0$ for all n . This is called zero-input response (ZIR)

The equation simplifies to $y[n] = \lambda y[n - 1]$.

The solution of this problem is given by:

$$y[n] = \lambda^n y[0]$$

This is a very simple case. Note that the steady state solution depends on the value of λ .

If $|\lambda| < 1$, then $y[\infty] = 0$.

If $\lambda = 1$, then $y[\infty] = y[0]$.

If $\lambda = -1$, then $y[n] = (-1)^n y[0]$. The solution does not converge.

If $|\lambda| > 1$, then $|y[\infty]| \rightarrow \infty$. The solution does not converge.

Case 2: $x[n] = 1$ for all n , and $y[0] = 0$. This is called zero-state response (ZSR).

Note: $x[n] = 1$ is not limiting. Through invoking linearity and time invariance (next lecture), we can relax the solution form by letting $x[n]$ be any arbitrary function.

In this case, equation (1) becomes

$$y[n] = \lambda y[n-1] + b$$

First, assuming the solution converges, let us find $y[\infty]$. We have

$$y[\infty] = \lambda y[\infty] + b$$

$$y[\infty] = \frac{b}{1 - \lambda}$$

Next, let's find $y[n]$. We can write $y[n]$ iteratively, as:

$$\begin{aligned} y[0] &= 0 \\ y[1] &= \lambda y[0] + b = b \\ y[2] &= \lambda y[1] + b = \lambda b + b \\ y[3] &= \lambda y[2] + b = \lambda^2 b + \lambda b + b \end{aligned}$$

Following this pattern, we get:

$$y[n] = \sum_{m=0}^{n-1} \lambda^m b \quad \text{and} \quad y[\infty] = \sum_{m=0}^{\infty} \lambda^m b$$

This implies

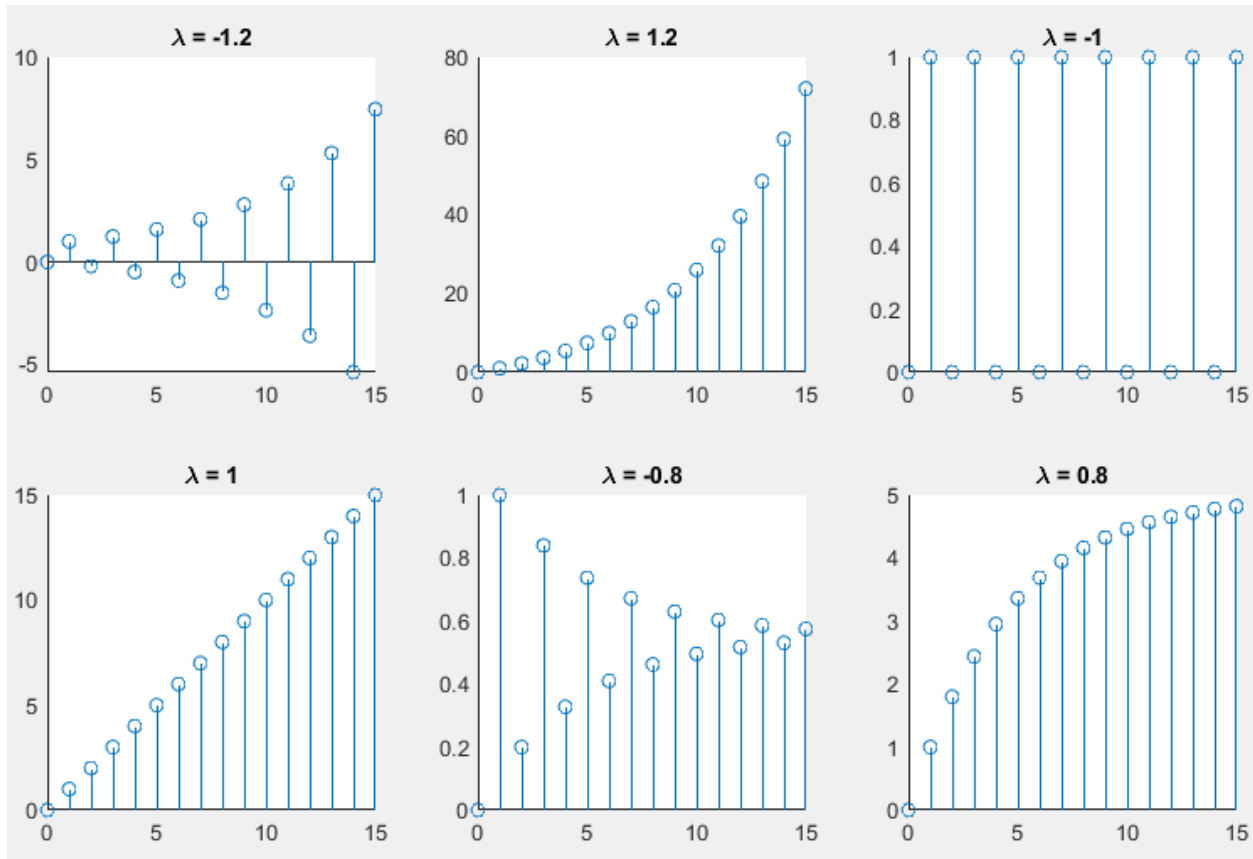
$$y[n] = y[\infty] - \sum_{m=n}^{\infty} \lambda^m b = y[\infty] - \lambda^n \sum_{m=0}^{\infty} \lambda^m b = y[\infty] - \lambda^n y[\infty] = y[\infty](1 - \lambda^n)$$

Substituting the solution of $y[\infty]$, we obtain:

$$y[n] = \frac{b}{1 - \lambda} (1 - \lambda^n)$$

Let's interpret what the solution looks like. Suppose $b = 1$, we consider 6 scenarios:

- (1) $\lambda > 1$. Solution diverges
- (2) $\lambda < -1$. Solution diverges
- (3) $\lambda = -1$. Solution diverges
- (4) $\lambda = 1$. Solution diverges
- (5) $0 < \lambda < 1$. Solution converges
- (6) $-1 < \lambda < 0$. Solution converges



Now we know how to solve 1st order DT systems, let's return to our 3D-printing controller example. It's worthwhile to emphasize again that the value of λ is crucial for the solution to either diverge or converge. In a control system, we need to design stable systems through setting the value of λ .

3. Choosing K_p for a first order system: stability, steady-state error, and convergence

Returning to the 3D-printing example, the system equation is given by:

$$T_m[n] = (1 - \gamma K_p \Delta T) T_m[n - 1] + \gamma \Delta T K_p T_d[n - 1]$$

The key question is how should we choose K_p to construct a "good" controller?

First, we can pattern-match to find λ and b . We have:

$$\begin{aligned}\lambda &= 1 - \gamma K_p \Delta T \\ b &= \gamma \Delta T K_p T_d[n]\end{aligned}$$

Here we can assume the desired temperature is constant.

There are several key metrics we need to consider:

(1) Stability:

$$\begin{aligned}-1 &< \lambda < 1 \\ -1 &< 1 - \gamma K_p \Delta T < 1 \\ \frac{2}{\gamma \Delta T} &> K_p > 0\end{aligned}$$

For this control problem, K_p must be chosen in the desired range to guarantee system stability (that is $T_m[\infty]$ is a finite number).

(2) Steady-state error:

We can use the steady-state solution to evaluate if there is any steady state error. We have:

$$T_m[\infty] = y[\infty] = \frac{b}{1 - \lambda} = \frac{\gamma \Delta T K_p T_d[\infty]}{1 - (1 - \gamma K_p \Delta T)} = T_d[\infty]$$

In this particular problem, $T_m[\infty] = T_d[\infty]$. As long as the system is stable, then there is no steady-state error. This is only true for this particular example. In the next class, we are going to see an example where K_p influences the steady state error.

(3) Convergence rate:

Thus far, the two conditions only give us a range of valid K_p . What is the optimal K_p ? There are many metrics to optimize for. In this example, let's consider the goal of making the measured temperature $T_m[n]$ approach its desired value $T_d[n]$ as soon as possible. Going back to the general solution:

$$y[n] = \frac{b}{1 - \lambda} (1 - \lambda^n)$$

What if we let $\lambda = 0$? Then we have:

$$y[1] = \frac{b}{1} (1) = b$$

This is a very nice result because the temperature approaches the desired value in 1 step. This is very fast convergence. Realistically, it may be influenced by external noises, and it usually requires a large control input. Those tradeoffs are things we need to consider when designing a realistic controller.