

$$1) \quad 1. G_1(s) = \frac{5s+5}{s^4+3s^3+3s^2+3s+2}$$

s^4	1	3	2
s^3	3	3	0
s^2	2	2	0
s^1	4	0	0
s^0	2	0	0

$$\rightarrow a(s) = 2s^2 + 2$$

$$\frac{da(s)}{ds} = 4s$$

marginally stable

$$2. G_2(s) = \frac{8s+1}{s^4+3s^3+3s^2+3s+1}$$

s^4	1	3	1
s^3	3	9	0
s^2	6	1	0
s^1	6	0	0
s^0	1	0	0

$$c = \frac{96-3}{6}$$

$$\lim_{c \rightarrow \infty} c = -\infty$$

unstable (two unstable poles)

$$3. G_3(s) = \frac{4}{s^4+4s^3+3s^2-4s-4}$$

s^4	1	3	-4
s^3	4	-4	0
s^2	4	-4	0
s^1	8	0	0
s^0	-4	0	0

$$\rightarrow a(s) = 4s^2 - 4$$

$$\frac{da(s)}{ds} = 8s$$

unstable (one unstable pole)

$$4. G_4(s) = \frac{2s-1}{s^5+2s^4+2s^3+4s^2+11s+10}$$

s^5	1	2	11
s^4	2	4	10
s^3	6	6	0
s^2	6	10	0
s^1	6	0	0
s^0	10	0	0

$$c = \frac{46-12}{6}$$

$$\lim_{c \rightarrow \infty} c = -\infty$$

$$d = \frac{6c-106}{6}$$

$$\lim_{c \rightarrow -\infty} d = 6$$

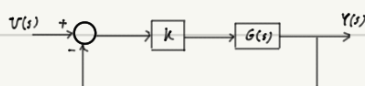
unstable (two unstable poles)

$$5. G_5(s) = \frac{2}{s^6+7s^5+20s^4+30s^3+25s^2+11s+2}$$

s^6	1	20	25	2
s^5	7	30	11	0
s^4	119	169	2	0
s^3	19.56	10.1	0	0
s^2	15.3	2	0	0
s^1	7.55	0	0	0
s^0	2	0	0	0

stable

$$2) \quad G(s) = \frac{2s^3+s^2+s-1}{s^3-5s^2+2s+8}$$



$$\frac{Y(s)}{V(s)} = \frac{KG(s)}{1+KG(s)} \Rightarrow 1+KG(s) = 0 \Rightarrow \frac{(2k+1)s^3+(k-5)s^2+(k+2)s-k+8}{s^3-5s^2+2s+8} = 0$$

s^3	$2k+1$	$k+2$
s^2	$k-5$	$-k+8$
s^1	c	0
s^0	$-k+8$	0

$$2k+1 > 0 \rightarrow k > -\frac{1}{2}$$

$$k-5 > 0 \rightarrow k > 5$$

$$c > 0 \rightarrow k > 3+\sqrt{15} \quad \& \quad 3-\sqrt{15} < k < 5$$

$$-k+8 > 0 \rightarrow k < 8$$

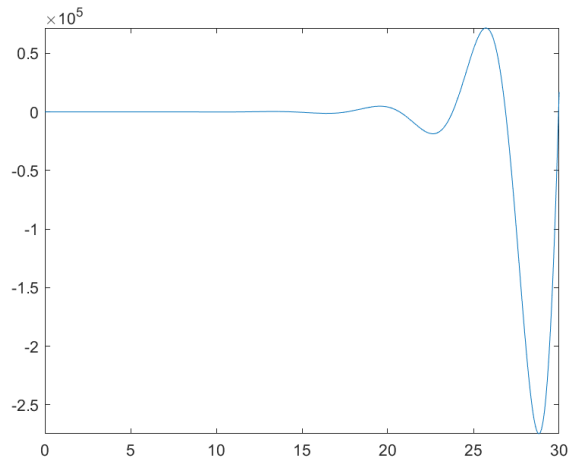
$$c = \frac{(2k+1)(-k+8)-(k+2)(k-5)}{-(k-5)} = \frac{3(k^2-6k-6)}{k-5}$$

$$\left. \begin{array}{l} 2k+1 > 0 \rightarrow k > -\frac{1}{2} \\ k-5 > 0 \rightarrow k > 5 \\ c > 0 \rightarrow k > 3+\sqrt{15} \quad \& \quad 3-\sqrt{15} < k < 5 \\ -k+8 > 0 \rightarrow k < 8 \end{array} \right\} \rightarrow 3+\sqrt{15} < k < 8$$

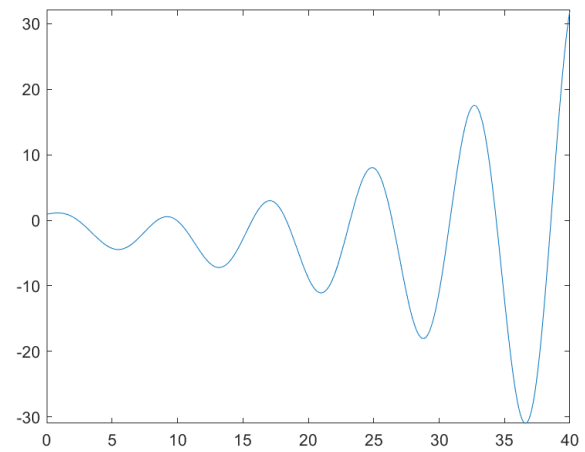
$$\downarrow \\ \approx 6.87$$

2 Range of a Gain for Closed-Loop Stability

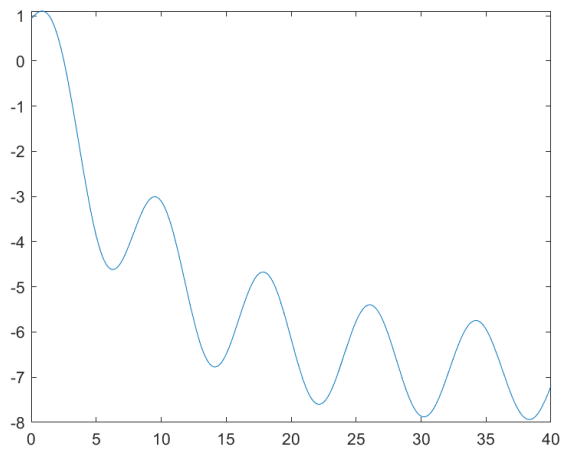
For different values of k we have these step responses:



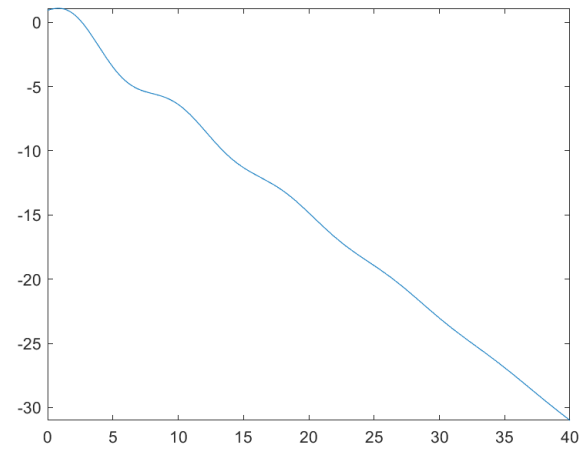
$k = 3$



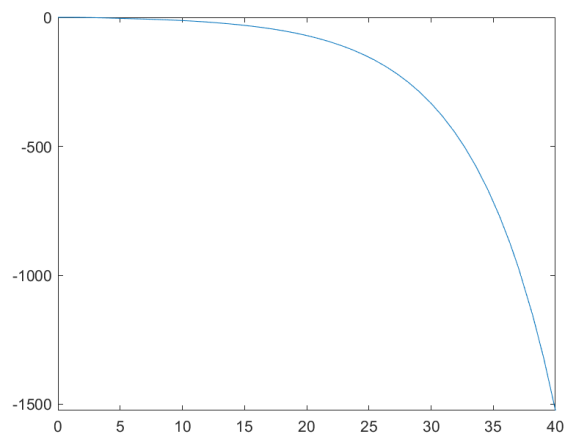
$k = 6$



$k = 7$



$k = 8$



$k = 10$

3- a) $y(t) = \int_{-\infty}^t u(t-\tau) h(\tau) d\tau$

$\xrightarrow{t=\tau_2} \Rightarrow \|y\|_{\infty} = y(\tau_2) = \int_0^{\tau_2} u(\tau_2-\tau) h(\tau) d\tau \leq \int_0^{\tau_2} |u(\tau_2-\tau)| |h(\tau)| d\tau \leq \int_0^{\infty} |u(\tau_2-\tau)| |h(\tau)| d\tau \leq \int_0^{\infty} \|u\|_{\infty} |h(\tau)| d\tau \leq \|u\|_{\infty} \int_0^{\infty} |h(\tau)| d\tau$

\downarrow
[u] هو مقدار ثابت

b) $u = a \operatorname{sgn}(h(t)) u(t) \quad (a > 0)$

$\Rightarrow \|y\|_{\infty} = \|u\|_{\infty} \int_0^{\infty} |h(\tau)| d\tau \quad \Rightarrow \|z\| = \int_0^{\infty} |h(\tau)| d\tau$

c) 1) $s_1: G_1(s) = \frac{1}{s+1} \xrightarrow{\mathcal{L}^{-1}} e^{-t} \rightarrow \|z\| = \int_0^{\infty} |e^{-\tau}| d\tau = \int_0^{\infty} e^{-\tau} d\tau = -e^{-\tau} \Big|_0^{\infty} = 1 \rightarrow \text{stable}$

2) $s_2: G_2(s) = \frac{1}{s-1} \xrightarrow{\mathcal{L}^{-1}} e^t \rightarrow \|z\| = \int_0^{\infty} |e^{\tau}| d\tau = \int_0^{\infty} e^{\tau} d\tau = e^{\tau} \Big|_0^{\infty} = \infty \rightarrow \text{unstable}$

3) $s_3: G_3(s) = \frac{1}{s^2+1} \xrightarrow{\mathcal{L}^{-1}} \sin(t) \rightarrow \|z\| = \int_0^{\infty} |\sin(\tau)| d\tau = \frac{\sin(\tau)}{|\sin(\tau)|} \int_0^{\infty} \sin(\tau) d\tau = \frac{-\cos(\tau) \sin(\tau)}{|\sin(\tau)|} \Big|_0^{\infty} = ? \rightarrow \text{unstable}$

4) $s_4: G_4(s) = \frac{1}{(s+1)^2} \xrightarrow{\mathcal{L}^{-1}} t e^{-t} \rightarrow \|z\| = \int_0^{\infty} |\tau e^{-\tau}| d\tau = \int_0^{\infty} e^{-\tau} |\tau| d\tau = -e^{-\tau} |\tau| \Big|_0^{\infty} + \int_0^{\infty} \frac{\tau e^{-\tau}}{|\tau|} d\tau$

$= -e^{-\tau} |\tau| \Big|_0^{\infty} - \frac{\tau e^{-\tau}}{|\tau|} \Big|_0^{\infty} = \frac{-(\tau^2 + \tau) e^{-\tau}}{|\tau|} \Big|_0^{\infty} = 1 \rightarrow \text{stable}$

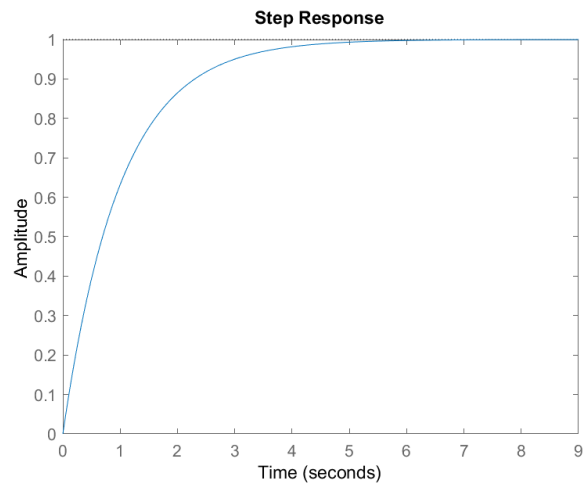
5) $s_5: h_5(t) = \frac{1}{t+1} \rightarrow \|z\| = \int_0^{\infty} \left| \frac{1}{\tau+1} \right| d\tau = \frac{\tau+1}{|\tau+1|} \int_0^{\infty} \frac{1}{\tau+1} d\tau = \frac{\tau+1}{|\tau+1|} \ln(\tau+1) \Big|_0^{\infty} = \infty \rightarrow \text{unstable}$

6) $s_6: h_6(t) = \frac{1}{t^2+1} \rightarrow \|z\| = \int_0^{\infty} \left| \frac{1}{\tau^2+1} \right| d\tau = \int_0^{\infty} \frac{1}{\tau^2+1} d\tau = \arctan(\tau) \Big|_0^{\infty} = \frac{\pi}{2} \rightarrow \text{stable}$

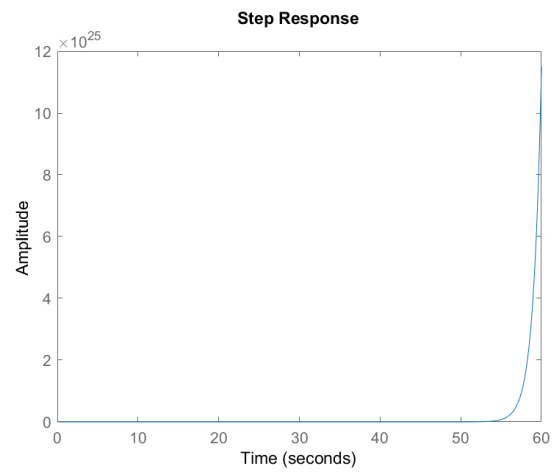
3 Evaluating Stability Condition Using Impulse Response

c)

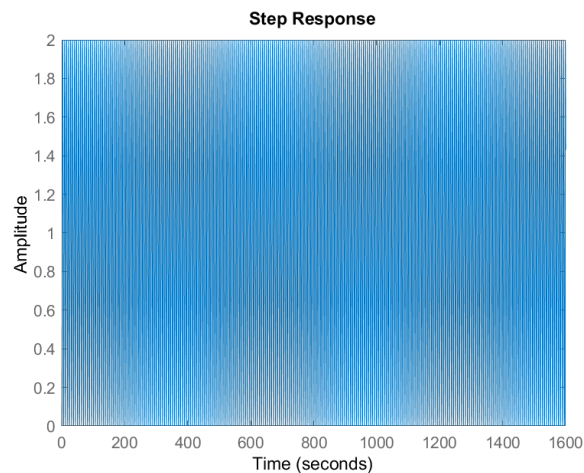
1)



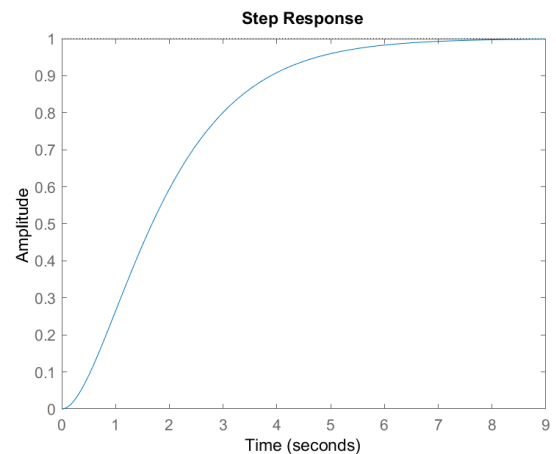
2)



3)



4)



5)

Input

$$\int_0^{\infty} \frac{1}{1 + (t - T)} dT$$

Result

(integral does not converge)

$$(u(t) * h_5(t))$$

6)

Definite integral

$$\int_0^{\infty} \frac{1}{1 + (t - T)^2} dT = \tan^{-1}(t) + \frac{\pi}{2}$$

$$(u(t) * h_6(t))$$

$$4) a) \left. \begin{aligned} \frac{1}{\sqrt{1-z^2}} e^{-z\omega_n t} < 0.02 &\Rightarrow T_s = \frac{\ln(0.02\sqrt{1-z^2})}{-z\omega_n} \\ 0 < z < 1 \rightarrow z^2 < 1 \end{aligned} \right\} \rightarrow T_s = \frac{4}{z\omega_n}$$

$$z=1 \Rightarrow Y(s) = \frac{\omega_n^2}{s(s^2 + 2\omega_n s + \omega_n^2)} = \frac{\omega_n^2}{s(s+\omega_n)^2} = \frac{1}{s} - \frac{1}{s+\omega_n} - \frac{\omega_n}{(s+\omega_n)^2} \rightarrow y(t) = 1 - e^{-\omega_n t} - \omega_n t e^{-\omega_n t}$$

$$\Rightarrow e^{-\omega_n t} + \omega_n t e^{-\omega_n t} < 0.02 \Rightarrow \ln(1 + \omega_n T_s) - \omega_n T_s = 4 \Rightarrow \omega_n T_s = 6 \Rightarrow T_s = \frac{6}{\omega_n}$$

$$b) M_p = \max(y(t)) - 1 = e^{\frac{-z\pi}{\sqrt{1-z^2}}}$$

$$Y(s) = \frac{\omega_n^2}{s(s^2 + 2z\omega_n s + \omega_n^2)} \Rightarrow sY(s) = \frac{\omega_n^2}{s^2 + 2z\omega_n s + \omega_n^2}$$

$$\Rightarrow y(t) = \mathcal{L}^{-1}\{sY(s)\} = \frac{\omega_n}{\sqrt{1-z^2}} e^{-z\omega_n t} \sin(\omega_n \sqrt{1-z^2} t) = 0 \Rightarrow t = \frac{n\pi}{\omega_n \sqrt{1-z^2}} \Rightarrow T_p = \frac{\pi}{\omega_n \sqrt{1-z^2}}$$

$$y(T_p) = 1 - \frac{1}{\sqrt{1-z^2}} e^{\frac{-z\pi}{\sqrt{1-z^2}}} \sin(\pi + z\omega_n T_p) \left. \begin{aligned} &\rightarrow \max(y(t)) = 1 + e^{\frac{-z\pi}{\sqrt{1-z^2}}} \Rightarrow M_p = e^{\frac{-z\pi}{\sqrt{1-z^2}}} \\ z \rightarrow 0 &\Rightarrow \begin{cases} z\omega_n T_p \rightarrow \frac{\pi}{2} \\ \sin(\pi + z\omega_n T_p) \rightarrow -1 \\ \frac{1}{\sqrt{1-z^2}} \rightarrow 1 \end{cases} \end{aligned} \right\}$$

for not having overshoot the poles must be real-valued

for fastest response we should make dominant pole as fast as possible

so we want two real-valued poles at the same location $\Rightarrow z=1 \rightarrow s_{1,2} = -\omega_n$

$$z=1 \rightarrow G(s) = \frac{\omega_n^2}{(s+\omega_n)^2}$$

MATLAB Assignments

5 Curve Fitting Toolbox

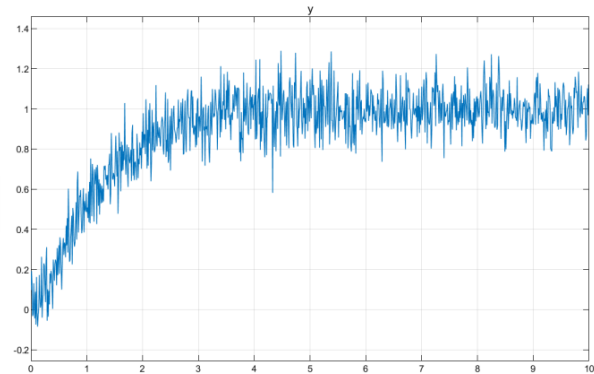
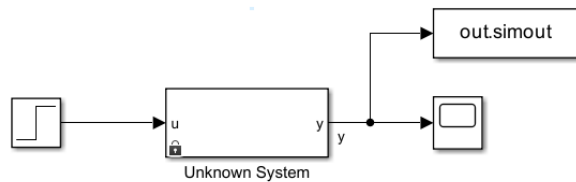
a)

$$G(s) = \frac{K}{\tau s + 1}, \quad U(s) = \frac{1}{s}$$

```
%% 5_a
syms K T s
step_res1 = ilaplace(K/(T*s+1) * (1/s));
```

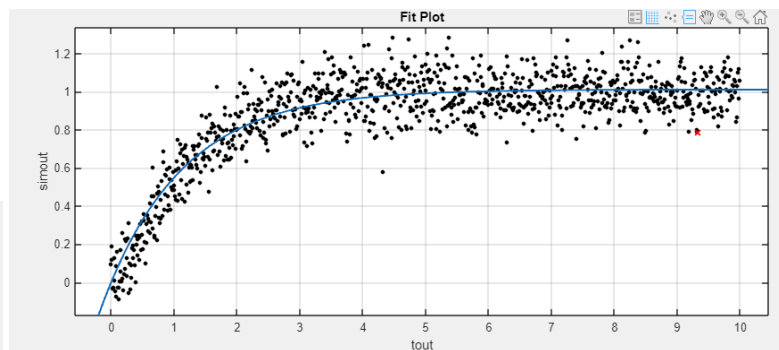
```
val =
K - K*exp(-t/T)
```

b)



c)

General model:
 $f(x) = a*(1-\exp(-b*x))$
 Coefficients (with 95% confidence bounds):
 a = 1.012 (1.003, 1.021)
 b = 0.7834 (0.7488, 0.818)



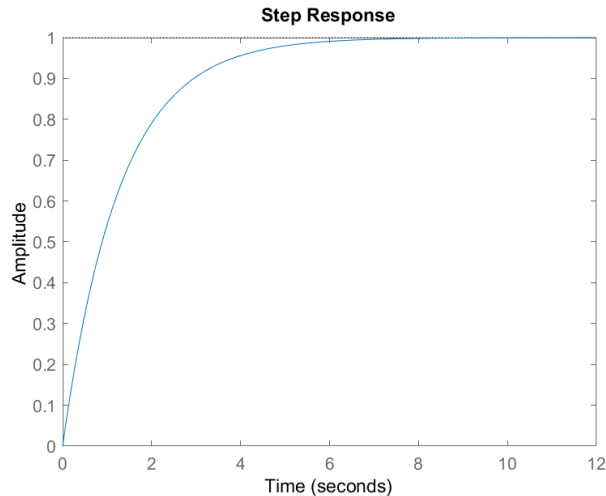
From data we have:

$$f(x) = a (1 - e^{-bx}) = 1.012 (1 - e^{-0.7834x}) \approx 1 (1 - e^{-0.78x})$$

compare to result in part a:

$$K = 1, \quad \tau = \frac{1}{0.78} = 1.28$$

d)

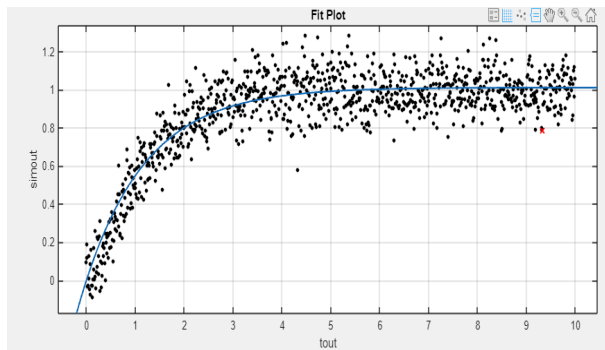


```
%% 5_d

K2 = 1;
T2 = 1.28;

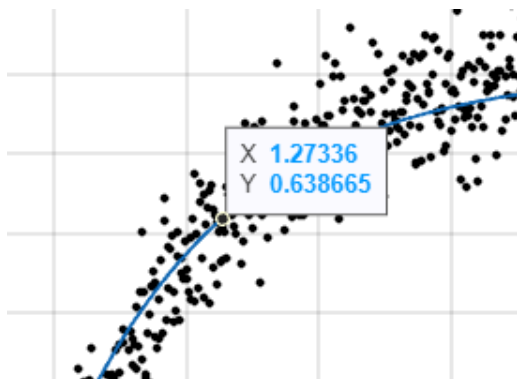
step( tf([K2],[T2,1]) )
```

As we see the result of step response for our system is similar to the output from the unknown system.



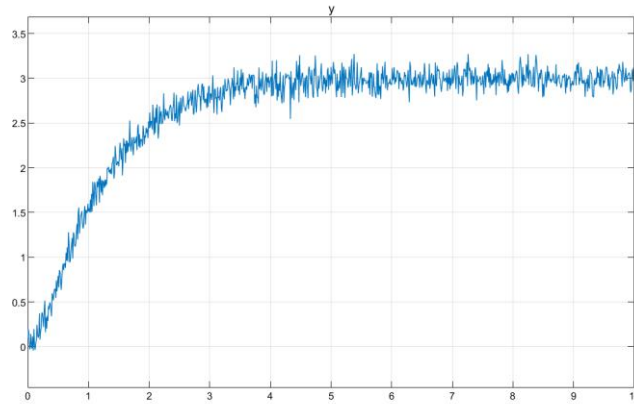
e) For the DC gain we should consider the value of response as the time goes to infinity. We can say that DC gain could be a little more than one (because of a). so 1 is a proper estimation for K (DC gain)

We know: $1 - e^{-1} = 0.632$, so the time constant could be estimated to be the time when the curve reaches to 63.2% of its maximum value.

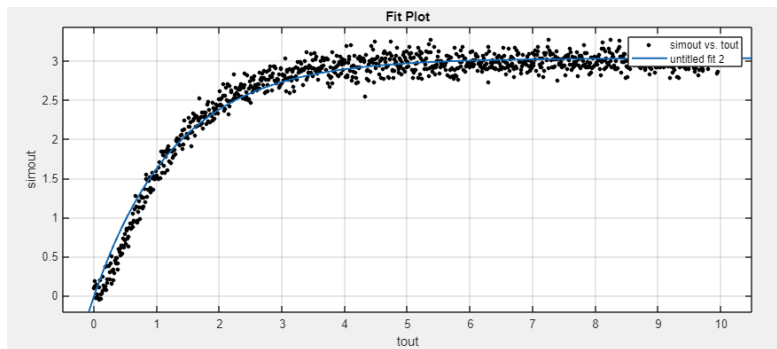


As we see the value of time constant that we've estimated (1.28) could be a proper value for it.

f) We set the input amplitude on 3.



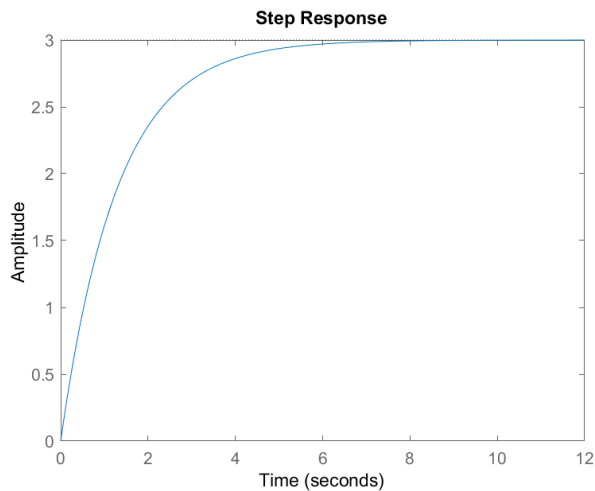
General model:
 $f(x) = a \cdot (1 - \exp(-b \cdot x))$
 Coefficients (with 95% confidence bounds):
 $a = 3.034 \ (3.023, 3.044)$
 $b = 0.769 \ (0.7552, 0.7827)$



$$f(x) = a (1 - e^{bx}) = 3.034 (1 - e^{-0.769x}) \approx 3 (1 - e^{-0.77x})$$

compare to results in part a:

$$K = 3, \quad \tau = \frac{1}{0.77} = 1.3$$



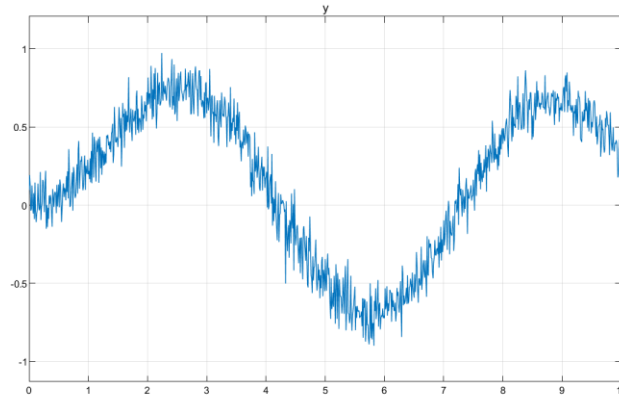
```
%% 5_f
```

```
K3 = 3;  
T3 = 1.3;
```

```
step( tf([K3],[T3,1]) )
```

As we see the result of step response for our system is similar to the output from the unknown system.

g) input : $\sin(t)$

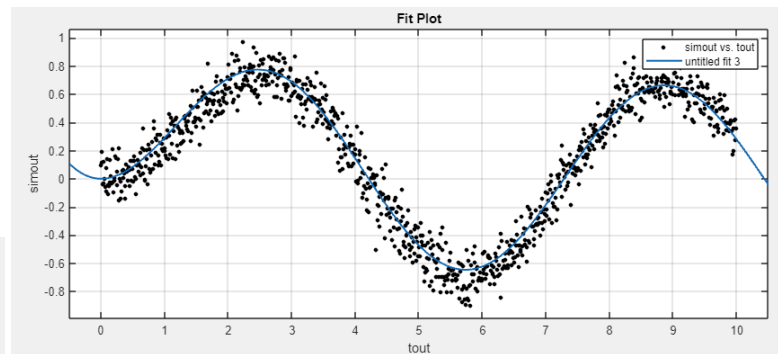


val =

$$(K \sin(t) - K T \cos(t)) / (T^2 + 1) + (K T \exp(-t/T)) / (T^2 + 1)$$

$$L\{\sin(t)\} = \frac{1}{s^2 + 1} \Rightarrow \text{response:}$$

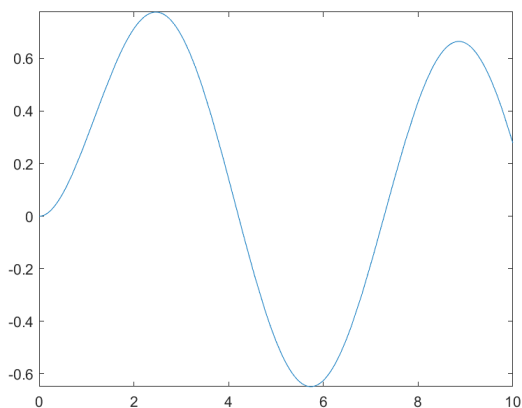
General model:
 $f(x) = (K \sin(x) - K T \cos(x) + K T \exp(-x/T)) / (T^2 + 1)$
 Coefficients (with 95% confidence bounds):
 K = 1.238 (1.21, 1.266)
 T = 1.582 (1.535, 1.63)



$$f(x) = \frac{1}{T^2 + 1} (K \sin(t) - K T \cos(t) + K T e^{-\frac{t}{T}})$$

$$= \frac{1}{(1.582)^2 + 1} (1.238 \sin(t) - 1.238 \times 1.582 \cos(t) + 1.238 \times 1.582 e^{-\frac{t}{1.582}})$$

$$\approx 0.35 \sin(t) - 0.56 \cos(x) + 0.56 e^{-0.63x} \quad (K = 1.24, T = 1.58)$$



%% 5_g_2

K4 = 1.24;

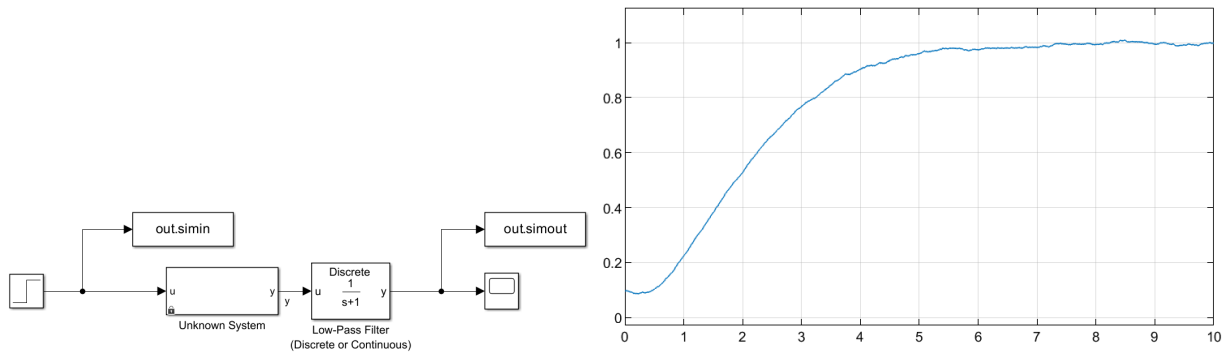
T4 = 1.58;

```
fplot( ilaplace( (K4/(T4*s+1))*(1/(s^2+1))
) )
xlim([0 10])
```

As we see the result of step response for our system is similar to the output from the unknown system.

h) Increasing data length means to have more data to calculate the proper equation for the output curve. This also helps us to understand the steady state behavior of the response better. So we can say that increasing data length will increase the accuracy of estimating output equation and coefficients.

i)



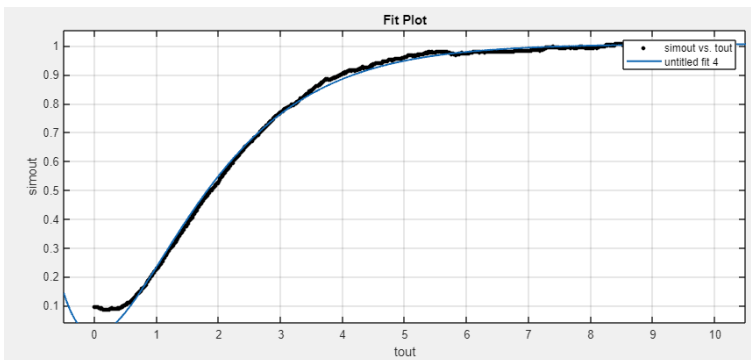
Time constant: $a = 1 \Rightarrow f_c = \frac{1}{2\pi a} = 0.16 \text{ Hz}$

$$G(s)F(s) = \frac{k}{\tau s + 1} \frac{1}{as + 1}$$

```
val =
K - (K*T*exp(-t/T))/(T - a) + (K*a*exp(-t/a))/(T - a)
```

\Rightarrow step response:

General model:
 $f(x) = K + (-K*T*exp(-x/T) + K*exp(-x))/(T-1)$
 Coefficients (with 95% confidence bounds):
 K = 1.006 (1.004, 1.008)
 T = 1.186 (1.174, 1.198)



$K \approx 1$, $\tau \approx 1.18 \Rightarrow G(s) = \frac{1}{1.18s + 1}$

j)

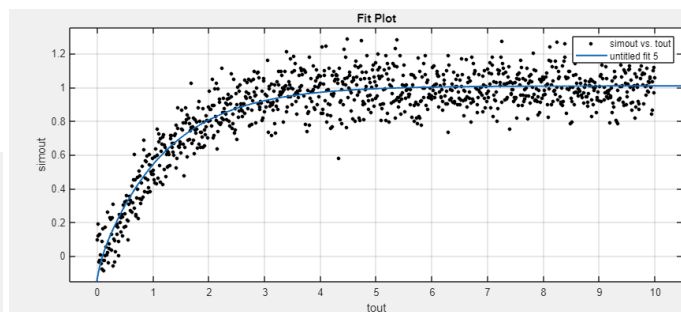
Answers from different inputs have near values for K and τ . Although the values we got from sinusoidal input were a little different from the others.

k)

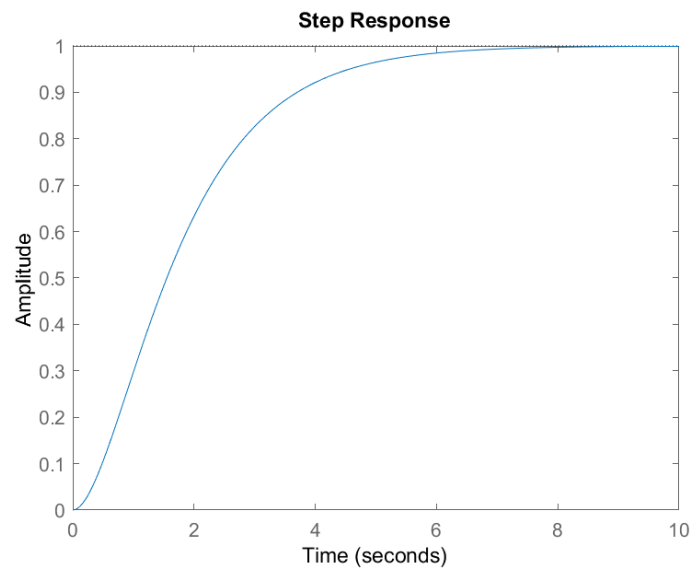
val =

$$K - (K \cdot T_1 \cdot \exp(-t/T_1)) / (T_1 - T_2) + (K \cdot T_2 \cdot \exp(-t/T_2)) / (T_1 - T_2)$$

General model:
 $f(x) = K \cdot (K \cdot T_1 \cdot \exp(-x/T_1) + K \cdot T_2 \cdot \exp(-x/T_2)) / (T_1 - T_2)$
 Coefficients (with 95% confidence bounds):
 $K = 1.009 (0.9999, 1.017)$
 $T_1 = 1.197 (1.135, 1.258)$
 $T_2 = 0.06852 (0.04278, 0.09426)$

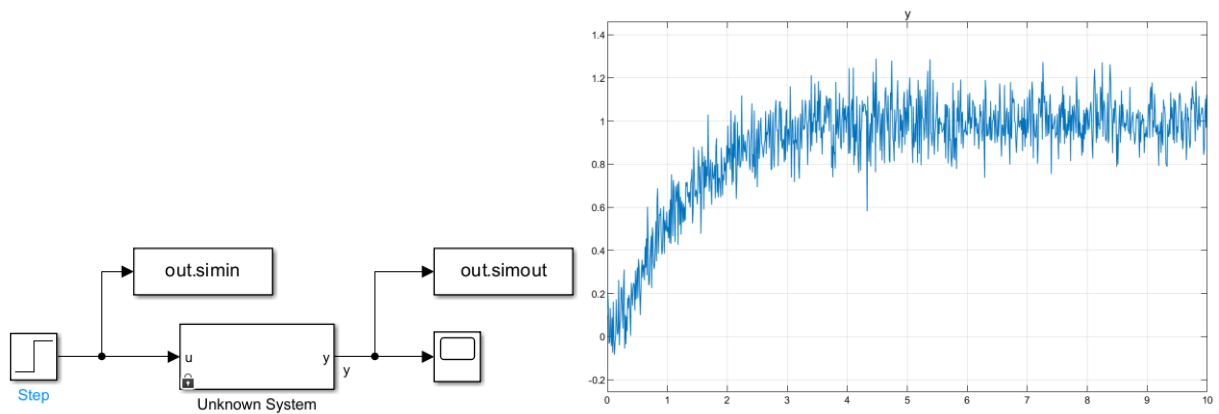


$$K \approx 1, \quad \tau_1 \approx 1.2, \quad \tau_2 \approx 0.68$$



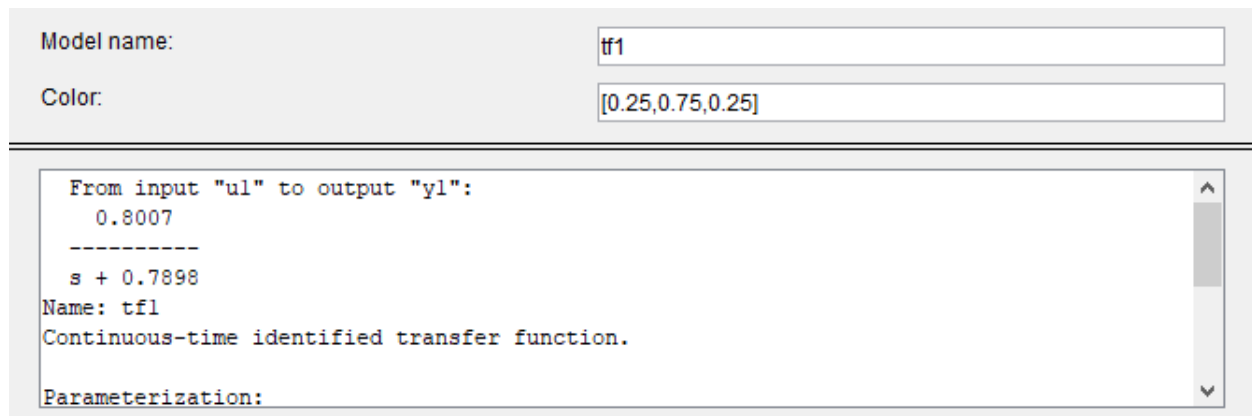
6 System Identification Toolbox

a)



b)

Result from System Identification toolbox :



$u(t)$

$$K = 0.8007 \quad a = 0.7898$$

c)

Model name:	tf2
Color:	[0.25,0.75,0.25]

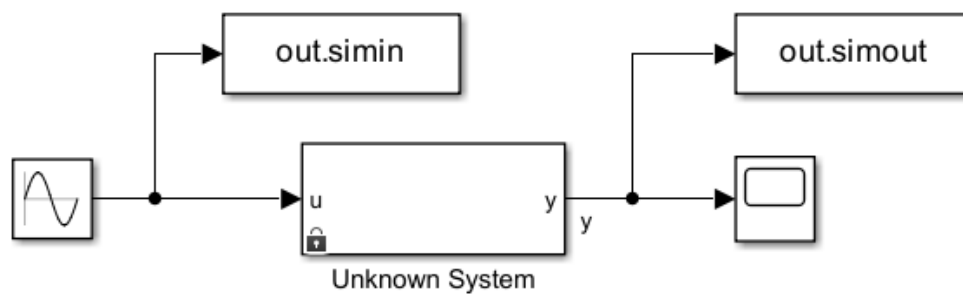
From input "u1" to output "y1":
0.7865

 $s + 0.7765$
Name: tf2
Continuous-time identified transfer function.
Parameterization:

$3u(t)$

$$K = 0.7865 \quad a = 0.7765$$

d)



Model name:	tf3
Color:	[0.25,0.75,0.25]

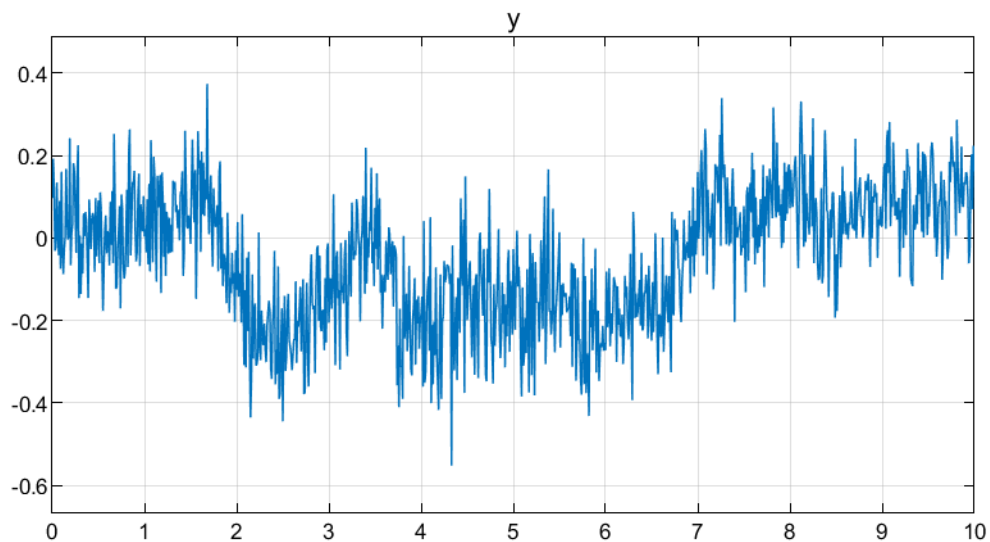
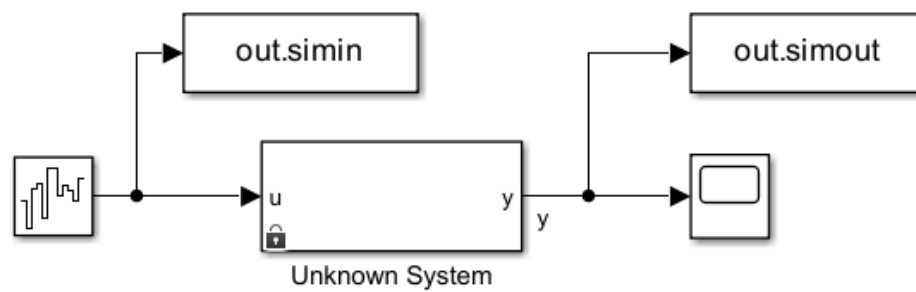
From input "u1" to output "y1":
0.7859

 $s + 0.6404$
Name: tf3
Continuous-time identified transfer function.
Parameterization:

$\sin(t)$

$$K = 0.7859 \quad a = 0.6404$$

e)



output

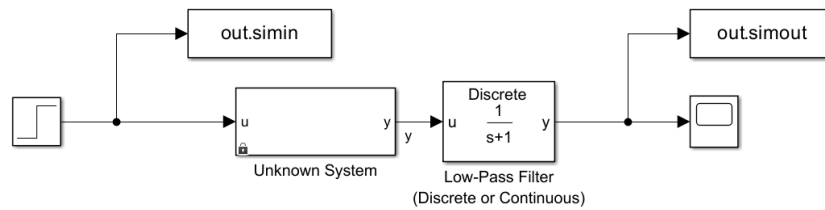
Model name:	<input type="text" value="tf4"/>
Color:	<input type="text" value="[0.75,0.75,0.2]"/>

```
From input "u1" to output "y1":
  0.7051
-----
  s + 0.8805
Name: tf4
Continuous-time identified transfer function.
Parameterization:
```

White noise

$$K = 0.7051 \quad a = 0.8805$$

f)



Model name:	tf5
Color:	[0.75,0.75,0.75]

From input "u1" to output "y1":

0.4161

s + 0.3852

Name: tf5

Continuous-time identified transfer function.

Parameterization:

u(t) with low-pass filter

$$K = 0.4161 \quad a = 0.3852$$

g)

Answers from different inputs have near values for K and a. Although the values we got from white noise and sinusoidal input and filtered system were a little different from the others.

h) Result from System Identification toolbox :

Model name:	tf6
Color:	[0,0,1]

From input "u1" to output "y1":

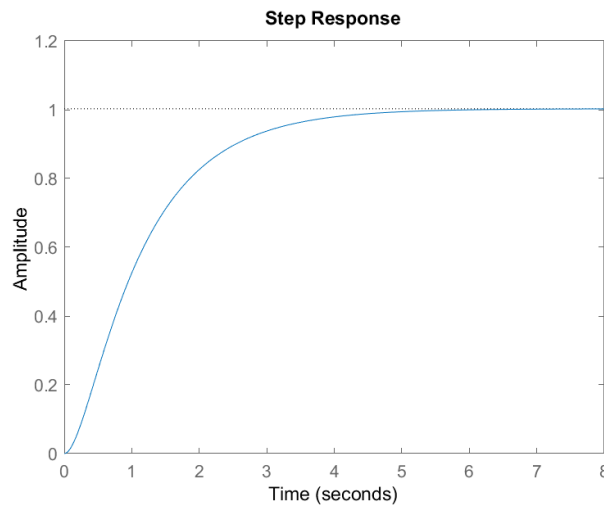
4.25

s^2 + 5.228 s + 4.238

Name: tf6

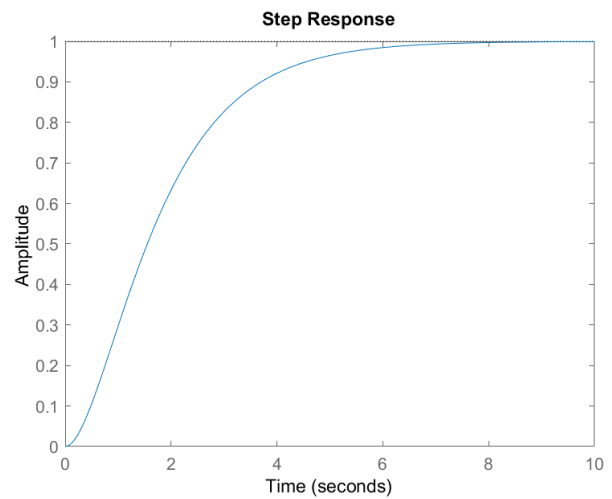
Continuous-time identified transfer function.

Parameterization:



Step response of result (part h)

$$\frac{4.25}{s^2 + 5.228 s + 4.238} = \frac{4.25}{(s+1.0031)(s+4.2249)}$$

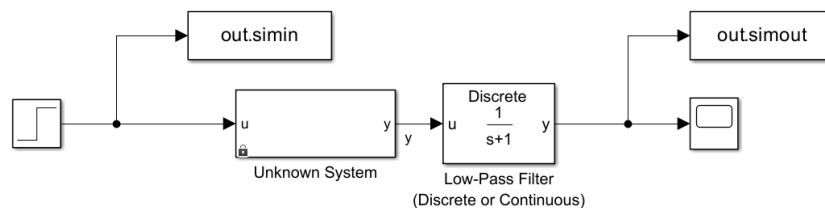


Step response of Curve Fitting

$$\frac{1}{(s+1.2)(s+0.68)} = \frac{1}{s^2 + 1.88 s + 0.816}$$

Result from Curve Fitting toolbox is similar to our system and had better performance compare to System Identification toolbox.

*) suggested method:



Step response:

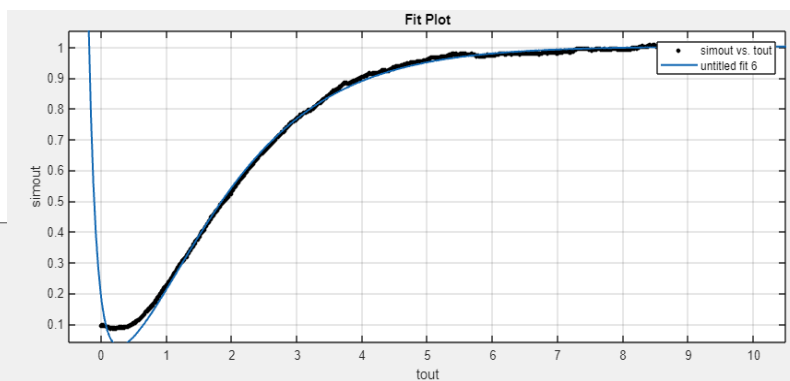
```
val =
K - (K*a^2*exp(-t/a))/((T1 - a)*(T2 - a)) - (K*T1^2*exp(-t/T1))/((T1 - T2)*(T1 - a)) + (K*T2^2*exp(-t/T2))/((T1 - T2)*(T2 - a))

=
K - (K*exp(-t))/((T1 - 1)*(T2 - 1)) - (K*T1^2*exp(-t/T1))/((T1 - T2)*(T1 - 1)) + (K*T2^2*exp(-t/T2))/((T1 - T2)*(T2 - 1))
```


General model:

$$f(x) = K \cdot K' (\exp(-x) (T1 - T2) + (T1^2)^2 \exp(-x/T1) (T2 - 1) - (T2^2)^2 \exp(-x/T2) (T2 - 1)) / ((T1 - T2) (T1 - 1) (T2 - 1))$$

Coefficients (with 95% confidence bounds):
K = 1.003 (1.002, 1.004)
T1 = 1.074 (1.068, 1.079)
T2 = 0.109 (0.1058, 0.1122)



$$K \approx 1 \quad , \quad \tau_1 \approx 1.07 \quad , \quad \tau_2 \approx 0.11$$