
```
% MECH 6325 - Final Project
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%2020-11-28
```

```
clear
close all;
```

```
% Problem 1
```

```
-----
disp('Problem 1
-----')
pblm1()
```

```
clear
```

```
% Problem 2
```

```
-----
disp('Problem 2
-----')
pblm2()
```

```
clear
```

```
% Problem 3
```

```
-----
disp('Problem 3
-----')
pblm3()
```

```
clear
```

```
% Problem 4
```

```
-----
disp('Problem 4
-----')
pblm4()
```

```
clear
```

```
% Problem 5
```

```
-----
disp('Problem 5
-----')
pblm5()
```

```
clear
```

```
close all %Comment this out if figures are wanted...
```

```
Problem 1 -----
If using the following  $F$ , there is a discrepancy for state 2
```

```
 $F =$ 
```

```
    0.5000    2.0000
         0         0
```

$P_{\text{minus_inf}} =$

52.0974 -0.0000
-0.0000 10.0000

$k_{\text{inf}} =$

0.8390
-0.0000

$P_{\text{plus_inf}} =$

8.3896 -0.0000
-0.0000 10.0000

$P_{\text{plus_inf_std}} =$

2.8965 + 0.0000i 0.0000 + 0.0000i
0.0000 + 0.0000i 3.1623 + 0.0000i

$X_{\text{hat_error_std1}} =$

9.3206 11.1483

There is a discrepancy for the second state.

*This is believed to be because of the constant input not = 0,
so the steady-state Kalman filter for that state is not as accurate.*

The alternative method of simulation:

$F =$

0.5000 2.0000
0 1.0000

$G =$

0
1

$P_{\text{minus_inf}} =$

58.1889 26.1130
26.1130 22.0991

$k_{\text{inf}} =$

0.8533
0.3830

$P_{plus_inf} =$

8.5335 3.8295
3.8295 12.0991

$P_{plus_inf_std} =$

2.9212 1.9569
1.9569 3.4784

$X_{hat_error_std2} =$

8.8327 10.9833

The alternative method has a discrepancy for the food term as well
Problem 2 -----
Discretized System:

$F =$

0.9930 0.0997
-0.0997 0.9950

$G =$

0.0060
0.0998

$H =$

1 0
0 1

$Q =$

0.9000

Standard diviations of errors

$Y_{error_std} =$

0.8822

$X_{hat_error_std} =$

0.1330

Problem 3 -----

Part a -----

$P_{ss_a} =$

2.3868	0.2774
0.2774	4.2188

$P_{ss_a_care} =$

2.3868	0.2774
0.2774	4.2188

Part b -----

$P_{ss_b} =$

0.2899	0.5798
0.5798	1.1596

$P_{ss_b_care} =$

0.2899	0.5798
0.5798	1.1596

Part c -----

$P_{ss_c} =$

2.2899	0.5798
0.5798	3.1596

$P_{ss_c_care} =$

2.2899	0.5798
0.5798	3.1596

Part d -----

$P_{ss_d} =$

0.6899	1.3798
1.3798	2.7596

$P_{ss_d_care} =$

2.2899	0.5798
--------	--------

0.5798 3.1596

$K_{int} =$

0.6899 1.3798
1.3798 2.7596

$A_{KC_{int}} =$

0.3101 -1.3798
-1.3798 -1.7596

$eig_{int} =$

-2.4495
1.0000

This is an unstable observer, so it does not result in a steady-state Kalman Filter

Problem 4 -----

$sys =$

$A =$

	x_1	x_2
x_1	0	1
x_2	-36	-1.92

$B =$

	u_1
x_1	0
x_2	1

$C =$

	x_1	x_2
y_1	1	0
y_2	0	1

$D =$

	u_1
y_1	0
y_2	0

Continuous-time state-space model.

$Q =$

0.0050

`dt_sys =`

`A =`

	<code>x1</code>	<code>x2</code>
<code>x1</code>	<code>-0.5908</code>	<code>0.01873</code>
<code>x2</code>	<code>-0.6743</code>	<code>-0.6267</code>

`B =`

	<code>u1</code>
<code>x1</code>	<code>0.04419</code>
<code>x2</code>	<code>0.01873</code>

`C =`

	<code>x1</code>	<code>x2</code>
<code>y1</code>	<code>1</code>	<code>0</code>

`D =`

	<code>u1</code>
<code>y1</code>	<code>0</code>

Sample time: 0.5 seconds

Discrete-time state-space model.

Problem 5 -----

Part a -----

`x_dot_sym =`

	<code>r_dot</code>
<code>w + r*theta_dot^2 - (G*M)/r^2</code>	<code>theta_dot</code>
	<code>-(2*r_dot*theta_dot)/r</code>

Part b -----

`eq =`

`0 == r*theta_dot^2 - (G*M)/r^2`

`theta_dot_sym =`

`(G^(1/2)*M^(1/2))/r^(3/2)`
`-(G^(1/2)*M^(1/2))/r^(3/2)`

Part c -----

`A =`

<code>[</code>	<code>0,</code>	<code>1, 0,</code>	<code>0]</code>
<code>[omega_0^2 + (2*G*M)/r_0^3,</code>	<code>0, 0,</code>	<code>2*omega_0*r_0]</code>	
<code>[</code>	<code>0,</code>	<code>0, 0,</code>	<code>1]</code>
<code>[</code>	<code>0, -(2*omega_0)/r_0,</code>	<code>0,</code>	<code>0]</code>

$L =$

0
1
0
0

$\omega_{0_const} =$

0.0012

$v_{const} =$

7.7940e+03

$A_{const} =$

1.0e+04 *

0	0.0001	0	0
0.0000	0	0	1.5588
0	0	0	0.0001
0	-0.0000	0	0

$eig_A =$

0.0000 + 0.0000i
0.0000 + 0.0012i
0.0000 - 0.0012i
0.0000 + 0.0000i

$\tau =$

5.2964e+03

$min_int_step =$

529.6442

Part d -----

The performance of the linear Kalman Filter is reduced by the lack of linearity in the actual system dynamics. This may be improved with faster

measurements and Kalman Filter updates; however this is not guaranteed.

Part e -----

As is clear between the two estimate error plots, the Extended Kalman Filter

has a lot less error then the Linear one.This is very evident from the order of magnitude on each of the plots.

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