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Class: GEOPH 605

Consider the following quadratic polynomial

$$y(t) = m_1 + m_2 t - \frac{1}{2} m_3 t^2$$

Estimate:  $m_1, m_2, m_3$

Observed altitude:  $y(t_j), j=1, \dots, m$

1. Matrix vector form

(a) When  $m=3$ , altitude observed at  $t_1, t_2$  and  $t_3$

$$\begin{bmatrix} 1 & t_1 & -\frac{1}{2} t_1^2 \\ 1 & t_2 & -\frac{1}{2} t_2^2 \\ 1 & t_3 & -\frac{1}{2} t_3^2 \end{bmatrix} \begin{bmatrix} m_1 \\ m_2 \\ m_3 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}$$

(b) When  $m=2$ , altitude observed at  $t_1$  and  $t_2$

$$\begin{bmatrix} 1 & t_1 & -\frac{1}{2} t_1^2 \\ 1 & t_2 & -\frac{1}{2} t_2^2 \end{bmatrix} \begin{bmatrix} m_1 \\ m_2 \\ m_3 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$$

2. Assume that

$$m_{\text{true}} = \begin{bmatrix} 0.18 \\ 16.21 \\ 9.81 \end{bmatrix}$$

```
mtrue = [0.18 16.21 9.81]' %true m
```

```
mtrue = 3×1
    0.1800
   16.2100
    9.8100
```

```
m = 20; %rows
n = 3; %columns
```

```
tj = linspace(0,3,m) %temporal domain
```

```
tj = 1×20
      0      0.1579      0.3158      0.4737      0.6316      0.7895      0.9474      1.1053 ...
```

## N0.2 a)

### (i). Form G

```
%forming G matrix
G = zeros(m,n); %m by n container
```

```
for i = 1:n
    for j = 1:m
        G(j,1) = 1;
        G(j,2) = tj(j);
        G(j,3) = -0.5*tj(j)^2;
    end
end
disp(G)
```

```
1.0000      0      0
1.0000    0.1579  -0.0125
1.0000    0.3158  -0.0499
1.0000    0.4737  -0.1122
1.0000    0.6316  -0.1994
1.0000    0.7895  -0.3116
1.0000    0.9474  -0.4488
1.0000    1.1053  -0.6108
1.0000    1.2632  -0.7978
1.0000    1.4211  -1.0097
1.0000    1.5789  -1.2465
1.0000    1.7368  -1.5083
1.0000    1.8947  -1.7950
1.0000    2.0526  -2.1066
1.0000    2.2105  -2.4432
1.0000    2.3684  -2.8047
1.0000    2.5263  -3.1911
1.0000    2.6842  -3.6025
1.0000    2.8421  -4.0388
1.0000    3.0000  -4.5000
```

### (ii). Form dtrue

```
dtrue = G*mtrue;
disp(dtrue)
```

```
0.1800
2.6172
4.8098
```

```
6.7579
8.4613
9.9202
11.1346
12.1043
12.8295
13.3102
13.5462
13.5377
13.2846
12.7869
12.0447
11.0579
9.8265
8.3506
6.6301
4.6650
```

### (iii). Form noisy data d

```
mean = 0;
std = 2;

%noise
noise = 2*randn(m,1);

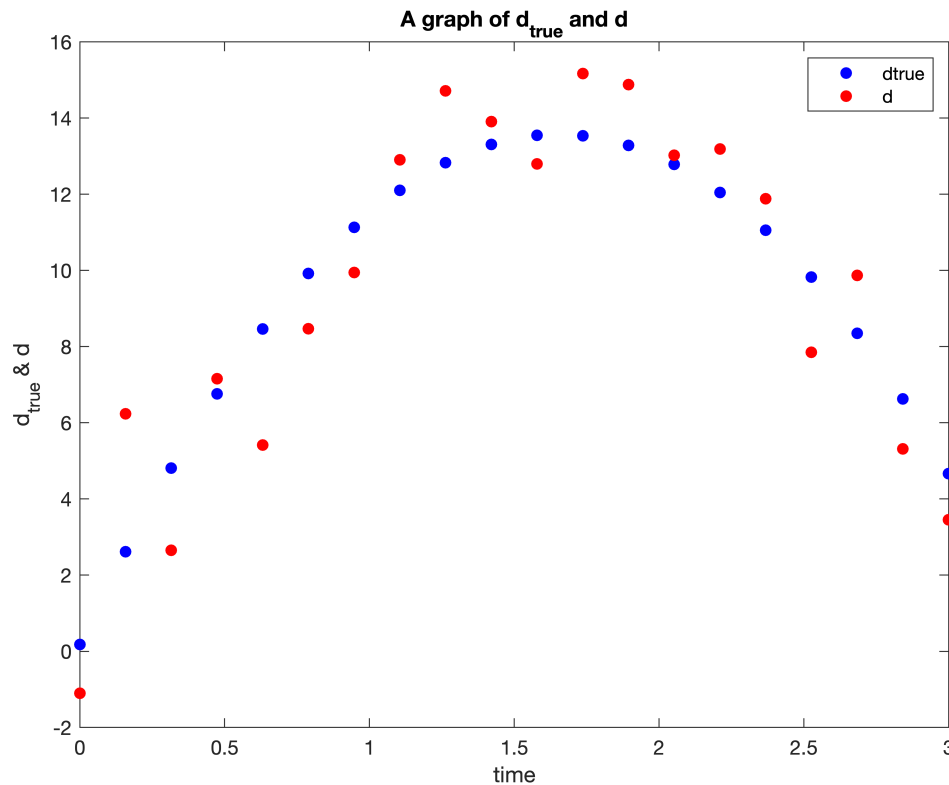
%noise data
d = dtrue + noise;
disp(d)
```

```
-1.1014
6.2349
2.6501
7.1562
5.4193
8.4730
9.9481
12.9070
14.7138
13.9111
12.8001
15.1687
14.8824
13.0273
13.1872
11.8835
7.8526
9.8697
5.3157
3.4572
```

### Plot dtrue and d as points on the same graph

```
figure(1)
plot(tj,dtrue,'b.','MarkerSize',20)
hold on
plot(tj,d,'r.','MarkerSize',20)
title('A graph of d_{true} and d')
legend('dtrue','d')
xlabel('time');
```

```
ylabel('d_{true} & d')
```



### Discuss the difference between the data

Comparing the true data,  $d_{true}$ , (blue) and the noisy data,  $d$ , (red), it is clearly shown that  $d$  has a lot of dispersed data points due to the effect of noise while  $d_{true}$  is a smooth curve without any disturbances. Even though the noisy data is distorted, it still depicts the trend of the true data.

### N0.2 b)

Solve  $Gm = d$  for  $m$  and plot  $y(t)$  using  $m$  and  $m_{true}$

```
m = G\d
```

```
m = 3x1
    -0.6776
    17.8078
    10.8129
```

```
%function
y = @(t,m1,m2,m3) m1 + m2*t - 0.5*m3*t.^2;

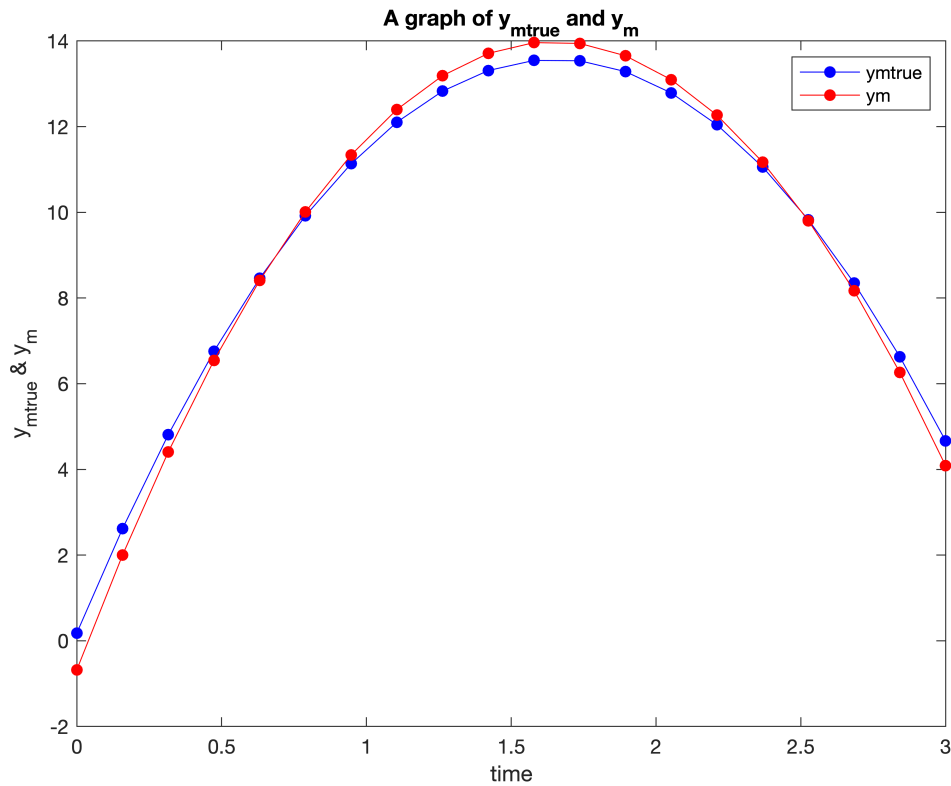
%using mtrue
ymtrue = y(tj, mtrue(1), mtrue(2), mtrue(3));
%using m
ym = y(tj, m(1), m(2), m(3));

%plot y
figure(2)
```

```

plot(tj,ymtrue,'b.-','MarkerSize',20)
hold on
plot(tj, ym,'r.-','MarkerSize',20)
title('A graph of  $y_{mtrue}$  and  $y_m$ ')
legend('ymtrue','ym')
xlabel('time');
ylabel('y_{mtrue} & y_m')

```



### Discuss the difference between the trajectories.

There is a discrepancy between the trajectories:  $y_{mtrue}$  (blue) and  $y_m$  (red) obtained using  $m_{true}$  and  $m$  values respectively. This is due to the noisy data  $d$ , used to obtain  $m$  and the forward model  $(y(t))$  used to obtain  $y$  which is approximate, hence even the trajectory won't fit the observed trajectory ( $y_{true}$ ) due to the distorted values of  $m$ . Therefore for whatever values of the  $m$ , the trajectories will never fit each other but will always have the same trend.