# Homework I

#### **Question 1:**

- (a) we use the function in Matlab rand() to randomly assign masses, m = rand(1,100)
- **(b)** The problem can be restated as a set of equations,

$$\begin{cases} m_1 = d_1 \\ m_1 + m_2 = d_2 \\ m_1 + m_2 + m_3 = d_3 \\ m_2 + m_3 + m_4 = d_4 \\ \dots \\ m_{98} + m_{99} + m_{100} = d_{100} \end{cases}$$

Which can be rewritten as matrix forms, and the size is  $100 \times 100$ .

$$\begin{bmatrix} 1 & 0 & 0 & 0 & \dots & 0 & 0 \\ 1 & 1 & 0 & \dots & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & \dots & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & 1 & 1 & \dots & 0 \\ 0 & 0 & \dots & 0 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} m_1 \\ m_2 \\ m_3 \\ m_4 \\ \dots \\ m_{99} \\ m_{100} \end{bmatrix} = \begin{bmatrix} d_1 \\ d_2 \\ d_3 \\ d_4 \\ \dots \\ d_{99} \\ d_{100} \end{bmatrix}$$

And G is,

$$\begin{bmatrix} 1 & 0 & 0 & 0 & \dots & 0 & 0 \\ 1 & 1 & 0 & \dots & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & \dots & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & 1 & 1 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 0 & 1 & 1 & 1 \end{bmatrix}$$

- (c) we use the function in Matlab to create observed data with Gaussian random numbers, dobs = G1\*m' + normrnd(0, sigmad, 100, 1)
- (d) Solve the inverse problem by simple least squares,

$$m_{est} = (G^T G)^{-1} (G^T d)$$

- (e) The variance of the estimated model parameters can use the command, var = std2(mest), and different calculation will return different values.
- **(f)** The counted number is 100 after calculating.

#### **Question 2:**

This method can be restated as follows,

$$\begin{cases} m_1 = d_1 \\ m_1 + m_2 = d_2 \\ m_1 + m_2 + m_3 = d_3 \\ m_1 + m_2 + m_3 + m_4 = d_4 \\ \dots \\ m_1 + \dots + m_{98} + m_{99} + m_{100} = d_{100} \end{cases}$$

The second weighing method will construct a different kernel G, which is quite easy, a lower triangular matrix.

$$G = \begin{bmatrix} 1 & 0 & \dots & 0 & 0 \\ 1 & 1 & \dots & 0 & 0 \\ \vdots & \ddots & \vdots & \vdots \\ 1 & 1 & \dots & 1 & 0 \\ 1 & 1 & \dots & 1 & 1 \end{bmatrix}$$

The same as the Method 1, After calculating the variance of the estimated model parameters, we get the answer for (f), counted number is 100.

# **(g)** In fig.1 and fig.2 we can see the difference.

According to the two figures below, comparing the two ways to weigh a set of boxes, we can find the second way can be better, for the inverse value fits the initial value well.

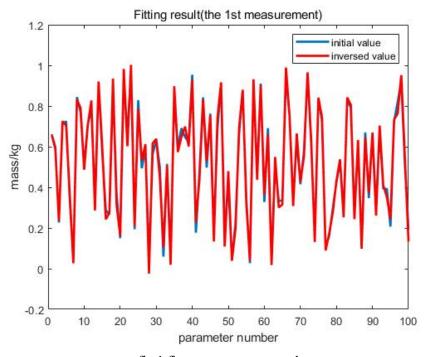


fig.1 first measurement result

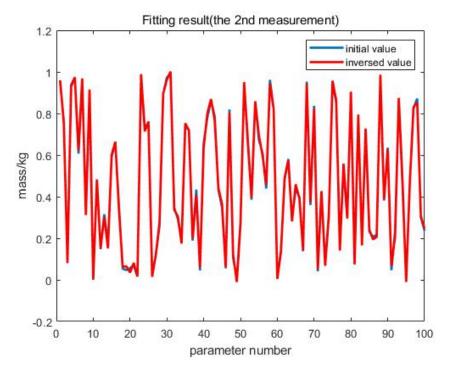


fig.2 second measurement result

### **Question 3:**

### Consider the cubic equation and answer the questions.

The problem can be restated as a set of equations as follows,

$$d_j = m_1 + m_2 z_j + m_3 z_j^2 + m_4 z_j^3, j = 1,2,3,...,11$$

which can be rewritten as the matrix form,

$$\begin{bmatrix} 1 & z_1 & z_1^2 & z_1^3 \\ 1 & z_2 & z_2^2 & z_2^3 \\ \dots & \dots & \dots & \dots \\ 1 & z_{11} & z_{11}^2 & z_{11}^3 \end{bmatrix} \begin{bmatrix} m_1 \\ m_2 \\ m_3 \\ m_4 \end{bmatrix} = \begin{bmatrix} d_1 \\ d_2 \\ \dots \\ d_{11} \end{bmatrix}$$

$$G = \begin{bmatrix} 1 & z_1 & z_1^2 & z_1^3 \\ 1 & z_2 & z_2^2 & z_2^3 \\ \dots & \dots & \dots & \dots \\ 1 & z_{11} & z_{11}^2 & z_{11}^3 \end{bmatrix}$$

We use the function in Matlab: normrnd(0,0.05,11,1) to create synthetic data with Gaussian random numbers with zero means and  $\sigma_d = 0.05$ .

The formula of simple least squares method is,

$$m_{est} = (G^T G)^{-1} (G^T d)$$

And the predicted data can be calculated by,

$$d_{mre} = Gm_{est}$$

Finally, we get the figure below.

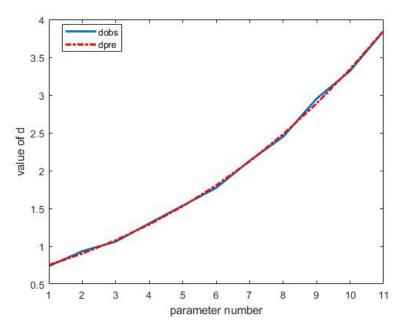


fig.3 plot in question 3(no constraint added)

## **Question 4:**

Add a constraint that the predicted data pass through a fixed point, and go through the steps above in Question 3.

We can solve this problem only to reconstruct the kernel G, let the row 5 in G be zero. In this way, the fifth element in vector d can always be the value we want.

And the result(fig.4) is as follows,

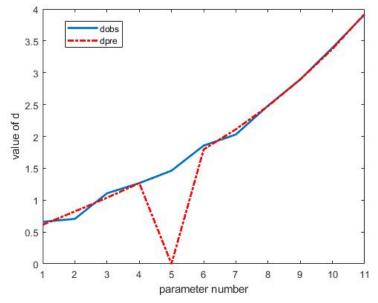


fig.4 plot in question 4(constraint added)

## **Programming code**

```
(1) question1:
clear;
clc;
m = rand(1,100);%randomly assigns masses mtrue in the range of 0-1kg
% build the appropriate kernel G
% G is a 100*100 matrix
G1 = zeros(100,100);
G1(1,1) = 1;
G1(2,1) = 1;
G1(2,2) = 1;
for i = (3:100)
G1(i,i-2) = 1;
G1(i,i-1) = 1;
G1(i,i) = 1;
end
G = sparse(G1);
% create synthetic observed data by adding Gaussian random vector
sigmad = 0.01;
dobs = G1*m' + normrnd(0, sigmad, 100, 1);
%solve the inverse problem by simple least squares
mest = (G'*G)\setminus(G'*dobs);
\% to calculate the variance of each of the estimated model parameters
var = std2(mest);
sigmam = sqrt(var);
count = 0;
for j = (1:length(mest))
a = mest(j);
b = m(j);
if abs(a-b) <= 2*sigmam</pre>
count = count + 1;
end
end
fprintf("the number of estimated model parameters that are within 2σ of their
true value is %d\n",count)
%draw the picture
x = [1:100];
clf;
plot(x,m,x,mest,'r-','linewidth',2),title('Fitting result(the 1st
measurement)'),xlabel('parameter number'),ylabel('mass/kg'),legend('initial
value','inversed value')
```

```
(2) question 2:
clear;
clc;
m = rand(1,100);%randomly assigns masses mtrue in the range of 0-1kg
% build the appropriate kernel G
% G is a 100*100 matrix
G1 = tril(ones(100));
G = sparse(G1);
% create synthetic observed data by adding Gaussian random vector
sigmad = 0.01;
dobs = G1*m' + normrnd(0, sigmad, 100, 1);
%solve the inverse problem by simple least squares
mest = (G'*G)\setminus(G'*dobs);
% to calculate the variance of each of the estimated model parameters
var = std2(mest);
sigmam = sqrt(var);
count = 0;
for j = (1:length(mest))
a = mest(j);
b = m(j);
if abs(a-b) <= 2*sigmam</pre>
count = count + 1;
end
end
fprintf("the number of estimated model parameters that are within <math>2\sigma of their
true value is %d\n",count)
%draw the picture
figure(2)
x = [1:100];
clf;
plot(x,m,x,mest,'r-','linewidth',2),title('Fitting result(the 2nd
measurement)'),xlabel('parameter number'),ylabel('mass/kg'),legend('initial
value','inversed value')
(3) question 3:
clear;
clc;
%construct a vector z with 11 elements equally spaced 0.1
z = [1:0.1:2];
%randomly assigns elements of mtrue
m = [0.1, 0.3, -0.2, 0.5]';
```

```
%build the appropriate kernel G
G = zeros(11,4);
G(:,1) = ones(1,11)';
G(:,2) = z';
G(:,3) = z'.^2;
G(:,4) = z'.^3;
%create synthetic data with Gaussian random numbers
mean = 0;
sigmad = 0.05;
dobs = G*m + normrnd(mean, sigmad, 11, 1);
%solve the inverse problem by simple least squares
mest = (G'*G)\setminus(G'*dobs);
dpre = G*mest;
figure(3)
clf;
x = [1:11]';
plot(x,dobs,x,dpre,'r-.','linewidth',2),legend('dobs','dpre')
(4) question 4:
clear;
clc;
%construct a vector z with 11 elements equally spaced 0.1
z = [1:0.1:2];
%randomly assigns elements of mtrue
m = [0.1, 0.3, -0.2, 0.5]';
%build the appropriate kernel G
G = zeros(11,4);
G(:,1) = ones(1,11)';
G(:,2) = z';
G(:,3) = z'.^2;
G(:,4) = z'.^3;
%create synthetic data with Gaussian random numbers
mean = 0;
sigmad = 0.05;
dobs = G*m + normrnd(mean, sigmad, 11, 1);
%solve the inverse problem by simple least squares
%add the prior constriants
```

```
G(5,:) = zeros(1,4);
mest = (G'*G)\(G'*dobs);

dpre = G*mest;
figure(3)
clf;
x = [1:11]';
plot(x,dobs,x,dpre,'r-.','linewidth',2),legend('dobs','dpre')
```