

# Lab 1b: Recovering matrix from data

MATH 5110: Applied Linear Algebra and Matrix Analysis

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## 1 Introduction

The goal of this Lab is to learn how to construct a matrix which will map a given set of input vectors to a given set of output vectors.

### 1.1 Modeling with a linear map

We are given a ‘black box’ that takes an input vector  $\vec{x}$  and produces an output vector  $\vec{b}$ . We want to model the black box using a linear map  $L$ , so the linear map should satisfy

$$L(\vec{x}) = \vec{b} \tag{1}$$

Of course if the black box is actually nonlinear then this cannot work for all pairs of input/output vectors. However if we limit the number of pairs of vectors  $(\vec{x}, \vec{b})$  for which we want the relation (1) to hold, then it is possible to find such a map  $L$  even in the case where the black box is nonlinear.

To be precise about this, suppose that both the input and output vectors for the black box are  $n$ -dimensional vectors for some integer  $n$ . That is  $\vec{x} \in \mathbb{R}^n$  and  $\vec{b} \in \mathbb{R}^n$ . Then  $L$  should be a linear map from  $\mathbb{R}^n$  to  $\mathbb{R}^n$ . Every linear map from  $\mathbb{R}^n$  to itself is equal to multiplication by a  $n \times n$  matrix, and we want to find this matrix.

In order to handle the problem that the black box may not be linear, we limit the number of pairs of input/output vectors to be exactly  $n$ . That is we assume that we are given  $n$  input vectors  $\vec{x}^{(1)}, \dots, \vec{x}^{(n)}$  and  $n$  output vectors  $\vec{b}^{(1)}, \dots, \vec{b}^{(n)}$  such that for every  $k = 1, \dots, n$  we have

$$L(\vec{x}^{(k)}) = \vec{b}^{(k)} \tag{2}$$

Let us write  $A$  to denote the  $n \times n$  matrix which implements the linear map  $A$ . So we want to find the matrix  $A$  such that

$$L(\vec{x}^{(k)}) = A \vec{x}^{(k)} = \vec{b}^{(k)} \quad \text{for } k = 1, \dots, n \tag{3}$$

Counting the number of parameters shows that this should be possible: there are  $n$  vector equations in (3), and hence  $n^2$  equations in total. The matrix  $A$  has  $n^2$  entries, and these are the unknown values that we want to find. So we have  $n^2$  equations with  $n^2$  unknowns.

## 1.2 Special case $n = 2$

Let's focus for now on the case  $n = 2$  in order to keep the notation simpler. And let's write the two input vectors as  $\vec{x}, \vec{y}$  and the output vectors as  $\vec{b}, \vec{c}$ . So we have the two vector equations

$$A \vec{x} = \vec{b}, \quad A \vec{y} = \vec{c} \quad (4)$$

We will write these equations more explicitly. First we have

$$\vec{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, \quad \vec{b} = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}, \quad \vec{y} = \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}, \quad \vec{c} = \begin{pmatrix} c_1 \\ c_2 \end{pmatrix} \quad (5)$$

and also

$$A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \quad (6)$$

Then the four equations (4) are

$$\begin{aligned} a_{11}x_1 + a_{12}x_2 &= b_1 \\ a_{21}x_1 + a_{22}x_2 &= b_2 \\ a_{11}y_1 + a_{12}y_2 &= c_1 \\ a_{21}y_1 + a_{22}y_2 &= c_2 \end{aligned} \quad (7)$$

We rewrite the first and third equations as a matrix equation:

$$\begin{pmatrix} x_1 & x_2 \\ y_1 & y_2 \end{pmatrix} \begin{pmatrix} a_{11} \\ a_{12} \end{pmatrix} = \begin{pmatrix} b_1 \\ c_1 \end{pmatrix} \quad (8)$$

We define

$$M = \begin{pmatrix} x_1 & x_2 \\ y_1 & y_2 \end{pmatrix}, \quad C_1 = \begin{pmatrix} a_{11} \\ a_{12} \end{pmatrix}, \quad D_1 = \begin{pmatrix} b_1 \\ c_1 \end{pmatrix} \quad (9)$$

so that the equation can be written

$$MC_1 = D_1 \quad (10)$$

Assuming that  $\text{rank}(M) = 2$ , there is a unique vector  $C_1$  which is the solution of (10). From (9) we see that this vector  $C_1$  gives us the first row of the unknown matrix  $A$ .

We apply similar reasoning with the second and fourth equations in (7): define

$$C_2 = \begin{pmatrix} a_{21} \\ a_{22} \end{pmatrix}, \quad D_2 = \begin{pmatrix} b_2 \\ c_2 \end{pmatrix} \quad (11)$$

then we have

$$MC_2 = D_2 \quad (12)$$

Note that the same matrix  $M$  appears again. Therefore we can solve for  $C_2$  which gives us the row of  $A$ .

To summarize: by solving (10) and (12) we get the column vectors  $C_1$  and  $C_2$ . By taking their transposes we get the rows of  $A$ , so we have

$$A = \begin{pmatrix} C_1^T \\ C_2^T \end{pmatrix} \quad (13)$$

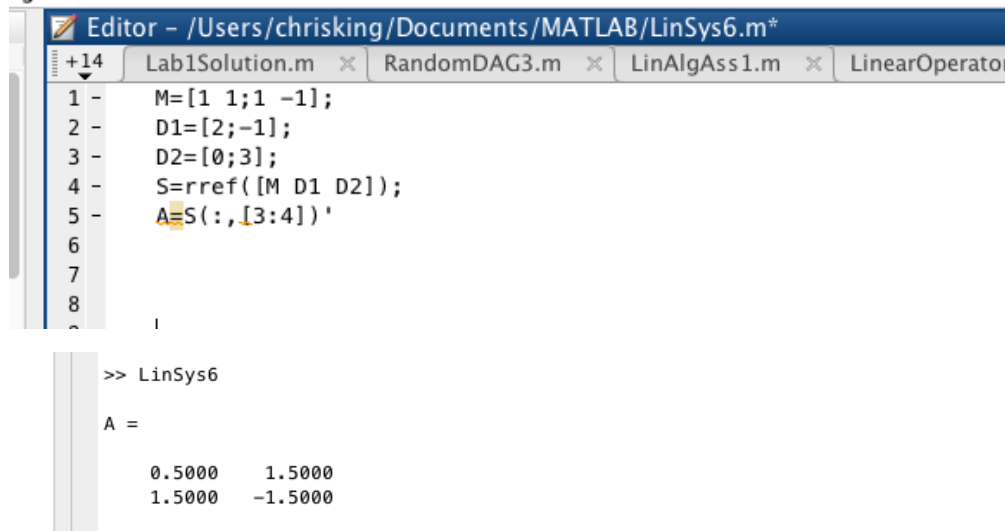
### 1.3 Example for $n = 2$

Let's see a concrete example. Suppose that

$$\vec{x} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \quad \vec{b} = \begin{pmatrix} 2 \\ 0 \end{pmatrix}, \quad \vec{y} = \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \quad \vec{c} = \begin{pmatrix} -1 \\ 3 \end{pmatrix} \quad (14)$$

Then we get

$$M = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}, \quad D_1 = \begin{pmatrix} 2 \\ -1 \end{pmatrix}, \quad D_2 = \begin{pmatrix} 0 \\ 3 \end{pmatrix} \quad (15)$$



The image shows a MATLAB Editor window with the following script in `LinSys6.m`:

```

1 - M=[1 1;1 -1];
2 - D1=[2;-1];
3 - D2=[0;3];
4 - S=rref([M D1 D2]);
5 - A=S(:,[3:4])'
```

The Command Window shows the execution of `>> LinSys6` resulting in the matrix `A`:

```

A =
    0.5000    1.5000
    1.5000   -1.5000
```

We can check that this is correct, namely that

$$A\vec{x} = A \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 & 3 \\ 3 & -3 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 2 \\ 0 \end{pmatrix} = \vec{b}, \quad (16)$$

and similarly for  $A\vec{y}$ .

### 1.4 Independence of input and output vectors

The existence of the solutions  $C_1$  and  $C_2$  depended on  $M$  being nonsingular, that is having rank equal to 2. This is equivalent to the independence of its columns, and also equivalent to the independence of its rows. The rows of  $M$  are the vectors  $\vec{x}$  and  $\vec{y}$ . So the condition that we can solve the system of equations to find  $A$  is the condition that the input vectors are independent. This condition must be satisfied for all  $n$  in the general case.

What happens if the vectors  $\vec{b}$  and  $\vec{c}$  are dependent? Then the vectors  $C_1$  and  $C_2$  will also be dependent, so the matrix  $A$  will be singular, and its rank will be less than two.

### 1.5 The setup for general dimension $n$

We follow the analysis given above and extend to the case of general dimension  $n$ . So now we have  $n$  input and  $n$  output vectors, and we want to find the  $n \times n$  matrix  $A$ . The main new ingredient is developing adequate notation. First define the input vectors as

$$\vec{x}^{(1)}, \dots, \vec{x}^{(n)} \in \mathbb{R}^n \quad (17)$$

and similarly the output vectors

$$\vec{b}^{(1)}, \dots, \vec{b}^{(n)} \in \mathbb{R}^n \quad (18)$$

The equations we need to solve are

$$A\vec{x}^{(k)} = \vec{b}^{(k)}, \quad k = 1, \dots, n \quad (19)$$

Proceeding as in the case  $n = 2$ , we rewrite these as another set of equations where the unknown vectors are the rows of  $A$ . The input vectors will again be combined into the matrix  $M$ . However recall that the *rows* of  $M$  in (9) were the input vectors. The same thing happens now, and the  $n$  input vectors are put in the rows of our  $n \times n$  matrix  $M$ :

$$M = \begin{pmatrix} (\vec{x}^{(1)})^T \\ (\vec{x}^{(2)})^T \\ \vdots \\ (\vec{x}^{(n)})^T \end{pmatrix} \quad (20)$$

Note that while  $\vec{x}^{(1)}$  is a column vector ( $n \times 1$ ), its transpose  $(\vec{x}^{(1)})^T$  is a row vector ( $1 \times n$ ).

Next we see how the output vectors enter the story. To set up the notation we write the components of each vector as

$$\vec{b}^{(k)} = \begin{pmatrix} b_1^{(k)} \\ b_2^{(k)} \\ \vdots \\ b_n^{(k)} \end{pmatrix}, \quad k = 1, \dots, n \quad (21)$$

Then we define the vector

$$D_1 = \begin{pmatrix} b_1^{(1)} \\ b_1^{(2)} \\ \vdots \\ b_1^{(n)} \end{pmatrix} \quad (22)$$

Notice that we have taken the *first* component of each vector and combined these all together to make the column vector  $D_1$ . We proceed similarly with  $D_2$ , using the second components of all the output vectors, and so on. So we have

$$D_k = \begin{pmatrix} b_k^{(1)} \\ b_k^{(2)} \\ \vdots \\ b_k^{(n)} \end{pmatrix} \quad \text{for all } k = 1, \dots, n \quad (23)$$

Notice again that the  $k$ th entries of the output vectors are combined to create  $D_k$ . Then we consider the  $n$  vector equations

$$MC_k = D_k, \quad k = 1, \dots, n \quad (24)$$

The solutions of these equations give the rows of  $A$ , and we get

$$A = \begin{pmatrix} C_1^T \\ C_2^T \\ \vdots \\ C_n^T \end{pmatrix} \quad (25)$$

## 2 Data example

The data is drawn from measurements of PM2.5 pollution in Beijing over several years in the early twenty-first century. [Source: the UCI archive. <https://archive.ics.uci.edu/ml/index.php> ]

We use 8 data characteristics measured at 4 time points. The characteristics are:

- Month (represented by integer in  $\{1, \dots, 12\}$ )
- Time of Day (represented by hour in  $[0, 24]$ )
- Temperature (degrees Celcius)
- Relative humidity (percentage)
- CO concentration (mg per cubic m)
- benzene concentration ( $\mu\text{g}$  per cubic m)
- nitrogen oxide (parts per billion)
- nitrogen dioxide ( $\mu\text{g}$  per cubic m)

and the data values are

<i>Characteristic</i>	<i>Time 1</i>	<i>Time 2</i>	<i>Time 3</i>	<i>Time 4</i>
<i>Month</i>	3	5	7	9
<i>TOD</i>	0	6	12	18
<i>Temp</i>	11.3	14.5	36.9	25.3
<i>Rel.Hum.</i>	56.8	78.3	17.2	33
<i>CO</i>	1.2	1.3	1.5	5.6
<i>C6H6</i>	3.6	6.9	8.3	31
<i>NOX</i>	62	108	78	578
<i>NO2</i>	77	81	94	204

### 2.1 First task

Your first task is to use this data to build a linear model whose inputs are the (Month, TOD, Temp, Rel Hum) for the 4 times, and whose outputs are the corresponding values of (CO, C6H6, NOX, NO2). So you should find the  $4 \times 4$  matrix which implements the linear map from these 4 input characteristics to the 4 output characteristics.

## 2.2 Second task

There is missing data for a fifth time, see below. Use the linear model to fill in the missing first four items.

<i>Characteristic</i>	<i>Time 1</i>	<i>Time 2</i>	<i>Time 3</i>	<i>Time 4</i>	<i>Time 5</i>
<i>Month</i>	3	5	7	9	
<i>TOD</i>	0	6	12	18	
<i>Temp</i>	11.3	14.5	36.9	25.3	
<i>Rel.Hum.</i>	56.8	78.3	17.2	33	
<i>CO</i>	1.2	1.3	1.5	5.6	1.7
<i>C6H6</i>	3.6	6.9	8.3	31	9.3
<i>NOX</i>	62	108	78	578	62
<i>NO2</i>	77	81	94	204	66