

Matrix Analysis with Applications
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Lecture - 30
Introduction to Ill-Conditioned Systems

Hello friends. So, welcome to the lecture on Introduction to Ill-Conditioned Systems. So, in this lecture we will learn when a system will be called ill-conditions, apart from that we will learn how to measure the ill-conditioning of a given system. So, for introducing the ill-conditioned system let us take one example.

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Ex-1

$$x_1 + x_2 = 2$$

$$x_1 + 1.0001x_2 = 2.0001$$

$$\begin{bmatrix} 1 & 1 \\ 1 & 1.0001 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 2.0001 \end{bmatrix}$$

the solution of this system is $x_1 = 1$ and $x_2 = 1$.

Now, make a small change in the one of entry of the right hand side vector.

$$\begin{bmatrix} 2 & 2.0001 \end{bmatrix}^T \rightarrow \begin{bmatrix} 2 & 2 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 \\ 1 & 1.0001 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

the solution is $x_1 = 2$ and $x_2 = 0$

Ex-2 consider 2x2 system

$$\begin{bmatrix} 400 & -201 \\ -800 & 401 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 200 \\ -200 \end{bmatrix}$$

$x_1 = -100$ and $x_2 = -200$

Make a small change in an entry of the coefficient matrix ($a_{11} = 400 \Rightarrow a_{11} = 401$)

the new system is

$$\begin{bmatrix} 401 & -201 \\ -800 & 401 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 200 \\ -200 \end{bmatrix}$$

the solution of the new system becomes

$x_1 = 40000$ and $x_2 = 79800$

So, consider a 2 by 2 system $x_1 + x_2 = 2$ and then $x_1 + 1.0001x_2 = 2.0001$. So, here the coefficient matrix is $\begin{bmatrix} 1 & 1 \\ 1 & 1.0001 \end{bmatrix}$ the unknown variable column is $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ and right hand side column vector is $\begin{bmatrix} 2 \\ 2.0001 \end{bmatrix}$ now the solution of this system is $x_1 = 1$ and $x_2 = 1$ this you can verify now make a small change in the in one of the entry of the right hand side vector that is you are having earlier the vector $\begin{bmatrix} 2 \\ 2.0001 \end{bmatrix}$ transpose.

Let us say due to some sensor error or due to some computer what I will say round of sort of thing this 2.0001 will be read as 2. So, now, the new system is $\begin{bmatrix} 1 & 1 \\ 1 & 1.0001 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$ if we see the solution of this system the solution is $x_1 = 2$ and $x_2 = 0$. So, what we have seen if I make a small change in one of the entry of the

right hand side vector then my solution is having a large change because earlier solution was x_1 equals to 1 x_2 equals to 1.

But due to this small change now it will become x_1 equals to 2 and x_2 equals to 0 in the same way take one more example. So, here I am having again a 2 by 2 system where the coefficient matrix is $\begin{bmatrix} 4 & 0 \\ 0 & 1 \end{bmatrix}$ minus $\begin{bmatrix} 2 & 0 \\ 1 & 800 \end{bmatrix}$ this is the coefficient matrix the vector is again x_1 x_2 and the right hand side vector is 200 and minus 200.

In the example one we have made a small change in the in entry of the right hand side vector now let us make a small change in one of the entry from the coefficient matrix. So, this is the 2 by 2 system now if I look for the solution of this system the solution is given as x_1 equals to minus 100 and x_2 equals to minus 200 now if I make a small change in an entry of the coefficient matrix because in the earlier example I have made change in the right hand side vector now I am making the change in coefficient matrix.

So, let us say earlier a 1 1 entry was $\begin{bmatrix} 4 & 0 \\ 0 & 0 \end{bmatrix}$ now make it a 1 1 as $\begin{bmatrix} 4 & 0 \\ 1 & 1 \end{bmatrix}$. So, the new system is $\begin{bmatrix} 4 & 0 \\ 1 & 1 \end{bmatrix}$ minus $\begin{bmatrix} 2 & 0 \\ 1 & 800 \end{bmatrix}$ and then $\begin{bmatrix} 4 & 0 \\ 1 & 1 \end{bmatrix}$ together with x_1 x_2 and right hand side vector is the same 200 and minus 200. So, now, the solution of this new system becomes x_1 equals to 40000 and x_2 equals to 79000 and 800.

So, now you just notice I am having only a small change earlier it was 400 now it becomes $\begin{bmatrix} 4 & 0 \\ 1 & 1 \end{bmatrix}$ how much change I am getting in my solution earlier x_1 was minus 100 now x_1 become 40000 x_2 was minus 200 now x_2 becomes 79000 and 800. So, what we have observed that if you make a small change either in an entry of the right hand side vector or in an entry of the coefficient matrix a very small change tiny change.

We are getting a very large change in our solution so, such type of systems are called ill-conditioned systems.

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If we make small change in my input data (an entry of right hand side vector or coefficient matrix) and get a large change in the solution, then the system is called ill-conditioned system.

Measure of ill-conditioning :-

Conditional number.. Let A be an invertible matrix, then the conditional number of A is denoted by $K(A)$,

$$K(A) = \|A\| \cdot \|A^{-1}\|$$

→ When the conditional number $K(A)$ of a ~~matrix~~ the coefficient matrix becomes large, the system $AX=b$ is regarded as ill-conditioned.

→ If $K(A)$ is near to 1, then the system is called well conditioned.

So, if we make a small change in my input data for the linear system it will become an entry of right hand side vector or coefficient matrix and get a large change in the solution which is output then the system is called ill-conditioned systems. Now, let us how to measure the ill-conditioning of a system for that we need to define conditional number. So, let A be an invertible matrix then the conditional number of A is denoted by K_A and it is defined as K_A equals to the product of norm of A with norm of A inverse now when the conditional number K_A of a matrix or of the coefficient matrix A will write becomes large the system AX equals to b is regarded as ill-conditioned.

So, you find out the conditional number of the coefficient matrix and if it is very large then we will say the system is ill-conditioned if the conditional number of A where A is the coefficient matrix is near to one then the system is called well conditioned. So, hence we can measure the ill-conditioning of a system by calculating the conditional number and if it is quite large then the system is ill-conditioned.

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Ex. 1

$$A = \begin{bmatrix} 1 & 1 \\ 1 & 1.0001 \end{bmatrix}; \quad A^{-1} = \begin{bmatrix} 10001 & -10000 \\ -10000 & 10001 \end{bmatrix}$$

$$\|A\| = 2.0001; \quad \|A^{-1}\| = 20001$$

$$K(A) = \|A\| \|A^{-1}\| \approx 40002 \quad (\text{large, that's why the system } Ax=b \text{ is ill conditioned})$$

Alternate definition:-
 If $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n$ be the singular values of an invertible matrix A, then

$$K(A) = \frac{\sigma_1}{\sigma_n}$$

$$A = V S V^T \Rightarrow S = \begin{bmatrix} 2.0005 & 0 \\ 0 & 0.0005 \end{bmatrix}$$

$$K(A) = \frac{2.0005}{0.0005} \approx 4002$$

So, if we see those 2 examples. So, in example 1 my coefficient matrix was one 1 1 and 1.0001. So, if I calculate A inverse here A inverse comes out to be 10000 101 and then minus 10000 minus 10000 and then finally, 10000. So, here norm of a will become the maximum of sum of the rows of A. So, I am taking the row norm. So, it will be 2.0001 and if I calculate a norm of a inverse this comes out to be 20001.

So, now if I find out the conditional number of a this will be norm of a into norm of A inverse which comes out to be something around 42002 almost of this 1. So, hence point something like this. So, hence it is quite large and that is why the system ax equals to b is ill-conditioned similar kind of analysis we can make for another matrix alternate definition for conditional number is if be the singular values of an invertible matrix a.

So, all will be greater than 0 then conditional number of A is the ratio of the largest singular values upon the smallest singular values. So, if we see in the above example the singular values of a will be. So, if a equals to U S V transpose then S will be 2.000 0 5 0 0 and 0.0005. So, from here conditional number a will be 2.0005 upon 0.00005 and which will be the same 40002 which you have obtained with the definition norm of a into norm of A inverse. So, by this alternate definition you can calculate the conditional number if you know the singular values of A given matrix.

Now, let us make some investigation why we are having this change in what I will say the change in solution a large change in the solution by making a very tiny change in the

input data means either an entry of coefficient matrix or an entry of right hand side vector.

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Investigation:-

① If we have a small change $b + \Delta b$ in b

Let x be the solution of the original system $Ax = b$.
 By making a small change in b ($b \rightarrow b + \Delta b$), we get
 new solution as $x + \Delta x$.

$$A(x + \Delta x) = b + \Delta b$$

$$\Rightarrow Ax + A\Delta x = b + \Delta b$$

$$\Rightarrow A\Delta x = \Delta b$$

$$\Rightarrow \Delta x = A^{-1} \Delta b$$

$$\Rightarrow \|\Delta x\| = \|A^{-1} \Delta b\|$$

$$\leq \|A^{-1}\| \|\Delta b\| \text{--- (1)}$$

Moreover,

$$\|Ax\| \leq \|A\| \|x\|$$

$$\|b\| \leq \|A\| \|x\| \text{--- (2)}$$

or

$$\|A\| \|x\| \geq \|b\| \text{--- (2)}$$

Dividing (1) by (2), we get

$$\frac{\|\Delta x\|}{\|A\| \|x\|} \leq \frac{\|A^{-1}\| \|\Delta b\|}{\|b\|}$$

$$\frac{\|\Delta x\|}{\|x\|} \leq \underbrace{\|A^{-1}\| \|A\|}_{K(A)} \cdot \frac{\|\Delta b\|}{\|b\|}$$

\Rightarrow If $K(A)$ is very large, then the product of small Δb with $K(A)$ will be significant.

So, first we will check if you we have a small change in b plus delta b in b means in right hand side vector. So, let X be the solution of the original system AX equals to b by making a small change in b that is b becomes b plus delta b we get new solution as X plus delta X . So, what we are having $A X$ plus delta X equals to b plus delta b ok.

So, this I can write ax plus A into delta X equals to b plus delta b , but as you know this ax equals to b . So, this ax I can replace with b . So, from here I am getting A into delta and b will be cancel out from both side A into delta X equals to delta b or from here delta x equals to a inverse into delta b or norm of delta x will be norm of A inverse into delta b this equals less than equals to norm of a inverse, because norm of A into b will be less than equals to norm of A into norm of b into norm of delta b . So, this let us say equation 1 moreover we have norm of AX is less than equals to norm of A into norm of X and X equals to b .

So, norm of b less than equals to norm of A into norm of X let us say 2 or I can write in other way norm of a into norm of X is greater than equals to norm of b . So, let us say this is my equation 2 dividing 1 by 2 we get norm of delta X upon norm of A into norm of X because left hand side of 1 is delta X and left hand side of 2 is norm of a into norm of X and this will be less than equals to norm of a inverse into norm of b norm of a inverse

into norm of b delta b , sorry because it is delta b here upon norm of b . If I multiply the this inequality by norm of A I will get norm of delta X upon norm of X in the left hand side which is less than equals to norm of A into norm of A inverse and norm of A into norm of A inverse is my conditional number into delta norm of delta b upon norm of b .

So, from here you can observe if I am if the conditional number is very large if I am having a small delta b I will get a large delta X . This means if K is very large then the product of small delta b with large K A will be significant and that is why I will get a significant delta X means significant change in my final solution. So, this is the investigation when I am making the change in the right hand side vector and from here we have seen that why for the large conditional number a small change in right hand side vector will give a large change in the solution.

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$$\begin{aligned}
 \textcircled{11} \quad & A \rightarrow A + \Delta A \\
 & X \rightarrow X + \Delta X \\
 & (A + \Delta A)(X + \Delta X) = b \\
 \Rightarrow & \frac{\|\Delta X\|}{\|X + \Delta X\|} \leq \underline{K(A)} \cdot \frac{\|\Delta A\|}{\|A\|} \\
 \Rightarrow & \text{-----} X \text{ -----} X \text{ -----} .
 \end{aligned}$$

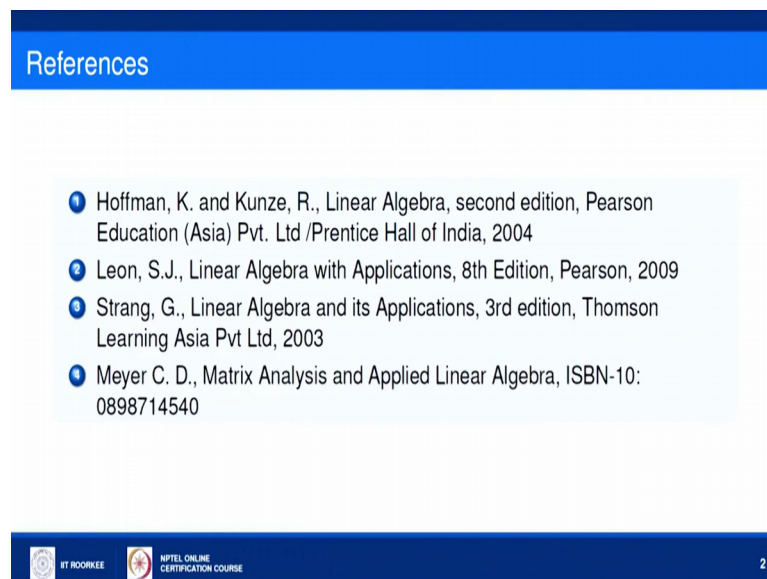
Now, take the second case when I am making a small change in my coefficient matrix means if I am having the original coefficient matrix A . And now I am making a small change and it is becoming a plus delta A and if earlier my solution was X now it becomes X plus delta X then what I will be having A plus delta A into X plus delta X equals to b , because there is no change in the right hand side vector by making the same kind of analysis which we have done earlier in the first case I will get delta X upon X plus delta X will be less than equals to conditional number of A into delta norm of delta A upon norm of A .

So, again if conditional number of A is quite large a small δA multiplied with a large KA will give you a very large value and that is why I am getting a large change in δX . So, this 2 kind of investigations still such that why conditional number is very important for ill-conditioning and if conditional number is near about 1 then a small change multiplied with an entry which is quite close to 1 will give you the small change δX .

In this lecture we have learnt what are the ill-conditioning systems what is conditional number what is the relation between conditional number and singular values and we have investigated why the system becomes ill-conditioned when we are having the large conditional number.

In the next lecture we will learn how to solve such type of ill-conditioned system by defining a proper regularization term.

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References

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So, these are the references.

Thank you very much.