Lecture 05





Math 22 Summer 2017 Section 2 June 30, 2017

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- Finish up §1.5
- Part of §1.6 on network flows
- §1.7 Linear Independence







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Suppose now we choose a specific A in RREF given by

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How many free variables do we have in this case? As we've seen before with a single free variable, we can write a general solution to this system using a **parametric vector equation**...



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$\mathbf{x} =$	<i>x</i> 3	$= x_{3}$	1	$+ x_{5}$	0	
	-3 <i>x</i> 5		0		-3	
	<i>X</i> 5		0		1	

What is the geometric interpretation of the solution set?



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What is the geometric interpretation of the solution set? Now let $\mathbf{b} = \begin{bmatrix} 1\\1\\0\\0 \end{bmatrix}$ and consider the *nonhomogeneous linear system* $A\mathbf{x} = \mathbf{b}$.





For A and **b** defined above, we see that a general solution to $A\mathbf{x} = \mathbf{b}$ is given by



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§1.5 Theorem 6





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Thus $\mathbf{v}_h := \mathbf{w} - \mathbf{p}$ satisfies $A\mathbf{x} - \mathbf{0}$ and $\mathbf{w} = \mathbf{p} + \mathbf{v}_h$. This shows that $\mathbf{w} \in S$ and concludes the proof.



Consider the following network flow diagram.



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Can you see how this defines a linear system?







```
500 + 300 + 100 + 400 = 300 + x_3 + 600.
```



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Now we can use linear algebra to answer questions about the network!



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٢0	0	1	0	0	400		[10	0	0	1	600
1	1	0	0	0	800		01	. 0	0	-1	200
0	1	-1	1	0	300	\sim	00) 1	0	0	400
0	0	0	1	1	500		00	0	1	1	500
1	0	0	0	1	600		0 0	0	0	0	0



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So a general solution to this linear system is given by

$$\begin{cases} x_1 = 600 - x_5 \\ x_2 = 200 + x_5 \\ x_3 = 400 \\ x_4 = 500 - x_5 \\ x_5 & \text{free} \end{cases}$$



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When dealing with network flows, a general solution of this form is called a **general flow pattern** for the network.





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§1.6 Network flows



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The above constraint on the *free* variable x_5 allows us to get conditions on the other variables written in terms of x_5 . In particular, what can we say about x_1 ?



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Definition



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§1.7 Example





$$\left\{ \mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3 \right\} = \left\{ \begin{bmatrix} 1\\0 \end{bmatrix}, \begin{bmatrix} 0\\1 \end{bmatrix}, \begin{bmatrix} 2\\-7 \end{bmatrix} \right\}.$$



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Is S linearly independent?



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Is S linearly independent? No.



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Is S linearly independent? No. $2\mathbf{v}_1 - 7\mathbf{v}_2 - \mathbf{v}_3 = \mathbf{0}$ is a dependence relation among these vectors. What about subsets of S? How can we be more systematic about this?

§1.7 Linear independence and $A\mathbf{x} = \mathbf{0}$





Let $\mathbf{a}_1, \ldots, \mathbf{a}_n \in \mathbb{R}^m$.



Let $\mathbf{a}_1, \ldots, \mathbf{a}_n \in \mathbb{R}^m$. How can we tell if these vectors form a linearly independent set?



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But as we saw, this is equivalent to asking about solutions to the matrix equation $A\mathbf{x} = \mathbf{0}$ where $A = [\mathbf{a}_1 \cdots \mathbf{a}_n]$.



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Let's do an example with 4 vectors in \mathbb{R}^3 .

§1.7 Theorem 8





Theorem

If a set contains more vectors than there are entries in each vector, then the set is linearly dependent.



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Proof.

See previous slide.

§1.7 More examples





Is the set $\{\boldsymbol{0}\}$ linearly independent?


Is the set $\{\boldsymbol{0}\}$ linearly independent? No.







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Could all of the scalars on the RHS of the equation be zero? No! Since we assumed $\mathbf{v}_j \neq 0$. Thus, dividing by c_j we get \mathbf{v}_j as a linear combination of the other vectors.



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Suppose $v\neq 0$ and we want to find a vector w so that $\{v,w\}$ is linearly independent. By the theorem, such a w cannot be in the span of v which is just all scalar multiples of v.

§1.7 Classwork





Find all values of $h \in \mathbb{R}$ for which the vectors

$$\mathbf{v}_1 = \begin{bmatrix} 1\\ -1\\ 4 \end{bmatrix}, \mathbf{v}_2 = \begin{bmatrix} 3\\ -5\\ 7 \end{bmatrix}, \mathbf{v}_3 = \begin{bmatrix} -1\\ 5\\ h \end{bmatrix}$$

form a linearly dependent set.



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So the set is dependent if and only if h = 6.
§1.7 Classwork





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§1.7 Classwork





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Yes! Although this seems like a trivial result, the significance is that if we have a space that is spanned by vectors, we can eliminate redundant vectors until we have a linearly independent set.



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Yes! Although this seems like a trivial result, the significance is that if we have a space that is spanned by vectors, we can eliminate redundant vectors until we have a linearly independent set. Such a set is called a **basis**.