MATH 18.700, DAY 1 SYSTEMS OF LINEAR EQUATIONS

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CONTENTS

I. PRE-CLASS PLANNING

I.1. **Goals for lesson.**

- (1) Students will understand expectations for the course.
- (2) Students will understand the definition of a system of linear equations.
- (3) Students will learn the elementary row operations.
- (4) Students will learn how to row reduce a matrix.
- (5) Students will learn how to write solutions in parametric form.

I.2. **Methods of assessment.**

- (1) Student responses to questions posed during lecture
- (2) Student responses to questions posed during lecture

I.3. **Materials to bring.** (1) Laptop + adapter (2) Colored chalk

II. LESSON ^PLAN **(0:00)**

Hi everyone and welcome to 18.700: Linear Algebra. I'm Sam. I'm a postdoc here at MIT and I'll be your instructor for this course. [Name game. Remind students about online survey.]

II.1. **Organizational details.** Before we talk about math, I'd like to cover some organizational details and show you the course website. [Show the website. Mention Canvas. Mention possible office hours: Monday 1 - 2pm, Tuesday 2 - 3pm. Ask if there are conflicts for the midterm exams. Mention that first homework assignment will be posted later today, due next Wednesday.]

(0:10)

(0:05)

II.2. **Why linear algebra?** Motivating question: How do we solve systems of equations? Answer:

- Arbitrary equations? Hard! Numerical analysis provides approximate solutions.
- Polynomial equations? Still hard! Computational algebraic geometry and Gröbner bases can sometimes provide solutions, but as the number of variables and degree increases, these methods become impractical.
- Linear equations, i.e., polynomial equations of degree 1. This is what we'll study.

Roughly speaking, a linear system is one in which the output is proportional to the input: an input twice as large yields an output twice as large. Linear systems are very well understood thanks to the success of the theory of linear algebra. Even when considering a nonlinear system, we often approximate it with a linear one. One example of this idea of linear approximation is the derivative in calculus.

Linear algebra is also used in virtually every branch of math: graph theory, functional analysis, differential geometry, algebraic geometry, algebraic number theory, differential equations. It is also used in virtually every area in science.

(0:15)

II.3. **What is linear algebra?** Short answers:

- The study of solutions sets to systems of linear equations and the transformations of these solution sets.
- The study of finite-dimensional vector spaces and linear maps.

Notation: Unless otherwise mentioned, throughout this course, the symbol **F** will stand for either **R** or **C**, the sets of real or complex numbers, respectively. (We use **F** because **R** and **C** are examples of *fields*, algebraic structures in which one can add, subtract, multiply, and divide.

Definition 1. A *linear equation* in the variables x_1, \ldots, x_n is an equation that can be written in the form

$$
a_1x_1+\cdots+a_nx_n=b
$$

for some coefficients $a_1, \ldots, a_n, b \in \mathbb{F}$. In other words, the lefthand side must be a polynomial of degree 1 in x_1, \ldots, x_n .

For instance, $4x_1x_2 + x_3 = 7$ and $\sqrt{x_1} - 7x_2 = 0$ are not linear. [Ask students: why?]

Definition 2.

• A *system of linear equations* or *linear system* is a collection of one or more linear equations (in the same variables). [Give example]

- A *solution* to a linear system is a list (s_1, \ldots, s_n) of numbers that satisfy each equation, i.e., that make each equation true when s_1, \ldots, s_n are substituted for x_1, \ldots, x_n .
- The *solution set* is the set of all possible solutions. Two linear systems are *equivalent* if they have the same solution set. **(0:25)**

Example 1. Intersection of two lines.

$$
\begin{array}{rcl}\nx_1 & -x_2 & = & 0 \\
2x_1 & +x_2 & = & 3\n\end{array}
$$

[Draw picture of lines.] The (unique) solution is $(1, 1)$. An equivalent system is

$$
\begin{array}{rcl}\nx_1 & = & 1 \\
x_2 & = & 1\n\end{array}
$$

[Ask students: Can you think of two lines that have no points of intersection? What does the picture look like?] There is one last possibility, illustrated by the following linear system:

$$
\begin{array}{rcl}\nx_1 & -x_2 & = & 1 \\
2x_1 & -2x_2 & = & 2\n\end{array}
$$

These two equations define the same line, so every point on this line is a solution. Thus there are infinitely many solutions.

Remark 1. This is a phenomenon that holds in general for linear systems: a linear system has either

- (1) exactly one solution,
- (2) infinitely many solutions, or
- (3) no solutions.

We will prove this later.

Definition 3. A linear system is *consistent* if it has a solution (at least one). If it has no solutions, it is *inconsistent*.

II.4. **Matrix notation.** Rewriting all the variables in a linear system can be tedious. We can instead record the relevant information more compactly in a *matrix* (a rectangular array of numbers) as follows. Consider the following linear system.

$$
\begin{array}{rcl}\n-2x_1 & +2x_2 & +10x_3 & = & 2 \\
-3x_1 & +x_2 & +9x_3 & = & 5 \\
4x_1 & -2x_2 & -14x_3 & = & -6\n\end{array}
$$

The *coefficient matrix* of the system is

$$
\begin{pmatrix}\n-2 & 2 & 10 \\
-3 & 1 & 9 \\
4 & -2 & -14\n\end{pmatrix}
$$

The i , j entry is the coefficient of x_i in the j^{th} equation of the system. The *augmented matrix* of the system is

$$
\begin{pmatrix}\n-2 & 2 & 10 & 2 \\
-3 & 1 & 9 & 5 \\
4 & -2 & -14 & -6\n\end{pmatrix}
$$

(0:35)

where the last column contains the constants from the righthand side of the equations.

II.5. **Solving a linear system.** [Switch to slides. Do an example of 3×3 example that has 1-dimensional solution space.]

$$
\begin{array}{rcl}\n-2x_1 & +2x_2 & +10x_3 & = & 2 \\
-3x_1 & +x_2 & +9x_3 & = & 5 \\
4x_1 & -2x_2 & -14x_3 & = & -6\n\end{array}
$$

$$
\begin{pmatrix}\n-2 & 2 & 10 & 2 \\
-3 & 1 & 9 & 5 \\
4 & -2 & -14 & -6\n\end{pmatrix}\n\rightsquigarrow\n\begin{pmatrix}\n1 & -1 & -5 & -1 \\
-3 & 1 & 9 & 5 \\
4 & -2 & -14 & -6\n\end{pmatrix}\n\rightsquigarrow\n\begin{pmatrix}\n1 & -1 & -5 & -1 \\
0 & -2 & -6 & 2 \\
0 & 2 & 6 & -2\n\end{pmatrix}
$$
\n
$$
\rightsquigarrow\n\begin{pmatrix}\n1 & -1 & -5 & -1 \\
0 & -2 & -6 & 2 \\
0 & 0 & 0 & 0\n\end{pmatrix}\n\rightsquigarrow\n\begin{pmatrix}\n1 & -1 & -5 & -1 \\
0 & 1 & 3 & -1 \\
0 & 0 & 0 & 0\n\end{pmatrix}
$$
\n
$$
\rightsquigarrow\n\begin{pmatrix}\n1 & 0 & -2 & -2 \\
0 & 1 & 3 & -1 \\
0 & 0 & 0 & 0\n\end{pmatrix}
$$
\n
$$
\implies\n\begin{pmatrix}\nx_1 \\
x_2 \\
x_3\n\end{pmatrix}\n=\n\begin{pmatrix}\n2t - 2 \\
-3t - 1 \\
t\n\end{pmatrix}\n=\n\begin{pmatrix}\n2 \\
-3 \\
1\n\end{pmatrix}t +\n\begin{pmatrix}\n-2 \\
-1 \\
0\n\end{pmatrix}
$$

[Ask students what geometric shape a given equation looks like. Hint: should be familiar from multivariable calc. Show picture of the planes intersecting in a line.]

In solving this system, we employed three basic operations, called *elementary row operations*:

(0:40)

(1:00) Elementary Row Operations:

- (1) (Add a multiple): Add a multiple of one row to another
- (2) (Swap) Swap two rows
- (3) (Rescale) Multiply all entries in a row by a nonzero constant.

We used these row operations to solve the above linear system, a process called *row reduction* or *Gaussian elimination*. We will explore this process in greater detail shortly. Two important questions about a linear system:

- 1) Does the system have a solution, i.e., is it consistent?
- 2) If so, is the solution unique? How many solutions are there?

(1:05)

II.6. **Echelon forms.** We saw an example of how we can use elementary row operations to take a given linear system and produce an equivalent but simpler linear system. Now we will formally write down this algorithm.

Definition 4.

- A row of a matrix is *nonzero* if at least one entry in the row is not zero. We define nonzero columns similarly.
- The *leading entry* of a nonzero row is the left-most nonzero entry.

Definition 5.

- A matrix is in *echelon form* (or *row echelon form*) if it satisfies the following three properties:
	- **–** All nonzero rows are above any rows of all zeros. (rows of all zeros are at the bottom)
	- **–** Each leading entry of a row is in a column to the right of the leading entry of the row above it.
	- **–** All entries in a column below a leading entry are zero.
- A matrix is in *reduced echelon form* (or *reduced row echelon form*), often abbreviated RREF, if it is in echelon form and satisfies two additional conditions:
	- **–** The leading entry in each nonzero row is a 1.
	- **–** Each leading 1 is the only nonzero entry in its column.
- [Give examples and non-examples.] **(1:15)**

Theorem (Uniqueness of reduced echelon form)**.** *Every matrix is row equivalent to a* unique *reduced echelon matrix.*

Remark 2. Although an echelon form of a matrix is not unique, the positions of the leading entries are uniquely determined.

Definition 6.

- A *pivot position* in a matrix *A* is a location in *A* corresponding to a leading 1 in the reduced echelon form for *A*.
- A *pivot column* is a column of *A* that contains a pivot position.
- A *pivot* is a nonzero entry in a pivot position.

II.7. **Row reduction algorithm.** The row reduction algorithm consists of 5 steps:

- (1) Take the leftmost nonzero column. This is the leftmost pivot column. Make the pivot position at the top of the column.
- (2) Select any nonzero entry in the leftmost pivot column as a pivot and move it into the pivot position (top) interchanging rows as necessary.
- (3) Use row operations to make all entries under the pivot zero.
- (4) Consider the submatrix obtained by removing the row and column containing the pivot. Go to the first step and repeat until there is no submatrix left to consider.
- (5) Use row operations to create zeros above each pivot.

[Label (1) - (4) as echelonization, (5) as reduction. Also, DON'T ERASE THE ALGO-

RITHM!] **(1:25)**

Example 2. Let's row reduce the following augmented matrix. [Ask students. Explicitly point out the various steps during the example.]

(Should get

\n
$$
\begin{pmatrix}\n0 & 0 & 3 & | 9 \\
1 & 5 & -2 & | 2 \\
1/3 & 2 & 0 & | 3\n\end{pmatrix}
$$
\n(Should get

\n
$$
\begin{pmatrix}\n1 & 0 & 0 & | 3 \\
0 & 1 & 0 & | 1 \\
0 & 0 & 1 & | 3\n\end{pmatrix}
$$
\nfor the RREF

for the RREF.)

(1:20)

Remark 3. There may be many ways to obtain the RREF of a matrix.

- II.8. **Worksheet.**
- **(1:40)**

(1:30)

II.9. **Parametric form for solutions.** Consider the augmented matrix

$$
\begin{pmatrix} 1 & 0 & 3 & 1 \ 0 & 1 & -2 & 5 \ 0 & 0 & 0 & 0 \end{pmatrix}.
$$

This corresponds to the linear system

$$
\begin{array}{|l|}\n x_1 & +3x_3 = 1 \\
 x_2 & -2x_3 = 5\n \end{array}\n \longrightarrow\n \begin{array}{|l|}\n x_1 = 1 - 3x_3 \\
 x_2 = 5 + 2x_3\n \end{array}
$$

The variables corresponding to pivot columns (i.e x_1 and x_2) are called *basic* or *leading variables*.

The other variables (i.e., *x*3) are called *free variables*.

We can use this description to write the set of solutions parametrically:

$$
\begin{cases}\nx_1 = 1 - 3x_3 \\
x_2 = 5 + 2x_3 \\
x_3 \text{ free}\n\end{cases}
$$

We call this the *parametric description* of the solutions. We can express this in vector notation as

$$
\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 - 3t \\ 5 + 2t \\ t \end{bmatrix} = \begin{bmatrix} -3t \\ 2t \\ t \end{bmatrix} + \begin{bmatrix} 1 \\ 5 \\ 0 \end{bmatrix} = \begin{bmatrix} -3 \\ 2 \\ 1 \end{bmatrix} t + \begin{bmatrix} 1 \\ 5 \\ 0 \end{bmatrix}.
$$

[Ask students how many solutions this system has.]

(1:45)

II.10. **Worksheet part 2.**

(1:50)

II.11. **Existence and uniqueness of solutions.**

Theorem.

(1) A linear system is consistent iff the right most column of the augmented matrix is not *a pivot column, i.e., iff an echelon form for the matrix does* not *contain a row of the form*

$$
\begin{bmatrix} 0 & \cdots & 0 & |b| \end{bmatrix}
$$

with b nonzero.

- *(2) If the linear system is consistent, then the solution set contains either (i) a unique solution, if there are no free variables*
	- *(ii) infinitely many solutions, if there is one or more free variables.*

[Remark on proofs briefly.]

(1:55)

II.12. **Summary.** Solving a linear system via row reduction:

- (1) Write the augmented matrix of the system.
- (2) Use the row reduction algorithm to put the matrix in echelon form. If the matrix is inconsistent, stop; otherwise continue to the next step.
- (3) Keep row reducing to put the matrix in reduced echelon form.
- (4) Write the linear system corresponding to the reduced echelon form.
- (5) Express the solution to the system in parametric form.