## Math 314 Matrix Theory

January 25, 2005

Vector equations 
$$x_1 \begin{bmatrix} a_{11} \\ \vdots \\ a_{m1} \end{bmatrix} + \cdots + x_n \begin{bmatrix} a_{1n} \\ \vdots \\ b_m \end{bmatrix} = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}$$
 can be useful in understand solutions to linear systems. One reason for this is that

addition and scalar multiplication are so well-beahved (because these operations are caried out component by component):

$$\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$$
  $(\mathbf{u} + \mathbf{v}) + \mathbf{w} = \mathbf{u} + (\mathbf{v} + \mathbf{w})$   $c(\mathbf{u} + \mathbf{v}) = c\mathbf{u} + c\mathbf{v}$   $\mathbf{u} + \mathbf{0} = \mathbf{u} + (0, \dots, 0) = \mathbf{u}$   $(cd)\mathbf{u} = c(d\mathbf{u})$   $1\mathbf{u} = \mathbf{u}$  with  $-\mathbf{u} = (-1)\mathbf{u}$ ,  $\mathbf{u} + (-\mathbf{u}) = (0, \dots, 0)$ 

The *span* of a collection of vectors is the collection of all linear combinations of the vectors;

$$\mathbf{Sp}(\mathbf{v_1},\ldots,\mathbf{v_n}) = \{x_1\mathbf{v_1} + \cdots + x_n\mathbf{v_n} : x_1,\ldots x_n \in \mathbf{R}\}$$

Then a vector equation  $x_1\mathbf{v_1} + \cdots + x_n\mathbf{v_n} = \mathbf{b}$  has a solution precisely when  $\mathbf{b} \in \mathbf{Sp}(\mathbf{v_1}, \dots, \mathbf{v_n})$ 

We can therefore understand linear systems (via vector equations) better, by understanding what the span of the column vectors of the coefficient matrix might look like. This will be a point of view that we will continue to develop throughout the course.

## Matrix equations:

There is still a third point of view that we will approach systems of equations from: matrix multiplication. We can interpret a linear combination

of vectors, 
$$x_1 \begin{bmatrix} a_{11} \\ \vdots \\ a_{m1} \end{bmatrix} + \cdots + x_n \begin{bmatrix} a_{1n} \\ \vdots \\ a_{mn} \end{bmatrix}$$
, as a product of the matrix  $A = \begin{bmatrix} a_{11} & \cdots & \vdots \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix}$  and the vector  $\mathbf{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$ , which we denote  $A\mathbf{x}$ .

In this notation a system of equations has a very compact form:  $A\mathbf{x} =$ 

This is really just a new notation for systems, but it will turn out to be remarkably useful. One reason for its utility is that the matrix  $A(c\mathbf{u} = c(A\mathbf{u}))$ product is *linear* (in the vector term):  $A(\mathbf{u} + \mathbf{v}) = A\mathbf{u} + A\mathbf{v}$ 

With this new notation, our basic goal becomes: understand the solutions  $\mathbf{x}$  to the equation  $A\mathbf{x} = \mathbf{b}$ .

Another example of how these different perspectives give different ways to view the same result:

If 
$$A = \begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix}$$
,  $v_1, \ldots, v_n$  are the column vectors of  $A$ , and  $\mathbf{b} = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}$ , then the system of equations  $\begin{pmatrix} a_{11} & \cdots & a_{1n} & b_1 \\ \vdots & & \vdots & & \vdots \\ a_{m1} & \cdots & a_{mn} & b_m \end{pmatrix}$  has a solution for every  $\mathbf{b}_1, \ldots, \mathbf{b}_m \Leftrightarrow A\mathbf{x} = \mathbf{b}_1$  has a solution for every  $\mathbf{b}_1, \ldots, \mathbf{b}_m \Leftrightarrow A\mathbf{x} = \mathbf{b}_1$  has a solution for every  $\mathbf{b}_1, \ldots, \mathbf{b}_m \Leftrightarrow A\mathbf{x} = \mathbf{b}_1$  has a solution for every  $\mathbf{b}_1, \ldots, \mathbf{b}_m \Leftrightarrow A\mathbf{x} = \mathbf{b}_1$  has a solution for every  $\mathbf{b}_1, \ldots, \mathbf{b}_m \Leftrightarrow A\mathbf{x} = \mathbf{b}_1$  has a solution for every  $\mathbf{b}_1, \ldots, \mathbf{b}_m \Leftrightarrow A\mathbf{x} = \mathbf{b}_1$ 

 $\mathbf{Sp}(v_1,\ldots,v_n)=\mathbf{R}^m$ .

These, in turn, are true  $\Leftrightarrow$  after row reducing A to RREF, every row has a pivot in it. To see this, note that if  $(A|\mathbf{b})$  is row reduced, then a pivot in every row means that there is no row  $(0 \cdots 0|1)$ , (because we can't have the row of 0's in the coefficient matrix), so the system is consistent, so there is a solution. Conversely, if the RREF does not have a pivot in every row, then its bottom row will be a row of 0's. Then

if we start with the inconsistent system  $\begin{pmatrix} RREF & | \vdots \\ |0 \\ |1 \end{pmatrix}$  and reverse the row reduction steps, we will arrive at  $(A|\mathbf{b})$  (for some  $\mathbf{b}$ ), which

we know to be inconsistent, so this equation has no solution. So if every system has a solution, then there must be a pivot in every row.