## LECTURE 9

## Hyperplanes

Let V be a finitely generated vector space over a field  $\mathbb{F}$ . Today we will focus our attention on a special class of subsets of V. These subsets will not in general be subspaces, but they arise quite naturally in linear algebra and have a lot of nice properties.

DEFINITION 9.1. Let V be a vector space over a field  $\mathbb{F}$ , let S be a subspace of V and let  $\mathbf{p} \in V$ . Let

$$M_{\mathbf{p},S} := \{\mathbf{p} + \mathbf{s} \mid \mathbf{s} \in S\}$$

We shall refer to such a subset as a linear submanifold of V. The subspace S is called the directing subspace of  $M_{\mathbf{p}_0,S}$ . The dimension of  $M_{\mathbf{p}_0,S}$  is defined as the dimension of S.

REMARK 9.2. The notion of a manifold is actually more germane to differential geometry than linear algebra, in the geometric setting the notion of a linear manifold is akin to the notion of a linear function in Calculus - it's so simple that it's not worth discussing except as a simplifying limit.

However, because we **are not discussing** differential geometry in this course, I don't see much point in mentioning manifolds. What I think would be more helpful will be to view sets of the form  $M_{\mathbf{p},S}$  as generalizations of lines and planes in  $\mathbb{R}^3$ . Indeed, you can generate a line in  $\mathbb{R}^3$  by starting at a particular point  $\mathbf{p} \in \mathbf{R}$  and then heading off an arbitrary distance d (forwards and backwards) along a particular direction  $\mathbf{v}$ : that is to say

$$line =$$
 set of the form  $\{\mathbf{p} + t\mathbf{v} \mid \mathbf{p}, \mathbf{v} \in \mathbb{R}^3 , t \in \mathbb{R}\}$ 

Since the vectors  $\{t\mathbf{v} \mid t \in \mathbb{R}\}$  constitute the span of  $\mathbf{v}$ , such a line is a linear submanifold of  $\mathbb{R}^3$  as defined defined above. Similarly, a plane in  $\mathbb{R}^3$  is formed by starting at a particular point and then heading of an arbitrary distance in two possible directrions; i.e. a subset of  $\mathbb{R}^3$  of the form

$$plane = \{\mathbf{p} + s\mathbf{u} + t\mathbf{v} \mid s, t \in \mathbb{R}\}$$
$$= \{\mathbf{p} + s \mid s \in span(\mathbf{u}, \mathbf{v})\}$$

To underscore this simple geometric picture, I shall henceforth refer to linear submanifolds of a vector space V as a hyperplane in V.

Definition 9.3.

NOTATION 9.4 (common but abusive notation). Let V be a vector space. If S is a subspace of a vector space V, and **b** is a point of V we shall write

 $\mathbf{b} + S$ 

for the corresponding hyperplane in V. (What's abusive about this is that you can't really add a subspace to a vector; on the other hand, if you interpret this expression as adding every vector in S to **b** then it does kind of make sense.)

I note also that we have already run into hyperplanes in two particular contexts. In Lecture 6, I defined quotient spaces V/S, S being some subspace of a vector space V, as the collection of sets of the form  $\mathbf{p}_0 + S$  (=  $[\mathbf{p}_0]_S$  in the notation of Lecture 6).

In Theorem 7.7 of Lecture 7, we saw hyperplanes arise as the solution sets of linear systems

THEOREM 7.7. (rephrased) Suppose Ax = b is an  $n \times m$  linear system. Then the solution set of this linear system can be expressed as

$$\mathbf{p}_0 + S = \{\mathbf{p}_0 + \mathbf{s} \mid \mathbf{s} \in S\}$$

where  $p_0$  is any particular solution of Ax = b and S is the solution set of Ax = 0.<sup>1</sup>

The first thing to point out about hyperplanes is that in general they are not subspaces. Here is a simple counter-example. Let  $V = \mathbb{R}^2$ ,  $\mathbf{p}_0 = [1, 0]$  and let S = span([0, 1)]. Then

 $\mathbf{p}_0 + S = \{ [1, y] \mid y \in \mathbb{R} \}$ 

which cannot be a subspace of  $\mathbb{R}^2$  since it does not contain the zero vector in  $\mathbb{R}^2$ .

THEOREM 9.5. Let  $\mathbf{b} + S$  be a hyperplane in a vector space V. Then

$$S = \{ \mathbf{v} \in V \mid \mathbf{v} = \mathbf{r} - \mathbf{q} \quad , \quad \mathbf{r}, \mathbf{q} \in \mathbf{b} + S \}$$

*Proof.* Let

$$S = \{ \mathbf{v} \in V \mid \mathbf{v} = \mathbf{r} - \mathbf{q} \quad , \quad \mathbf{r}, \mathbf{q} \in \mathbf{b} + S \}$$

Suppose  $\mathbf{v} \in \widetilde{S}$ . Then there exists  $\mathbf{r} = \mathbf{b} + \mathbf{s}_1$  and  $\mathbf{q} = \mathbf{b} + \mathbf{s}_2$  in  $\mathbf{b} + S$  such that

$$v = r - q = (b + s_1) - (b + s_2) = s_1 - s_2 \in S$$

and so every element  $\mathbf{v} \in \widetilde{S}$  is also a vector of S.

Suppose on the other hand that  $\mathbf{s} \in S$ , then we can always write

$$s = s + b - b = (b + s) - (b + 0)$$

which displays **s** as an element of  $\widetilde{S}$ . We conclude  $\widetilde{S} = S$  and thus prove the theorem.

Theorem 9.2 told us how to view the solution set of homogeneous  $n \times m$  linear system is subspace of  $\mathbb{F}^m$ . The following lemma provides a converse to this result.

LEMMA 9.6. Let S be an r-dimensional subspace of a vector space V of dimension m. Then there exists a set of m - r homogeneous linear equations in m unknowns whose solution set is exactly S.

Let  $\{b_1, \ldots, b_r\}$  be a basis for S. Consider the solution space  $S^*$  of

$$b_1 \cdot x = 0$$
  

$$b_2 \cdot x = 0$$
  

$$\vdots$$
  

$$b_r \cdot x = 0$$

We first note that  $S^*$  is not likely to coincide with S, simply because for example,  $b_1 \in S$  but  $b_1 \cdot b_1 \neq 0$ . On the other hand, since the vectors  $b_i$  are all linearly independent, it follows that the coefficient matrix **A** for this linear system has rank r (since the row space of **A** will be span of the r linearly independent vectors  $b_1, \ldots, b_r$ ). So the solution space  $S^*$  of  $\mathbf{A}x = \mathbf{0}$  will be of dimension m - r. Let  $\{c_1, \ldots, c_{m-r}\}$  be the basis for the  $S^*$  and consider the system

$$c_1 \cdot x = 0$$

$$\vdots$$

$$c_{m-r} \cdot x = 0$$
(\*)

Let  $S^{**}$  denote the solution set of (\*). Clearly, each  $b_i$  will be a solution of this system, and thus so will any linear combination of the vectors  $b_i$ , and thus, the entire subspace S lie in the solution set of (\*). On the other hand, Since the vectors  $c_1, \ldots, c_{m-r}$  are linearly independent, it is clear that the rank of this

 $\Box$ 

<sup>&</sup>lt;sup>1</sup>That the solution set of a homogeneous  $n \times m$  linear system  $\mathbf{Ax} = \mathbf{0}$  is actually a *subspace* of  $\mathbb{F}^m$  is the content of Theorem 9.2.

linear system is m - r and so solution set of dimension m - (m - r) = r. But we've seen that if a subspace has same dimension as the vector space containing it, the subspace must be the whole vector space. Since  $S \subset S^{**}$  and dim $(S) = \dim(S^{**})$  we conclude that S coincides with the solution set  $S^{**}$  of (\*).

EXAMPLE 9.7. Find a homogeneous linear system whose solution set coincides with the span of [1, 0, 1] and [1, 1, 0].

• We first find a basis for the solution set of

$$0 = [1, 0, 1] \cdot x = x_1 + x_3$$
  
$$0 = [1, 1, 0] \cdot x = x_1 + x_2$$

The augmented matrix for this system is

$$\left[\begin{array}{rrrr|rrr} 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 0 \end{array}\right] \quad \rightarrow \quad \left[\begin{array}{rrrr|rrr} 1 & 0 & 1 & 0 \\ 0 & 1 & -1 & 0 \end{array}\right]$$

and so the general solution will be

$$\begin{array}{rcl} x_1 & = & -x_3 \\ x_2 & = & x_3 \end{array}$$

or

$$\mathbf{x} = \begin{bmatrix} -x_3 \\ x_3 \\ x_2 \end{bmatrix} = x_3 \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix}$$

Thus, the solution space has basis  $c_1 = [-1, 1, 1]$ . The desired homogeneous linear system will be

$$[-1, 1, 1] \cdot x = 0$$

THEOREM 9.8. A necessary and sufficient condition for a subset M of vectors to form a hyperplane in  $\mathbb{F}^m$  of dimension r is that M be the set of solutions of a system of m - r equations in m unknowns whose coefficient matrix has rank r.

*Proof.* How, a solution set of a linear system constitutes a hyperplane was explained in at the start of this lecture. To see that every hyperplane  $\mathbf{b} + S$  corresponds to a linear system, we just observe that by Lemma 10.5 the directing subspace S can be viewed as the solution set an  $(m - r) \times m$  linear system  $\mathbf{Ax} = \mathbf{0}$ . Let

$$\mathbf{b} = \mathbf{A}\mathbf{b}$$

Then any  $\mathbf{b} + \mathbf{s}$  vector in  $\mathbf{b} + S$  will satisfy

$$A(b+s) = Ab + 0 = b$$
 .

 $Av = \tilde{b}$ 

This shows that the solution of

(\*\*)

will contain  $\mathbf{b} + S$ . On the other hand, by construction  $\mathbf{y} = \mathbf{b}$  is a solution of (\*\*) and by Theorem 7.7, any other solution of  $\mathbf{Ay} = \widetilde{\mathbf{b}}$  will be of the form

 $\mathbf{b}$  + some solution of  $\mathbf{A}\mathbf{x} = 0$ 

and so any solution  $\mathbf{y}$  of (\*\*) will be of the form ,

 $\mathbf{y} = \mathbf{b} + \mathbf{s}$  ,  $\mathbf{s} \in S$  .

Therefore  $\mathbf{b} + S$  will coincide with the solutions of (\*\*).

Finally, let me describe an algorithm by which one can identify a linear system whose solution set is a given hyperplane.

We have see above that if we had a hyperplane in  $\mathbb{R}^m$  which is also a subspace S of  $\mathbb{R}^m$ , then we could construct a corresponding equation set as follows:

- find a basis {v<sub>1</sub>,..., v<sub>k</sub>} for S
  find a basis {u<sub>1</sub>,..., u<sub>k</sub>} for the solution set of the linear system

$$\mathbf{v}_1 \cdot \mathbf{x} = 0$$
$$\mathbf{v}_2 \cdot \mathbf{x} = 0$$
$$\vdots$$
$$\mathbf{v}_k \cdot \mathbf{x} = 0$$
bace S will

• The equations that cut out the subsp

$$\mathbf{u}_1 \cdot \mathbf{x} = 0 \\ \mathbf{u}_2 \cdot \mathbf{x} = 0 \\ \vdots \\ \mathbf{u}_\ell \cdot \mathbf{x} = 0$$

Now suppose we have a hyperplane in  $\mathbb{R}^m$  of the form

$$H = \mathbf{p}_0 + S \equiv \{\mathbf{p}_0 + \mathbf{s} \mid \mathbf{s} \in S\}$$

S being some subspace of  $\mathbb{R}^m$ . Suppose also that we have followed the algorithm above and found  $\ell$  vectors  $\mathbf{u}_1,\ldots,\mathbf{u}_\ell$  such that

$$\mathbf{s} \in S \iff \mathbf{u}_i \cdot \mathbf{s} = 0$$

Then each vector in H will satisfy

$$\mathbf{u}_i \cdot (\mathbf{p}_0 + \mathbf{s}) = \mathbf{u}_i \cdot \mathbf{p}_0 + \mathbf{u}_i \cdot \mathbf{s} = \mathbf{u}_i \cdot \mathbf{p}_0 + 0 = \mathbf{u}_i \cdot \mathbf{p}_0 \qquad , \quad i = 1, \dots \ell$$

And so the linear equations whose solution set is the hyperplane  $H = \mathbf{p}_0 + S$  will be

$$\begin{aligned} \mathbf{u}_1 \cdot \mathbf{x} &= \mathbf{u}_1 \cdot \mathbf{p}_0 \\ \mathbf{u}_2 \cdot \mathbf{x} &= \mathbf{u}_2 \cdot \mathbf{p}_0 \\ &\vdots \\ \mathbf{u}_\ell \cdot \mathbf{x} &= \mathbf{u}_\ell \cdot \mathbf{p}_0 \end{aligned}$$