

# Linear Algebra II

## Notes for January 25, 2010

### Sums, direct sums, and relative bases

Let  $V$  be a vector space. Recall that the *linear span* of a set of vectors  $v_1, \dots, v_k \in V$  is the set of all linear combinations  $c_1v_1 + \dots + c_kv_k$ . It is denoted by  $\text{span}(v_1, \dots, v_k)$ . Vectors  $v_1, \dots, v_k$  are linearly independent if and only if they form a basis of their linear span. Our next definition provides a generalization of what is just said, dealing with subspaces, and not vectors.

**Definition.** Let  $V_1, \dots, V_k$  be subspaces of  $V$ . Their *sum*  $V_1 + \dots + V_k$  is defined as the set of vectors of the form  $v_1 + \dots + v_k$ , where  $v_1 \in V_1, \dots, v_k \in V_k$ . The sum of the subspaces  $V_1, \dots, V_k$  is said to be *direct* if  $0 + \dots + 0$  is the only way to represent  $0 \in V_1 + \dots + V_k$  as a sum  $v_1 + \dots + v_k$ . In this case, it is denoted by  $V_1 \oplus \dots \oplus V_k$ .

**Lemma.**  $V_1 + \dots + V_k$  is a subspace of  $V$ .

*Proof.* It is sufficient to check that  $V_1 + \dots + V_k$  is closed under addition and multiplication by numbers. Clearly,

$$(v_1 + \dots + v_k) + (v'_1 + \dots + v'_k) = ((v_1 + v'_1) + \dots + (v_k + v'_k))$$

and

$$c(v_1 + \dots + v_k) = ((cv_1) + \dots + (cv_k)),$$

and the lemma follows, since each  $V_i$  is a subspace and hence closed under the vector space operations.  $\square$

- Examples.**
1. Arithmetics of subspace addition:  $V + \{0\} = \{0\} + V = V$ ,  $V + V = V$  (why?).
  2. Consider the subspaces  $U_n$  and  $U_m$  of  $\mathbb{R}^{n+m}$ , the first one being the linear span of the first  $n$  standard unit vectors, and the second one being the linear span of the last  $m$  standard unit vectors. We have  $\mathbb{R}^{n+m} = U_n + U_m = U_n \oplus U_m$ .
  3. More generally, for a collection of vectors  $v_1, \dots, v_k \in V$ , consider the subspaces  $V_1, \dots, V_k$ , where  $V_i$  consists of all multiples of  $v_i$ . Then, clearly,  $V_1 + \dots + V_k = \text{span}(v_1, \dots, v_k)$ , and this sum is direct if and only if the nonzero vectors among  $v_j$  are linearly independent.
  4. For two subspaces  $V_1$  and  $V_2$ , their sum is direct if and only if  $V_1 \cap V_2 = \{0\}$ . Indeed, if  $v_1 + v_2 = 0$  is a nontrivial representation of  $0$ ,  $v_1 = -v_2$  is in the intersection, and vice versa.

**Theorem.** If  $V_1$  and  $V_2$  are subspaces of  $V$ , we have

$$\dim(V_1 + V_2) = \dim(V_1) + \dim(V_2) - \dim(V_1 \cap V_2).$$

In particular, the sum of  $V_1$  and  $V_2$  is direct if and only if  $\dim(V_1 + V_2) = \dim(V_1) + \dim(V_2)$ .

*Proof.* Let us pick a basis  $e_1, \dots, e_k$  of the intersection  $V_1 \cap V_2$ .  $V_1 \cap V_2$  is a subspace of both  $V_1$  and  $V_2$ , and we shall extend this basis to both a basis of  $V_1$  and a basis of  $V_2$ . Let  $e_1, \dots, e_k, f_1, \dots, f_l$  and  $e_1, \dots, e_k, g_1, \dots, g_m$  be the resulting bases of  $V_1$  and  $V_2$  respectively. Let us prove that

$$e_1, \dots, e_k, f_1, \dots, f_l, g_1, \dots, g_m$$

is a basis of  $V_1 + V_2$ . It is a complete system of vectors, since every vector in  $V_1 + V_2$  is a sum of a vector from  $V_1$  and a vector from  $V_2$ , and vectors there can be represented as linear combinations of  $e_1, \dots, e_k, f_1, \dots, f_l$  and  $e_1, \dots, e_k, g_1, \dots, g_m$  respectively. To prove linear independence, let us assume that

$$a_1e_1 + \dots + a_ke_k + b_1f_1 + \dots + b_lf_l + c_1g_1 + \dots + c_mg_m = 0.$$

Rewriting this formula as  $\mathbf{a}_1\mathbf{e}_1 + \dots + \mathbf{a}_k\mathbf{e}_k + \mathbf{b}_1\mathbf{f}_1 + \dots + \mathbf{b}_l\mathbf{f}_l = -(\mathbf{c}_1\mathbf{g}_1 + \dots + \mathbf{c}_m\mathbf{g}_m)$ , we notice that on the left we have a vector from  $V_1$  and on the right a vector from  $V_2$ , so both the left hand side and the right hand side is a vector from  $V_1 \cap V_2$ , and so can be represented as a linear combination of  $\mathbf{e}_1, \dots, \mathbf{e}_k$  alone. However, the vectors on the right hand side together with  $\mathbf{e}_i$  form a basis of  $V_2$ , so there is no nontrivial linear combination of these vectors that is equal to a linear combination of  $\mathbf{e}_i$ . Consequently, all coefficients  $\mathbf{c}_i$  are equal to zero, so the left hand side is zero. This forces all coefficients  $\mathbf{a}_i$  and  $\mathbf{b}_i$  to be equal to zero, since  $\mathbf{e}_1, \dots, \mathbf{e}_k, \mathbf{f}_1, \dots, \mathbf{f}_l$  is a basis of  $V_1$ . This completes the proof of the linear independence of the vectors  $\mathbf{e}_1, \dots, \mathbf{e}_k, \mathbf{f}_1, \dots, \mathbf{f}_l, \mathbf{g}_1, \dots, \mathbf{g}_m$ .

Summing up,  $\dim(V_1) = k + l$ ,  $\dim(V_2) = k + m$ ,  $\dim(V_1 + V_2) = k + l + m$ ,  $\dim(V_1 \cap V_2) = k$ , and our theorem follows.  $\square$

In practice, it is important sometimes to determine the intersection of two subspaces, each presented as a linear span of several vectors. This question naturally splits into two different questions.

First, it makes sense to find a basis of each of these subspaces. To determine a basis for a linear span of given vectors, the most straightforward way is to form a matrix whose columns are the given vectors, and find its reduced column echelon forms (like the reduced row echelon form, but with columns). Columns of the result form a basis of our subspace.

Second, once we know a basis  $\mathbf{v}_1, \dots, \mathbf{v}_k$  for the first subspace, and a basis  $\mathbf{w}_1, \dots, \mathbf{w}_l$  for the second one, the question reduces to solving the linear system  $\mathbf{c}_1\mathbf{v}_1 + \dots + \mathbf{c}_k\mathbf{v}_k = \mathbf{d}_1\mathbf{w}_1 + \dots + \mathbf{d}_l\mathbf{w}_l$ . For each solution to this system, the vector  $\mathbf{c}_1\mathbf{v}_1 + \dots + \mathbf{c}_k\mathbf{v}_k$  is in the intersection, and vice versa.

Let us introduce another bit of linear algebra vocabulary. It will prove extremely useful for various results we are going to prove in the coming weeks.

**Definition.** Let  $V$  be a vector space, and let  $U$  be a subspace of  $V$ .

A set of vectors  $\mathbf{v}_1, \dots, \mathbf{v}_k$  is said to be complete relative to  $U$ , if every vector in  $V$  can be represented in the form  $\mathbf{c}_1\mathbf{v}_1 + \dots + \mathbf{c}_k\mathbf{v}_k + \mathbf{u}$ , where  $\mathbf{u} \in U$ .

A set of vectors  $\mathbf{v}_1, \dots, \mathbf{v}_k$  is said to be linearly independent relative to  $U$ , if  $\mathbf{c}_1\mathbf{v}_1 + \dots + \mathbf{c}_k\mathbf{v}_k = \mathbf{u}$  with  $\mathbf{u} \in U$  implies  $\mathbf{c}_1 = \dots = \mathbf{c}_k = 0$  (and, consequently,  $\mathbf{u} = 0$ ).

A set of vectors  $\mathbf{v}_1, \dots, \mathbf{v}_k$  is said to form a basis of  $V$  relative to  $U$ , if it is complete and linearly independent relative to  $U$ .

- Examples.**
1. A system of vectors is complete (linearly independent, forms a basis) in the usual sense if and only if it is complete (linearly independent, forms a basis) relative to the zero subspace  $\{0\}$ .
  2. A system of vectors  $\mathbf{v}_1, \dots, \mathbf{v}_k$  is complete relative to  $U$  if and only if  $U + \text{span}(\mathbf{v}_1, \dots, \mathbf{v}_k) = V$ .
  3. A system of vectors forms a basis relative to  $U$  if and only if it is linearly independent and  $U \oplus \text{span}(\mathbf{v}_1, \dots, \mathbf{v}_k) = V$  (in particular, that sum should be direct).
  4. If  $\mathbf{f}_1, \dots, \mathbf{f}_k$  is a basis of  $U$ , then picking a basis of  $V$  relative to  $U$  is the same as extending  $\mathbf{f}_1, \dots, \mathbf{f}_k$  to a basis of  $V$ .

To compute a basis of  $\mathbb{R}^n$  relative to the linear span of several vectors, one may compute the reduced column echelon form for the corresponding matrix (like when looking for a basis, as before), and pick the standard unit vectors corresponding to “missing leading 1’s”, that is to the rows of the reduced column echelon form which do not have leading 1’s in them. For example, in the case of the

vectors  $\begin{pmatrix} -2 \\ 1 \\ 1 \\ 0 \end{pmatrix}$  and  $\begin{pmatrix} -2 \\ 1 \\ 0 \\ 1 \end{pmatrix}$ , the reduced column echelon form is  $\begin{pmatrix} 1 & 0 \\ -\frac{1}{2} & 0 \\ 0 & 1 \\ -\frac{1}{2} & -1 \end{pmatrix}$ , so the missing leading

1’s correspond to the second and the fourth row, and the vectors  $\begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \end{pmatrix}$  and  $\begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$  form a relative

basis.