Miscellaneous Notes on Linear Algebra

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Who ever suffered from learning too much linear algebra? These notes will seek to fill in my linear algebra gaps. New inclusion: these notes will also cover any miscellaneous material I should have learned in my undergraduate analysis, abstract algebra, topology, or whatever classes but didn't. Source files: https://git.simonxiang.xyz/math_notes/files.html

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Lecture 1

Miscellaneous Topics

TODO: affine spaces, inverse function, change of variables for multiple integrals (spivak 34,67) or tu appendix, rank, nullity, binomial theorem, freed's thing, maybe topology bases, subspace/product, tychonoff, convergnece, etc

1.1 The Inverse Function Theorem

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Lecture 2

Inner-Product Spaces

What is an inner product?? Let's find out.

2.1 Inner Products

The length of a vector x is the **norm** of x, denoted ||x||. If $x = (x_1, \dots, x_n) \in \mathbb{R}^n$, we have $||x|| = \sqrt{x_1^2 + \dots + x_n^2}$. Note that the norm is not linear. For $x, y \in \mathbb{R}^n$, the **dot product** of x and y, denoted $x \cdot y$, is defined by $x \cdot y = x_1y_1 + \dots + x_ny_n$. Note that this is a number, not a vector. Clearly $x \cdot x = ||x||^2$ for all $x \in \mathbb{R}^n$, which implies $x \cdot x \ge 0$ for all $x \in \mathbb{R}^n$ ($x \cdot x = 0$ only if x is the zero vector). The map that sends $x \in \mathbb{R}^n$ to $x \cdot y$ in \mathbb{R} for fixed y is linear since \mathbb{R} is a field. The dot product is also commutative, since \mathbb{R} is.

Inner products generalize dot products. Recall that $|\lambda|^2 = \lambda \overline{\lambda}$ for $\lambda \in \mathbb{C}$. For $z \in \mathbb{C}^n$, we define the norm of z by $||z|| = \sqrt{|z_1|^2 + \cdots + |z_n|^2}$. We take the modulus of z_i since we want the result to be nonnegative. Note that $||z||^2 = z_1 \overline{z_1} + \cdots + z_n \overline{z_n}$. We want to think of $||z||^2$ as the inner product of z with itself, like in \mathbb{R}^n . This suggests we define the inner product of $w = (w_1, \cdots, w_n) \in \mathbb{C}^n$ with z as $w_1 \overline{z_1} + \cdots + w_n \overline{z_n}$. We expect the inner product of w with z equal the complex conjugate of the inner product of z with w. With this motivation in mind, let us define inner products.

Definition 2.1 (Inner product). An **inner product** on an *F*-vector space *V* is a function that takes each ordered pair (u, v) of elements of *V* to a number $\langle u, v \rangle \in F$ such that

- (i) $\langle v, v \rangle \ge 0$ for all $v \in V$; (**positivity**)
- (ii) $\langle v, v \rangle = 0$ iff v = 0; (**definiteness**)
- (iii) $\langle u + v, w \rangle = \langle u, w \rangle + \langle v, w \rangle$ for all $u, v, w \in V$; (additivity in first slot)
- (iv) $\langle av, w \rangle = a \langle v, w \rangle$ for all $a \in F$ and all $v, w \in V$; (homogeneity in first slot)
- (v) $\langle v, w \rangle = \overline{\langle w, v \rangle}$ for all $v, w \in V$. (conjugate symmetry).

For real numbers, condition (v) simply becomes $\langle v, w \rangle = \langle w, v \rangle$ for all $v, w \in V$. An **inner product space** is a vector space V along with an inner product on V.

Example 2.1. The most important example is the **Euclidian inner product** on \mathbf{F}^n (Axler uses \mathbf{F} to denote either \mathbb{C} or \mathbb{R}). We define an inner product on \mathbf{F}^n by

$$\langle (w_1, \cdots, w_n), (z_1, \cdots, z_n) \rangle = w_1 \overline{z_1} + \cdots w_n \overline{z_n}.$$

An example of another inner product on \mathbf{F}^n is defined by $\langle (w_1, \dots, w_n), (z_1, \dots, z_n) \rangle = c_1 w_1 \overline{z_1} + \dots + c_n w_n \overline{z_n}$ for c_i positive constants. The case where $c_i = 1$ for all i is simply the standard Euclidean inner product.

Example 2.2. Consider the vector space $\mathscr{P}_m(\mathbf{F})$, the polynomial ring over \mathbf{F} of polynomials with degree at most m. We can define an inner product on $\mathscr{P}_m(\mathbf{F})$ by

$$\langle p,q\rangle = \int_0^1 p(x) \overline{q(x)} dx.$$

For fixed $w \in V$, the function that takes v to $\langle v, w \rangle$ is a linear map $V \to F$. So $\langle 0, w \rangle = 0$, and by condition (v) $\langle w, 0 \rangle = 0$ as well. Furthermore, $\langle u, v + w \rangle = \langle u, v \rangle + \langle u, w \rangle$ and $\langle u, av \rangle = \overline{a} \langle u, v \rangle$ hold as well: This second condition is known as conjugate homogeneity in the second slot.

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2.2 Norms

For $v \in V$, we define the **norm** of v, denoted ||v||, by $||v|| = \sqrt{\langle v, v \rangle}$. For example, if $p \in \mathscr{P}_m(\mathbf{F})$, then $||p|| = \sqrt{\int_0^1 |p(x)|^2 \, dx}$. Some properties: ||v|| = 0 iff v = 0, and ||av|| = |a|||v||. To see this, note that $||av||^2 = \langle av, av \rangle = a\langle v, av \rangle = a\overline{a}\langle v, v \rangle = |a|^2 ||v||^2$, taking square roots gives us our result. This illustrates a general idea: working with norms squared is easier than working directly with norms.

Two vectors $u, v \in V$ are **orthogonal** if $\langle u, v \rangle = 0$. The zero vector is orthogonal to every vector, and the only vector orthogonal to itself. Assume $V = \mathbb{R}^2$, now let us state a 2500 year old theorem.

Pythagorean Theorem. If u, v are orthogonal vectors in V, then

$$||u + v||^2 = ||u||^2 + ||v||^2.$$

Proof. Exercise.

□

Suppose $u, v \in V$. We want to write u as a scalar multiple of v plus a vector w orthogonal to v. Let $a \in F$ be a scalar, then u = av + (u - av). We need to choose a such that v is orthogonal to u - av, in other words, we want $0 = \langle u - av, v \rangle = \langle u, v \rangle - a||v||^2$. So we should choose $a = \langle u, v \rangle / ||v||^2$ (where $v \neq 0$). Then

$$u = \frac{\langle u, v \rangle}{\|v\|^2} v + \left(u - \frac{\langle u, v \rangle}{\|v\|^2} v \right).$$

Cauchy-Schwarz Inequality. If $u, v \in V$, then

$$|\langle u, v \rangle| \le ||u|| ||v||.$$

This inequality is an equality iff one of u, v is a scalar multiple of the other.

Proof. Let $u, v \in V$, and assume $v \neq 0$. Consider $u = \frac{\langle u, v \rangle}{\|v\|^2} v + w$, where w is orthogonal to v. By the Pythagorean theorem, we have

$$||u||^2 = \left\| \frac{\langle u, v \rangle}{||v||^2} v \right\|^2 + ||w||^2 = \frac{|\langle u, v \rangle|^2}{||v||^2} + ||w||^2 \ge \frac{|\langle u, v \rangle|^2}{||v||^2}.$$

Multiply both sides, take a square root, and we are done. This is an equality iff w = 0, but this is true iff u is a multiple of v.

Triangle Inequality. If $u, v \in V$, then

$$||u + v|| \le ||u|| + ||v||.$$

This is an equality iff one of u, v is a nonnegative multiple of the other.

Proof. Let $u, v \in V$. Then

$$||u+v||^2 = ||u||^2 + ||v||^2 + \langle u,v \rangle + \overline{\langle u,v \rangle} = ||u||^2 + ||v||^2 + 2\operatorname{Re}\langle u,v \rangle \le ||u||^2 + ||v^2|| + 2||u||||v|| = (||u|| + ||v||)^2.$$

The inequality step frollows from Cauchy-Schwartz, where $2\operatorname{Re}\langle u,v\rangle \leq 2|\langle u,v\rangle|$. Taking square roots gives the triangle inequality. This is an equality iff the two inequalities above are equalities, which is true iff $\langle u,v\rangle = \|u\|\|v\|$.

Parallelogram Equality. *If* $u, v \in V$, *then*

$$||u + v||^2 + ||u - v||^2 = 2(||u||^2 + ||v||^2).$$

Proof. Exercise.

□

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2.3 Orthonormal Bases

A list (e_1, \dots, e_m) of vectors in V is orthonormal if $\langle e_j, e_k \rangle = 0$ when $j \neq k$ and equals 1 when j = k, for $j, k \in \{1, \dots, m\}$. Orthonormal lists are nice.

Proposition 2.1. If (e_1, \dots, e_m) is an orthonormal list of vectors in V, then

$$||a_1e_1 + \dots + a_me_m||^2 = |a_1|^2 + \dots + |a_m|^2$$

 \boxtimes

for all $a_1, \dots, a_m \in \mathbf{F}$.

Proof. Since each e_i has norm 1, this follows from repeated applications of the Pythagorean theorem.

Corollary 2.1. Every orthonormal list of vectors is linearly independent.

An **orthonormal basis** of V is an orthonormal list of vectors in V that forms a basis for V. The standard basis is a good example. If we find an orthonormal list of length dim V, then this is automatically an orthonormal basis of V (since they must be LI). In general, given a basis (e_1, \dots, e_n) of V and a vector $v \in V$, we know there is some choice of scalars a_1, \dots, a_m such that $v = a_1e_1 + \dots + a_ne_n$, but finding the a_j 's can be difficult. This is not the case for an orthonormal basis.

Theorem 2.1. Suppose (e_1, \dots, e_n) is an orthonormal basis of V. Then

$$v = \langle v, e_1 \rangle e_1 + \dots + \langle v, e_n \rangle e_n$$

and

$$||v||^2 = |\langle v, e_1 \rangle|^2 + \dots + |\langle v, e_n \rangle|^2$$

for every $v \in V$.

Proof. Let $v \in V$. Since (e_1, \dots, e_n) is a basis of V, there exist scalars a_1, \dots, a_n such that $v = a_1e_1 + \dots + a_ne_n$. Taking the inner product of both sides with e_j , we get $\langle v, e_j \rangle = a_j$. The second part follows from the first proposition and our previous result.