Dot Product, Orthogonality, and Projections

August 13 - AM

1. Dot Product (Scalar Product, Inner Product)

Definition 1. (Dot Product). The dot product $\mathbf{v} \cdot \mathbf{w}$ of two vectors

$$\mathbf{v} = \langle a_1, b_1, c_1 \rangle, \qquad \mathbf{w} = \langle a_2, b_2, c_2 \rangle$$

is the scalar defined by

 $\mathbf{v} \cdot \mathbf{w} = a_1 a_2 + b_1 b_2 + c_1 c_2$

denoted by (\mathbf{v}, \mathbf{w}) or $\langle \mathbf{v}, \mathbf{w} \rangle$.

In words, to compute the dot product, multiply the corresponding components and add.

Theorem 1. (Properties of the Dot Product).

- *i.* $\mathbf{0} \cdot \mathbf{v} = \mathbf{v} \cdot \mathbf{0} = 0$
- *ii.* Commutativity: $\mathbf{v} \cdot \mathbf{w} = \mathbf{w} \cdot \mathbf{v}$
- *iii.* Pulling out scalars: $(\lambda \mathbf{v}) \cdot \mathbf{w} = \mathbf{v} \cdot (\lambda \mathbf{w}) = \lambda(\mathbf{v} \cdot \mathbf{w})$
- *iv.* Distributive Law: $\mathbf{u} \cdot (\mathbf{v} + \mathbf{w}) = \mathbf{u} \cdot \mathbf{v} + \mathbf{u} \cdot \mathbf{w}$

$$(\mathbf{v} + \mathbf{w}) \cdot \mathbf{u} = \mathbf{v} \cdot \mathbf{u} + \mathbf{w} \cdot \mathbf{u}$$

v. Relation with length: $\mathbf{v} \cdot \mathbf{v} = \|\mathbf{v}\|^2$

There is one convention for the angle between two vectors:

The angle between two vectors is chosen to satisfy $0 \le \theta \le \pi$

Theorem 2. (Dot Product and the Angle). Let θ be the angle between two nonzero vectors \mathbf{v} and \mathbf{w} . Then

$$\mathbf{v} \cdot \mathbf{w} = \|\mathbf{v}\| \|\mathbf{w}\| \cos \theta$$
 or $\cos \theta = \frac{\mathbf{v} \cdot \mathbf{w}}{\|\mathbf{v}\| \|\mathbf{w}\|}$

Proof. According to the Law of Cosines, the three sides of a triangle satisfy

$$c^2 = a^2 + b^2 - 2ab\cos\theta$$

If two sides of the triangle are \mathbf{v} and \mathbf{w} , then the third side is $\mathbf{v} - \mathbf{w}$, and the Law of Cosines gives

$$\|\mathbf{v} - \mathbf{w}\|^2 = \|\mathbf{v}\|^2 + \|\mathbf{w}\|^2 - 2\cos\theta \|\mathbf{v}\| \|\mathbf{w}\|$$

By property (v) of Theorem 1 and the Distributive Law.

$$\|\mathbf{v} - \mathbf{w}\|^2 = (\mathbf{v} - \mathbf{w}) \cdot (\mathbf{v} - \mathbf{w}) = \mathbf{v} \cdot \mathbf{v} - 2\mathbf{v} \cdot \mathbf{w} + \mathbf{w} \cdot \mathbf{w}$$
$$= \|\mathbf{v}\|^2 + \|\mathbf{w}\|^2 - 2\mathbf{v} \cdot \mathbf{w}$$

Thus, Equation (2) follows.

By definition of the arccosine, the angle $\theta = \cos^{-1} x$ is the angle in the interval $[0, \pi]$ satisfying $\cos \theta = x$. Thus, for nonzero vectors **v** and **w**, we have

$$\theta = \cos^{-1}\left(\frac{\mathbf{v} \cdot \mathbf{w}}{\|\mathbf{v}\| \|\mathbf{w}\|}\right)$$

2. Perpendicular (Orthogonal), Obtuseness and Projection

Definition 2. (Perpendicular). Two nonzero vectors \mathbf{v} and \mathbf{w} are called perpendicular or orthogonal if the angle between them is $\frac{\pi}{2}$, denoted as $\mathbf{v} \perp \mathbf{w}$.

$$\mathbf{v} \perp \mathbf{w} \quad iff \quad \mathbf{v} \cdot \mathbf{w} = 0$$

The standard basis vectors are mutually orthogonal and have length 1.

Definition 3. (Obtuseness).

The angle θ between \mathbf{v} and \mathbf{w} is obtuse if $\mathbf{v} \cdot \mathbf{w} < 0$.

The angle θ between \mathbf{v} and \mathbf{w} is acute if $\mathbf{v} \cdot \mathbf{w} > 0$.

Definition 4. (Projection). Assume $\mathbf{v} \neq 0$. The projection of \mathbf{u} along \mathbf{v} is the vector

$$\mathbf{u}_{\parallel} = (\mathbf{u} \cdot \mathbf{e}_{\mathbf{v}})\mathbf{e}_{\mathbf{v}} \quad or \quad \mathbf{u}_{\parallel} = \left(\frac{\mathbf{u} \cdot \mathbf{v}}{\mathbf{v} \cdot \mathbf{v}}\right)\mathbf{v}$$

The scalar $\mathbf{u} \cdot \mathbf{e}_{\mathbf{v}}$ is called the **component** of \mathbf{u} along \mathbf{v} , $\mathbf{u} \cdot \mathbf{e}_{\mathbf{v}} = \|\mathbf{u}\| \cos \theta$.

Proof.

$$\begin{split} \mathbf{u}_{\parallel} &= (\mathbf{u} \cdot \mathbf{e}_{\mathbf{v}}) \mathbf{e}_{\mathbf{v}} = \left(\mathbf{u} \cdot \frac{\mathbf{v}}{\|\mathbf{v}\|}\right) \frac{\mathbf{v}}{\|\mathbf{v}\|} \\ &= \left(\frac{\mathbf{u} \cdot \mathbf{v}}{\|\mathbf{v}\|^2}\right) \mathbf{v} = \left(\frac{\mathbf{u} \cdot \mathbf{v}}{\mathbf{v} \cdot \mathbf{v}}\right) \mathbf{v} \end{split}$$

Every vector \mathbf{u} can be written as the sum of the projection \mathbf{u}_{\parallel} and a vector \mathbf{u}_{\perp} with respect to another vector \mathbf{v} .

$$\mathbf{u} = \mathbf{u}_{\parallel} + \mathbf{u}_{\perp}$$

This is the decomposition of \mathbf{u} with respect to \mathbf{v} .

THE SPECTRAL THEOREM

Let A be an $n \times n$ symmetric real matrix. An amazing fact is that we can write A as

with $\vec{v}_1, \vec{v}_2, \ldots, \vec{v}_n$ orthonormal vectors.

Here is how we compute¹ the λ 's and the \vec{v} 's. The fact that this method works involves what seems to be some amazing pieces of good fortune.

Step 1 Compute the characteristic polynomial det(A - kId).

Factor this polynomial as $(\lambda_1 - k)(\lambda_2 - k)\cdots(\lambda_n - k)$. The Fundamental Theorem of Algebra² promises us that such a factorization is possible if we use complex numbers. However, it turns out in our case that life is much better than this:

Lucky Fact 1: All the roots of f are real.

Step 2 For each eigenvalue λ , compute an **orthonormal** basis for Ker $(A - \lambda Id)$.

Putting all these bases together gives us a list of vectors: $\vec{v}_1, \vec{v}_2, \ldots$

Lucky Fact 2: The geometric multiplicity of λ , meaning the dimension of this kernel, is equal to the number of times λ occurs as a root of f.

Thus, the total number of vectors in our list is equal to the number of roots of f, which is n.

Lucky Fact 3: These vectors form an orthonormal basis of \mathbb{R}^n .

In the rest of this note, we will explain why we got so lucky.

A KEY FACT

We will use the following key fact twice below:

Key Fact: Suppose that \vec{u} and \vec{v} are eigenvectors of A, with

$$A\vec{u} = \lambda \vec{u} \quad A\vec{v} = \mu \vec{v}.$$

Suppose that $\lambda \neq \mu$. Then \vec{u} and \vec{v} are orthogonal.

To see this, we compute $\vec{u}^T A \vec{v}$ in two ways. We have

$$\vec{u}^T A \vec{v} = \vec{u}^T (\lambda \vec{v}) = \lambda \vec{u}^T \vec{v}.$$

But, also, $\vec{u}^T A \vec{v} = (\vec{v}^T A^T \vec{u})^T = (\vec{v}^T A \vec{u})^T$, where we have replaced A^T by A because A is symmetric. And $(\vec{v}^T A \vec{u})^T = (\vec{v}^T \mu \vec{u})^T = \mu (\vec{v}^T \vec{u})^T = \mu (\vec{u}^T \vec{v})$. Putting it all together,

$$\lambda \vec{u}^T \vec{v} = \mu \vec{u}^T \vec{v}.$$

Since $\lambda \neq \mu$, we get $\vec{u}^T \vec{v} = 0$. In other words, the dot product $\vec{u} \cdot \vec{v}$ is 0 or, in still other words, \vec{u} and \vec{v} are orthogonal.

We now start explaining the Lucky Facts.

LUCKY FACT 1

Suppose that f had a complex root, $\lambda = a + bi$. Let the corresponding eigenvector be $\vec{x} + i\vec{y}$, with x and y each real vectors. So

$$A(\vec{x} + i\vec{y}) = (a + bi)(\vec{x} + i\vec{y}).$$

Taking complex conjugates of everything, we also have

$$A(\vec{x} - i\vec{y}) = (a - bi)(\vec{x} - i\vec{y})$$

So $\vec{x} - i\vec{y}$ is also an eigenvector, with eigenvalue a - bi.

Now, if λ is complex, then $b \neq 0$, and $a + bi \neq a - bi$. We can now use the Key Fact, with $\vec{u} = \vec{x} + i\vec{y}$ and $\vec{v} = \vec{x} - i\vec{y}$. (You probably expected us to use the Key Fact for real vectors, but there is nothing about our argument that needed the entries to be real.)

So we deduce that $\vec{u} \cdot \vec{v} = 0$. Plugging in the formulas for \vec{u} and \vec{v} , that $(\vec{x} + i\vec{y}) \cdot (\vec{x} - i\vec{y}) = 0$.

But
$$(\vec{x}+i\vec{y})\cdot(\vec{x}-i\vec{y}) = \vec{x}\cdot\vec{x}-i\vec{x}\cdot\vec{y}+i\vec{y}\cdot\vec{x}+\vec{y}\cdot\vec{y} = \vec{x}\cdot\vec{x}+\vec{y}\cdot\vec{y} = |\vec{x}|^2+|\vec{y}|^2.$$

There is no way this is zero. (Remember that \vec{x} and \vec{y} have real entries.) More precisely, the only way it could be zero is if $\vec{x} = \vec{y} = 0$, but then $\vec{x} + i\vec{y}$ is 0 and is not an example of a nontrivial eigenvector.

We started out supposing that f has a complex root, and reached the absurd conclusion that $|\vec{x}|^2 + |\vec{y}|^2 = 0$. The resolution is that, in fact, f doesn't have any complex roots, and all its roots are real. (Here we are using the Fundamental Theorem of Algebra to know that there are n roots total – real and complex.) Lucky Fact 1 explained!

LUCKY FACT 2

Let λ be a root of f(k), so det $(A - \lambda \text{Id}) = 0$. Let r be the geometric multiplicity of λ , meaning that there is an r dimensional space of solutions to the equation $A\vec{v} = \lambda \vec{v}$.

Let $\vec{v}_1, \vec{v}_2, \ldots, \vec{v}_r$ be an orthonormal basis for the space of solutions to $A\vec{v} = \lambda \vec{v}$. Find additional vectors $\vec{v}_{r+1}, \ldots, \vec{v}_n$ so that $\vec{v}_1, \vec{v}_2, \ldots, \vec{v}_r, \vec{v}_{r+1}, \ldots, \vec{v}_n$ is an orthonormal basis for \mathbb{R}^n . For $r+1 \leq i \leq n$, we can write

$$A\vec{v}_i = \sum_{j=1}^n c_{ij}\vec{v}_j$$

for some coefficients c_{ij} . We can organize these equations into a matrix. For concreteness, we take r = 2 and n = 5.

$$\begin{pmatrix} | & | & | & | & | \\ \vec{v}_1 & \vec{v}_2 & \vec{v}_3 & \vec{v}_4 & \vec{v}_5 \\ | & | & | & | & | \end{pmatrix} \begin{pmatrix} \lambda & & & & \\ \lambda & & & & \\ c_{31} & c_{32} & c_{33} & c_{34} & c_{35} \\ c_{41} & c_{42} & c_{43} & c_{44} & c_{45} \\ c_{51} & c_{52} & c_{53} & c_{54} & c_{55} \end{pmatrix} = A \cdot \begin{pmatrix} | & | & | & | & | \\ \vec{v}_1 & \vec{v}_2 & \vec{v}_3 & \vec{v}_4 & \vec{v}_5 \\ | & | & | & | & | \end{pmatrix}$$

You may recognize this argument from the November 13 notes.

Let
$$S = \begin{pmatrix} | & | & | & | & | \\ \vec{v}_1 & \vec{v}_2 & \vec{v}_3 & \vec{v}_4 & \vec{v}_5 \\ | & | & | & | & | \end{pmatrix}$$
. So $SAS^{-1} = \begin{pmatrix} \lambda & & & \\ \lambda & & & \\ c_{31} & c_{32} & c_{33} & c_{34} & c_{35} \\ c_{41} & c_{42} & c_{43} & c_{44} & c_{45} \\ c_{51} & c_{52} & c_{53} & c_{54} & c_{55} \end{pmatrix}$

But $\vec{v}_1, \vec{v}_2, \ldots, \vec{v}_n$ is an orthonormal basis, so S is orthogonal and $S^{-1} = S^T$. So we have

$$S^{T}AS = \begin{pmatrix} \lambda & & & \\ & \lambda & & \\ c_{31} & c_{32} & c_{33} & c_{34} & c_{35} \\ c_{41} & c_{42} & c_{43} & c_{44} & c_{45} \\ c_{51} & c_{52} & c_{53} & c_{54} & c_{55} \end{pmatrix}.$$

Notice that $(S^T A S)^T = S^T A^T S = S^T A S$, so the above matrix is symmetric. In particular, that lower left block is entirely zero and we have

$$S^{-1}AS = \begin{pmatrix} \lambda & & \\ & \lambda & \\ \hline & & C \end{pmatrix}$$

for some symmetric $(n-r) \times (n-r)$ matrix C. As in the November 13 notes, we see that

$$\det(A - k \cdot \operatorname{Id}_n) = \det\left(\left(\begin{array}{c|c} \lambda & & \\ & \lambda & \\ \hline & & \\ \end{array}\right) - k \cdot \operatorname{Id}_n\right) = \det\left(\begin{array}{c|c} \lambda - k & & \\ \hline & \lambda - k & \\ \hline & & \\ \end{array}\right) = \left(\lambda - k\right)^r \det(C - k \cdot \operatorname{Id}_{n-r}).$$

Our goal is to show that $\lambda - k$ divides det $(A - k \cdot \mathrm{Id}_n)$ exactly r times. So we want to know that $\lambda - k$ does **not** divide det $(C - k\mathrm{Id}_{n-r})$. In other words, we want to know that λ is **not** an eigenvalue of C.

Suppose, to the contrary that $C\vec{u} = \lambda \vec{u}$. Then

$$\begin{pmatrix} \lambda & & \\ & \lambda & \\ \hline & & C \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ \vec{u} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \lambda \vec{u} \end{pmatrix}.$$

So $\begin{pmatrix} 0\\ 0\\ \vec{u} \end{pmatrix}$ is another λ eigenvector for A, which should have been listed when we found the v_i .

LUCKY FACT 3

We need to explain two things: Why the \vec{v}_i are orthonormal, and why they are a basis.

If \vec{v}_i and \vec{v}_j both come from the same eigenvalue λ , then $\vec{v}_i \cdot \vec{v}_j = 0$ because we chose an orthonormal basis for the λ -eigenspace. Also, the \vec{v}_i all have length 1 because we chose an orthonormal basis in this place.

If $\vec{v_i}$ and $\vec{v_j}$ come from different eigenvalues, then the Key Fact tells us that $\vec{v_i}$ and $\vec{v_j}$ are perpendicular. So we also have perpendicularity between these vectors.

Since the \vec{v}_i are orthonormal, they are linearly independent. From Lucky Fact 2, the number of \vec{v}_i coming from each λ is the same as the number of times $(\lambda - k)$ divides f(k), and the polynomial f has degree n. So the total number of vectors \vec{v}_i is n. We have n linearly independent vectors in \mathbb{R}^n , so we have a basis for \mathbb{R}^n .

We have now explained all of our Lucky Facts, and shown that every symmetric matrix has a basis of eigenvectors.