

5.6 Gram-Schmidt

Let $(V, \langle, \rangle)$ be an n -dim inner product space, and suppose $\{x_1, \dots, x_n\}$ is any basis of the vector space V .

We want to use $\{x_i\}$ to construct a new basis of V that respects the inner product $\langle, \rangle$ in the sense that it is orthonormal.

This new basis $\{u_1, \dots, u_n\}$ is obtained by the following Gram-Schmidt Orthogonalization Process.

Idea: we want all of the u_i to be unit vectors, and we also want them to be orthogonal to each other:

1. $\|u_i\| = 1$ for all i
2. $\langle u_i, u_j \rangle = \delta_{ij} = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}$

property 1. is actually encoded in property 2.
if you recall what the definition of $\|u\|$ is.

Here we go:

Step 1: Put $u_1 = \frac{x_1}{\|x_1\|}$

Step 2: Put $p_1 = \text{proj}_{u_1} x_2 = \langle x_2, u_1 \rangle u_1$

Then $\text{orth}_{u_1} x_2 = (x_2 - p_1)$ is orthogonal to x_1 and u_1 .

Further,

$$\underline{x}_2 - p_1 \neq 0 \quad \text{since}$$

$$\underline{x}_2 - p_1 = -\frac{\langle \underline{x}_2, \underline{u}_1 \rangle}{\|\underline{x}_1\|} \underline{x}_1 + \underline{x}_2. \quad (*)$$

and $\underline{x}_2 \neq 0$ since it is a basis vector, and $\underline{x}_1, \underline{x}_2$ are linearly independent.

Step 3. Put $\underline{u}_2 = \frac{(\underline{x}_2 - p_1)}{\|\underline{x}_2 - p_1\|}$

Then $\underline{u}_2$ is a unit vector orthogonal to $\underline{u}_1$. Notice that $\text{span}\{\underline{x}_1, \underline{x}_2\} = \text{span}\{\underline{u}_1, \underline{u}_2\}$ by eqn (*). Thus we have produced an orthonormal basis of $\text{span}\{\underline{x}_1, \underline{x}_2\}$, constructed from $\{\underline{x}_1, \underline{x}_2\}$.

Step 4. Repeat this process to produce the rest of the basis.

Define $\underline{u}_2, \dots, \underline{u}_n$ recursively by

$$\underline{u}_{k+1} = \frac{(\underline{x}_{k+1} - p_k)}{\|\underline{x}_{k+1} - p_k\|} \quad \text{where}$$

$$p_k = \langle \underline{x}_{k+1}, \underline{u}_1 \rangle \underline{u}_1 + \langle \underline{x}_{k+1}, \underline{u}_2 \rangle \underline{u}_2 + \dots + \langle \underline{x}_{k+1}, \underline{u}_k \rangle \underline{u}_k$$

the projection of $\underline{x}_{k+1}$ onto the subspace $\text{span}\{\underline{u}_1, \dots, \underline{u}_k\} = \text{span}\{\underline{x}_1, \dots, \underline{x}_k\}$.

Ex. Find an orthonormal basis for P_3 w/ inner product defined by

$$\langle p, q \rangle = \sum_{i=1}^3 p(x_i) q(x_i)$$

where $x_1 = -1$, $x_2 = 0$, $x_3 = 1$.

Start w/ the standard basis $\{1, x, x^2\}$ and use G-S to produce the orthonormal basis $\{u_1, u_2, u_3\}$

$$u_1 = \frac{1}{\sqrt{3}}$$

$$u_2 = \frac{1}{\sqrt{2}} x$$

$$u_3 = \frac{\sqrt{6}}{2} \left(x^2 - \frac{2}{3} \right)$$

Ex. Let $A = \begin{pmatrix} 1 & -1 & 4 \\ 1 & 4 & -2 \\ 1 & 4 & 2 \\ 1 & -1 & 0 \end{pmatrix}$ Find an orthonormal basis for the column space.

~~But~~ Let $r_{11} = \|a_1\| = 2$

$$\text{put } \underline{q}_1 = \frac{a_1}{r_{11}} = \left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2} \right)^T$$

$$\text{let } r_{12} = \langle \underline{q}_1, a_2 \rangle = \underline{q}_1^T a_2 = 3$$

$$\underline{p}_1 = r_{12} \underline{q}_1 = 3 \underline{q}_1$$

$$\underline{a}_2 - \underline{p}_1 = \left(-\frac{5}{2}, \frac{5}{2}, \frac{5}{2}, -\frac{5}{2} \right)$$

$$r_{22} = \|\underline{a}_2 - \underline{p}_1\| = 5$$

$$\underline{q}_2 = \frac{\underline{a}_2 - \underline{p}_1}{r_{22}} = \left(-\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, -\frac{1}{2} \right)^T$$

Similarly

$$r_{13} = \langle \underline{g}_1, \underline{g}_3 \rangle = 2$$

$$r_{23} = \langle \underline{g}_2, \underline{g}_3 \rangle = -2$$

$$\underline{p}_2 = r_{13} \underline{g}_1 + r_{23} \underline{g}_2 = (2, 9, 0, 2)^T$$

$$\underline{q}_3 = \underline{g}_3 - \underline{p}_2 = (2, -2, 2, -2)^T$$

$$r_{33} = \|\underline{q}_3\| = 4$$

$$\underline{g}_3 = \frac{\underline{q}_3 - \underline{p}_2}{r_{33}} = \left(\frac{1}{2}, -\frac{1}{2}, \frac{1}{2}, -\frac{1}{2}\right)^T$$

so $Q = \begin{pmatrix} \frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} & -\frac{1}{2} \end{pmatrix}$ has the same column space as A , but is an orthogonal matrix.

If we collect all of the r_{ij} 's into a matrix, we get

$$R = \begin{pmatrix} r_{11} & r_{12} & r_{13} \\ 0 & r_{22} & r_{23} \\ 0 & 0 & r_{33} \end{pmatrix} = \begin{pmatrix} 2 & 3 & 2 \\ 0 & 5 & -2 \\ 0 & 0 & 4 \end{pmatrix}$$

RE. Verify that $A = QR$

This is called the Q-R factorization of A .

In general, we can always decompose an $m \times n$ matrix A w/ rank n into such a product where $Q \in \mathbb{R}^{m \times n}$ is orthogonal (its columns are an orthonormal basis of the column space) and $R \in \mathbb{R}^{n \times n}$ is non-singular, and whose diagonal entries are all positive.

Think of R as an orientation-preserving change of basis matrix. ~~orthogonal~~

Least Squares Problem

Let $A \in \mathbb{R}^{m \times n}$ w/ $\text{rank}(A) = n$, and let $b \in \mathbb{R}^m$.

The least squares problem is to find a vector $\hat{x} \in \text{Row}(A)$ that is "as close as possible" to b .
such $A\hat{x}$ (whoops!!)

To find $\hat{x}$, one solves the normal equation(s)

$$A^T A x = A^T b$$

This has a unique sol'n since $A^T \in \mathbb{R}^{n \times m}$ w/
 $\text{rank}(A^T) = \text{rank}(A) = n$; i.e., the column space (image)
of A^T is all of $\mathbb{R}^n$.

When the columns of A form an orthonormal
set of vectors in $\mathbb{R}^m$, then $A^T A = I \in \mathbb{R}^{n \times n}$.
Then the solution of the least squares problem
is

$$\hat{x} = A^T b.$$

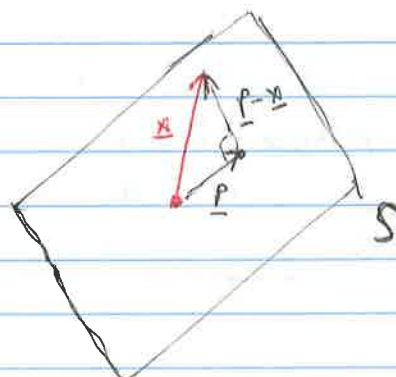
Thm. Let S be a subspace of an inner product space
 $(V, \langle \cdot, \cdot \rangle)$ and let $x \in V$. Let $\{u_1, u_2, \dots, u_n\}$ be an
orthonormal basis for S .

$$\text{If } p = \sum_{i=1}^n c_i u_i$$

where $c_i = \langle u_i, x \rangle \quad \forall i$,

then $p - x \in S^\perp$.

The picture looks as follows.



We can then decompose $\underline{x}$ into a sum of a vector in S and one in $S^\perp$.

Namely,

$$\underline{x} = \underline{p} + (\underline{p} - \underline{x}) = [\underline{x}]_S + [\underline{x}]_{S^\perp}$$

These are components of $\underline{x}$ that respect the subspace S .

(orthogonal)

we call $[\underline{x}]_S = \underline{p}$ the projection of $\underline{x}$ onto S .

Thm. Let S be a nonzero subspace of $\mathbb{R}^m$ and let $\underline{b} \in \mathbb{R}^m$. If $\{\underline{u}_1, \dots, \underline{u}_k\}$ is an orthonormal basis for S and $U = (\underline{u}_1, \underline{u}_2, \dots, \underline{u}_k)$, ~~then~~ then the projection $\underline{p}$ of $\underline{b}$ onto S is given by

$$\underline{p} = UU^T \underline{b}$$

Ex. Let S be the set of all vectors in $\mathbb{R}^3$ of the form $(x, y, 0)^T$. Find the vector $\underline{p} \in S$ closest to $\underline{w} = (5, 3, 4)^T$.

Sol'n. $\underline{e}_1$ and $\underline{e}_2$ form an orthonormal basis for S .

$$\text{Then } U = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \quad \text{and} \quad U^T = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$$

$$UU^T = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\text{and } \underline{p} = UU^T \underline{w} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 5 \\ 3 \\ 4 \end{pmatrix} = \begin{pmatrix} 5 \\ 3 \\ 0 \end{pmatrix}$$

This is exactly what one would expect!

Ex. Find the best linear approximation of e^x on the interval $[0, 1]$ by a linear function.

$$\text{Let } S = P_1 = \{p \in C[0, 1] \mid p(x) = ax + b.\}$$

Although $\{1, x\}$ span P_1 , they are not orthogonal wrt the inner product

$$\langle p, q \rangle = \int_0^1 p(x) q(x) dx$$

By Gram-Schmidt, $\{1, \sqrt{2}(x - \frac{1}{2})\}$ form an orthonormal basis.

$$\text{let } c_1 = \langle u_1, e^x \rangle = \int_0^1 e^x dx = e - 1$$

$$c_2 = \langle u_2, e^x \rangle = \int_0^1 \sqrt{2} \left(x - \frac{1}{2}\right) e^x dx = \sqrt{3} (3 - e)$$

$$\text{Then } p(x) = c_1 u_1(x) + c_2 u_2(x)$$

$$= (e-1) \cdot 1 + \sqrt{3} (3-e) \left[\sqrt{2} \left(x - \frac{1}{2}\right) \right]$$

$$= (4e - 10) + 6(3-e)x \approx e^x$$

This is the best linear approximation of e^x on this interval!

pretty cool, right?!

Ex. Show that the functions $\sin x, \sin 2x, \dots, \sin nx$ are orthonormal wrt the inner product

$$\langle f, g \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} f(x) g(x) dx$$

in $[-\pi, \pi]$.

let $f \in C[-\pi, \pi]$ be 2π -periodic. Then we can approximate ~~represent~~ f by a Fourier series or trigonometric polynomial.

$$f(x) \approx \sum_{k=1}^n a_k(x) \sin(kx)$$

$$\text{where } a_k(x) = \frac{1}{\pi} \int_{-\pi}^{\pi} f(x) \sin(kx) dx = \langle f, \sin(kx) \rangle.$$

This can also be done w/ cosine. See pp. 252-3.