

Anton 6.3 Gram-Schmidt process and $A = QR$ matrix factorization

Let V be a vector space with inner product $\langle \cdot, \cdot \rangle$

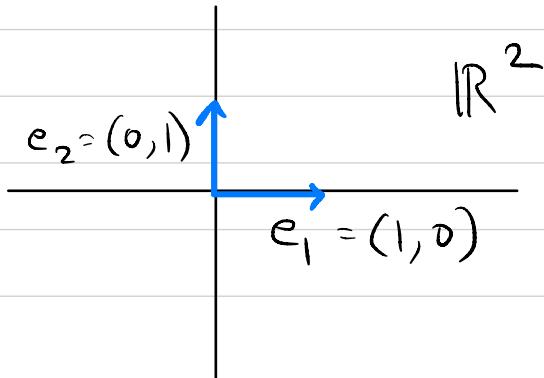
(e.g. $V = \mathbb{R}^n$ with dot product also called Euclidean product)

Definition: A set $S = \{q_1, q_2, \dots, q_n\}$ of vectors is an orthogonal set (or just orthogonal) if $\langle q_i, q_j \rangle = 0$ for all $i \neq j$

orthonormal set in addition to $\langle q_i, q_j \rangle = 0$ for $i \neq j$, each q_i is a unit vector i.e. $\langle q_i, q_i \rangle = 1$ for all i .

(Once you have an orthogonal set, it is easy to create an orthonormal set - just divide each vector by its norm to create a unit vector).

Standard basis $\{e_1, e_2, \dots, e_n\}$ of \mathbb{R}^n is an orthonormal basis for $\langle \cdot, \cdot \rangle = \text{dot product}$



$$\langle e_1, e_2 \rangle = 1 \cdot 0 + 0 \cdot 1 = 0$$

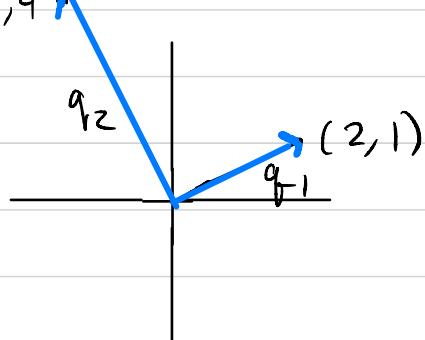
$$\langle e_1, e_1 \rangle = 1^2 + 0^2 = 1$$

$$\langle e_2, e_2 \rangle = 0^2 + 1^2 = 1$$

Example

\mathbb{R}^2 \langle , \rangle = dot product

$$q_1 = (2, 1) \quad q_2 = (-2, 4)$$



$\{q_1, q_2\}$ are orthogonal but not orthonormal

$$\langle q_1, q_2 \rangle = 2(-2) + 1 \cdot 4 = 0$$

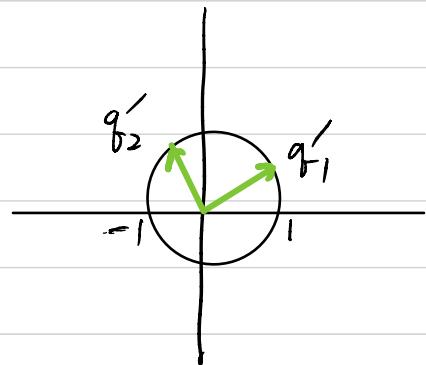
$$\langle q_1, q_1 \rangle = 2^2 + 1^2 = 5 \Rightarrow \|q_1\| = \sqrt{5}$$

$$\langle q_2, q_2 \rangle = (-2)^2 + 4^2 = 20 \Rightarrow \|q_2\| = \sqrt{20}$$

Let's normalize q_1 & q_2 to get an orthonormal set
 $\{q'_1, q'_2\}$

$$q'_1 = \frac{1}{\sqrt{5}} q_1 = \left(\frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}} \right)$$

$$q'_2 = \frac{1}{\sqrt{20}} (-2, 4) = \left(-\frac{2}{\sqrt{20}}, \frac{4}{\sqrt{20}} \right)$$



$\{q'_1, q'_2\}$ are orthonormal

$$\|q'_1\|^2 = \left(\frac{2}{\sqrt{5}} \right)^2 + \left(\frac{1}{\sqrt{5}} \right)^2 = \frac{2^2}{5} + \frac{1^2}{5} = \frac{5}{5} = 1.$$

$$\text{Similarly } \|q'_2\|^2 = 1$$

Why is an orthogonal / orthonormal basis better?

It is easy to find coordinates of a vector v with respect to an orthonormal basis $\{q_1, q_2, \dots, q_n\}$, by using the inner product (instead of solving system of linear equations)

Fact: If $\{q_1, \dots, q_n\}$ is an orthogonal set and $v \in \text{Span}\{q_1, \dots, q_n\}$

then $v = \frac{\langle v, q_1 \rangle}{\|q_1\|^2} q_1 + \frac{\langle v, q_2 \rangle}{\|q_2\|^2} q_2 + \dots + \frac{\langle v, q_n \rangle}{\|q_n\|^2} q_n$

If $\{q_1, \dots, q_n\}$ is an orthonormal set and $v \in \text{Span}\{q_1, \dots, q_n\}$

then $v = \langle v, q_1 \rangle q_1 + \langle v, q_2 \rangle q_2 + \dots + \langle v, q_n \rangle q_n$

Proof: $v = c_1 q_1 + c_2 q_2 + \dots + c_n q_n$ for some numbers c_1, c_2, \dots, c_n

We have to find what c_1, c_2, \dots, c_n are

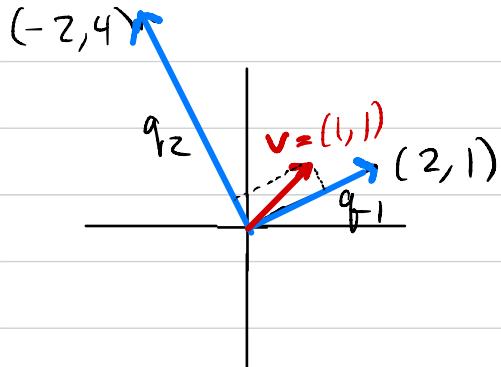
$$\begin{aligned}\langle v, q_1 \rangle &= \langle c_1 q_1 + c_2 q_2 + \dots + c_n q_n, q_1 \rangle \\ &= \underbrace{\langle c_1 q_1, q_1 \rangle}_{\|q_1\|^2} + \underbrace{\langle c_2 q_2, q_1 \rangle}_{0} + \dots + \underbrace{\langle c_n q_n, q_1 \rangle}_{0} \quad (\text{by orthogonal}) \\ &= c_1 \underbrace{\langle q_1, q_1 \rangle}_{\|q_1\|^2} + c_2 \underbrace{\langle q_2, q_1 \rangle}_{0} + \dots + c_n \underbrace{\langle q_n, q_1 \rangle}_{0}\end{aligned}$$

$$\langle v, q_1 \rangle = c_1 \|q_1\|^2 \quad \text{solve for } c_1$$

$\frac{\langle v, q_1 \rangle}{\|q_1\|^2} = c_1$, is the first coordinate of v in basis $\{q_1, q_2, \dots, q_n\}$.

similarly $c_2 = \frac{\langle v, q_2 \rangle}{\|q_2\|^2}$, etc. \square

Example. Express $v = (1, 1)$ as a linear combination of q_1 & q_2 from before



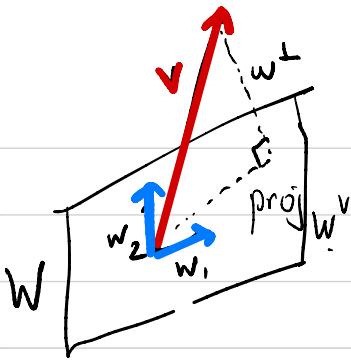
Solution: $\{q_1, q_2\}$ is orthogonal (but not orthonormal) so we use formula

$$v = \frac{\langle v, q_1 \rangle}{\|q_1\|^2} q_1 + \frac{\langle v, q_2 \rangle}{\|q_2\|^2} q_2$$

$$v = \frac{(1, 1) \cdot (2, 1)}{2^2 + 1^2} q_1 + \frac{(1, 1) \cdot (-2, 4)}{(-2)^2 + 4^2} q_2$$

$$= \frac{1 \cdot 2 + 1 \cdot 1}{5} q_1 + \frac{1 \cdot (-2) + 1 \cdot 4}{20} q_2$$

$$v = \boxed{\frac{3}{5} q_1 + \frac{1}{10} q_2}$$



Orthogonal
Projection of a vector \mathbf{v} onto a subspace
 $W = \text{Span} \{w_1, w_2, \dots, w_r\}$ where

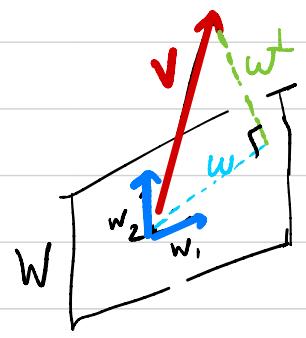
w_1, \dots, w_r is an orthogonal set of vectors,

Let $W^\perp = \{v \in V \mid \langle v, w \rangle = 0 \text{ for every } w \in W\}$ "orthogonal complement."

For $\mathbf{v} \in V$ we want to express $\mathbf{v} = \underbrace{\mathbf{w}}_{\text{proj}_W \mathbf{v}} + \underbrace{\mathbf{w}^\perp}_{\text{in } W^\perp}$

in W^\perp

in W



The formulas are:

$$\text{proj}_W \mathbf{v} = \mathbf{w} = \frac{\langle \mathbf{v}, \mathbf{w}_1 \rangle}{\|\mathbf{w}_1\|^2} \mathbf{w}_1 + \frac{\langle \mathbf{v}, \mathbf{w}_2 \rangle}{\|\mathbf{w}_2\|^2} \mathbf{w}_2 + \dots + \frac{\langle \mathbf{v}, \mathbf{w}_r \rangle}{\|\mathbf{w}_r\|^2} \mathbf{w}_r$$

$$w^\perp = \mathbf{v} - \frac{\langle \mathbf{v}, \mathbf{w}_1 \rangle}{\|\mathbf{w}_1\|^2} \mathbf{w}_1 - \frac{\langle \mathbf{v}, \mathbf{w}_2 \rangle}{\|\mathbf{w}_2\|^2} \mathbf{w}_2 - \dots - \frac{\langle \mathbf{v}, \mathbf{w}_r \rangle}{\|\mathbf{w}_r\|^2} \mathbf{w}_r$$

If w_1, w_2, \dots, w_r are orthonormal, the formulas become

$$\text{proj}_W \mathbf{v} = \mathbf{w} = \langle \mathbf{v}, \mathbf{w}_1 \rangle \mathbf{w}_1 + \langle \mathbf{v}, \mathbf{w}_2 \rangle \mathbf{w}_2 + \dots + \langle \mathbf{v}, \mathbf{w}_r \rangle \mathbf{w}_r$$

$$w^\perp = \mathbf{v} - \langle \mathbf{v}, \mathbf{w}_1 \rangle \mathbf{w}_1 - \langle \mathbf{v}, \mathbf{w}_2 \rangle \mathbf{w}_2 - \dots - \langle \mathbf{v}, \mathbf{w}_r \rangle \mathbf{w}_r$$

Why?

$$v = w + w^\perp$$

$$v = \underbrace{(c_1 w_1 + c_2 w_2 + \dots + c_r w_r)}_{w \in W} + w^\perp \quad \text{for some } c_1, c_2, \dots, c_r \in \mathbb{R}$$

$w \in W$ is a linear combination of w_1, \dots, w_r .

$$\langle v, w_1 \rangle = \langle c_1 w_1 + \dots + c_r w_r + w^\perp, w_1 \rangle$$

$$= \underbrace{\langle c_1 w_1, w_1 \rangle}_{=0} + \underbrace{\langle c_2 w_2, w_1 \rangle}_{=0} + \dots + \underbrace{\langle c_r w_r, w_1 \rangle}_{=0} + \underbrace{\langle w^\perp, w_1 \rangle}_{=0}$$

$$\langle v, w_1 \rangle = c_1 \langle w_1, w_1 \rangle$$

$$\frac{\langle v, w_1 \rangle}{\|w_1\|^2} = c_1 \quad (\|w_1\|^2 = \langle w_1, w_1 \rangle)$$

$$\text{Similarly } c_2 = \frac{\langle v, w_2 \rangle}{\|w_2\|^2}, \dots$$

$$\text{So } v = \underbrace{c_1 w_1 + \dots + c_r w_r}_{w} + w^\perp$$

$$v = \frac{\langle v, w_1 \rangle}{\|w_1\|^2} w_1 + \dots + \frac{\langle v, w_r \rangle}{\|w_r\|^2} w_r + w^\perp$$

$$\text{So } w^\perp = v - \frac{\langle v, w_1 \rangle}{\|w_1\|^2} w_1 - \frac{\langle v, w_2 \rangle}{\|w_2\|^2} w_2 - \dots - \frac{\langle v, w_r \rangle}{\|w_r\|^2} w_r$$

$$w = \frac{\langle v, w_1 \rangle}{\|w_1\|^2} w_1 + \frac{\langle v, w_2 \rangle}{\|w_2\|^2} w_2 + \dots + \frac{\langle v, w_r \rangle}{\|w_r\|^2} w_r$$

Anton 6.3 #23

Find the orthogonal projection of $b = (1, 2, 0, -2)$ on the subspace spanned by $v_1 = (1, 1, 1, 1)$ & $v_2 = (1, 1, -1, -1)$

Solution $\text{proj}_V b = \frac{\langle b, v_1 \rangle}{\|v_1\|^2} v_1 + \frac{\langle b, v_2 \rangle}{\|v_2\|^2} v_2$
($V = \text{span}\{v_1, v_2\}$)

$$v_1 = (1, 1, 1, 1) \quad \|v_1\|^2 = 1^2 + 1^2 + 1^2 + 1^2 = 4 \quad v_2 = (1, 1, -1, -1) \quad \|v_2\|^2 = 1^2 + 1^2 + (-1)^2 + (-1)^2 = 4$$
$$b = (1, 2, 0, -2) \quad b = (1, 2, 0, -2)$$

$$\begin{aligned} \langle b, v_1 \rangle &= 1 \cdot 1 + 2 \cdot 1 + 0 \cdot 1 + (-2) \cdot 1 \\ &= 1 + 2 + 0 - 2 \\ &= 1 \\ \langle b, v_2 \rangle &= 1 \cdot 1 + 2 \cdot 1 + 0 \cdot (-1) + (-2) \cdot (-1) \\ &= 1 + 2 + 0 + 2 \\ &= 5 \end{aligned}$$

$$\text{proj}_V b = \frac{\langle b, v_1 \rangle}{\|v_1\|^2} v_1 + \frac{\langle b, v_2 \rangle}{\|v_2\|^2} v_2$$

$$= \frac{1}{4} v_1 + \frac{5}{4} v_2$$

$$= \left(\frac{1}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4} \right) + \left(\frac{5}{4}, \frac{5}{4}, -\frac{5}{4}, -\frac{5}{4} \right)$$

$$\text{proj}_V b = \boxed{\left(\frac{3}{2}, \frac{3}{2}, -1, -1 \right)}$$

Now Gram-Schmidt is easy. Here is what Gram-Schmidt does:

Given a basis $\{v_1, v_2, \dots, v_n\}$ of V find an orthogonal/orthonormal basis $\{q_1, q_2, \dots, q_n\}$ of V .

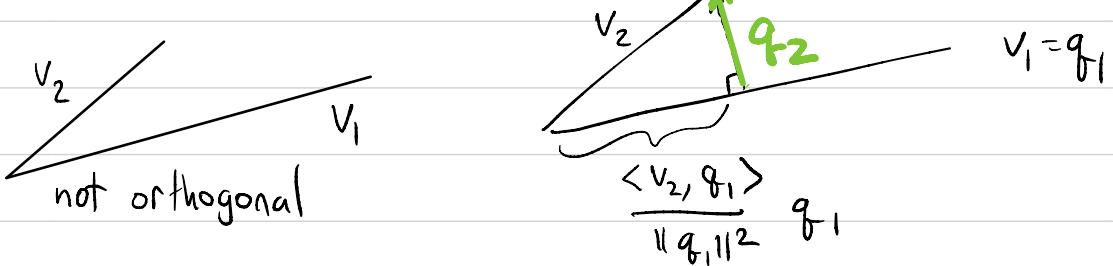
Key formula

$$w^\perp = v - \frac{\langle v, w_1 \rangle}{\|w_1\|^2} w_1 - \frac{\langle v, w_2 \rangle}{\|w_2\|^2} w_2 - \dots - \frac{\langle v, w_r \rangle}{\|w_r\|^2} w_r$$

1. Set $q_1 = v_1$ (or any scalar multiple)

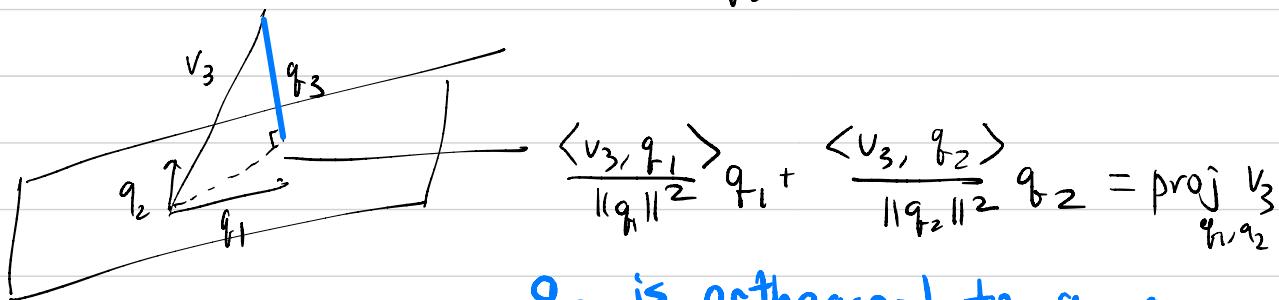
2. Set $q_2 = v_2 - \frac{\langle v_2, q_1 \rangle}{\|q_1\|^2} q_1$

q_2 is orthogonal to q_1



Can replace q_2 by a scalar multiple of it (e.g. to clear denominator)

3. set $q_3 = v_3 - \frac{\langle v_3, q_1 \rangle}{\|q_1\|^2} q_1 - \frac{\langle v_3, q_2 \rangle}{\|q_2\|^2} q_2$



4. And so on.

$q_i = v_i - \frac{\langle v_i, q_1 \rangle}{\|q_1\|^2} q_1 - \frac{\langle v_i, q_2 \rangle}{\|q_2\|^2} q_2 - \dots - \frac{\langle v_i, q_{i-1} \rangle}{\|q_{i-1}\|^2} q_{i-1}$

q_i is orthogonal to q_1, q_2, \dots, q_{i-1} .

(can replace q_i by any scalar multiple of it)

5. Get orthogonal basis $\{q_1, q_2, \dots, q_n\}$

To get orthonormal basis (each vector a unit vector)

Anton 6.3

31. Let \mathbb{R}^4 have the Euclidean inner product. Use the Gram–Schmidt process to transform the basis $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3, \mathbf{u}_4\}$ into an orthonormal basis.

$$\begin{aligned}\mathbf{u}_1 &= (0, 2, 1, 0), & \mathbf{u}_2 &= (1, -1, 0, 0), \\ \mathbf{u}_3 &= (1, 2, 0, -1), & \mathbf{u}_4 &= (1, 0, 0, 1)\end{aligned}$$

Solution. Set $q_1 = (0, 2, 1, 0) (= \mathbf{u}_1)$ $\|q_1\|^2 = 0^2 + 2^2 + 1^2 + 0^2 = 5$

Initially set $q_{f2} = \mathbf{u}_2 - \frac{\langle \mathbf{u}_2, q_1 \rangle}{\|q_1\|^2} q_1$ $\mathbf{u}_2 = (1, -1, 0, 0)$
 $= (1, -1, 0, 0) - \frac{-2}{5} q_1$ $\langle \mathbf{u}_2, q_1 \rangle = 0 - 2 - 0 - 0 = -2$

Scale through by 5 to clear denominator

$$\begin{aligned}5q_2 &= (5, -5, 0, 0) + 2(q_1) \\ &= (5, -5, 0, 0) + 2(0, 2, 1, 0)\end{aligned}$$

rename $5q_2$ $q_2 = (5, -5, 0, 0) + (0, 4, 2, 0)$
as q_{f2}

$$q_2 = (5, -1, 2, 0) \quad \|q_2\|^2 = 5^2 + 1^2 + 2^2 = 30$$

$$q_3 = \mathbf{u}_3 - \frac{\langle \mathbf{u}_3, q_1 \rangle}{\|q_1\|^2} q_1 - \frac{\langle \mathbf{u}_3, q_2 \rangle}{\|q_2\|^2} q_2 \quad \begin{aligned}u_3 &= (1, 2, 0, -1) \\ q_1 &= (0, 2, 1, 0)\end{aligned}$$

$$= (1, 2, 0, -1) - \frac{4}{5} q_1 - \frac{3}{30} q_2 \quad \langle \mathbf{u}_3, q_1 \rangle = 0 \quad 4 \quad 0 \quad 0 = 4$$

$$= (1, 2, 0, -1) - \frac{4}{5} q_1 - \frac{1}{10} q_2 \quad u_3 = (1, 2, 0, -1) \quad q_2 = (5, -1, 2, 0) \\ = 5 - 2 \quad 0 \quad 0$$

Rescale by 10

$$= (10, 20, 0, -10) - 8q_1 - q_2$$

$$\langle \mathbf{u}_3, q_2 \rangle = 3 \quad (\text{sum})$$

Rescale by 10

$$= (10, 20, 0, -10) - 8q_1 - q_2$$

$$= (10, 20, 0, -10) - 8(0, 2, 1, 0) - (5, -1, 2, 0)$$

$$= (5, 5, -10, -10)$$

Rescale by $\frac{1}{5}$

$$q_3 = (1, 1, -2, -2)$$

$$\|q_3\|^2 = 1^2 + 1^2 + (-2)^2 + (-2)^2 = 10$$

$$q_4 = u_4 - \frac{0}{5}q_1 - \frac{5}{30}q_2 - \frac{-1}{10}q_3$$

$$= (1, 0, 0, 1) - \frac{5}{30}(5, -1, 2, 0) + \frac{1}{10}(1, 1, -2, -2)$$

Rescale by 30

$$= (30, 0, 0, 30) - 5(5, -1, 2, 0) + 3(1, 1, -2, -2)$$

$$= (8, 8, -16, 24)$$

Rescale by 8

$$q_4 = (1, 1, -2, 3)$$

$$\|q_4\|^2 = 1^2 + 1^2 + (-2)^2 + 3^2 = 15$$

So an orthogonal basis of \mathbb{R}^4 that Gram-schmidt gives (up to scaling)

$$q_1 = (0, 2, 1, 0) \quad \|q_1\|^2 = 5$$

$$q_2 = (5, -1, 2, 0) \quad \|q_2\|^2 = 30$$

$$q_3 = (1, 1, -2, -2) \quad \|q_3\|^2 = 10$$

$$q_4 = (1, 1, -2, 3) \quad \|q_4\|^2 = 15$$

So the orthonormal basis Gram-Schmidt yields is found by replacing each q_i by $\frac{q_i}{\|q_i\|}$ ← not squared (which is a unit vector)

$$q_1 = (0, \frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}}, 0)$$

$$q_2 = \left(\frac{5}{\sqrt{30}}, -\frac{1}{\sqrt{30}}, \frac{2}{\sqrt{30}}, 0 \right)$$

$$q_3 = \left(\frac{1}{\sqrt{10}}, \frac{1}{\sqrt{10}}, -\frac{2}{\sqrt{10}}, -\frac{2}{\sqrt{10}} \right)$$

$$q_4 = \left(\frac{1}{\sqrt{15}}, \frac{1}{\sqrt{15}}, -\frac{2}{\sqrt{15}}, \frac{3}{\sqrt{15}} \right)$$

Sage Math

$A.\underset{\text{(underline)}}{\text{gram_schmidt}}()$ [0]

applies Gram-Schmidt to the

(under score)

rows (not columns) of A , and returns matrix G whose rows are orthogonal (not orthonormal)

Previous example in SageMath:

```
A=matrix(QQ, 4, 4, [0,2,1,0, 1, -1, 0, 0, 1, 2, 0, -1, 1, 0, 0, 1])
```

A

```
[ 0  2  1  0] ← rows of A are given vectors
[ 1 -1  0  0]
[ 1  2  0 -1]
[ 1  0  0  1]
```

```
A.gram_schmidt()
```

```
([ 0   2   1   0]
 [ 1 -1/5 2/5  0]
 [ 1/2 1/2 -1   -1]
 [ 4/15 4/15 -8/15 4/5], [ 1   0   0   0]
 [-2/5 1     0   0]
 [ 4/5 1/2 1   0]
 [ 0   5/6 -1/5 1])
```

?? a list of two matrices

```
typeset_mode() ← toggle to make output easier to read
```

```
A.gram_schmidt()
```

returns a pair of matrices (G, L)
with $A = LG$

this is the
matrix whose rows are
the output of
gram-schmidt

$$\text{row 2} = \left(1, -\frac{1}{5}, \frac{2}{5}, 0\right)$$

our answer by hand = $(5, -1, 2, 0)$ is a scalar multiple

Answer

$$\boxed{\begin{pmatrix} 0 & 2 & 1 & 0 \\ 1 & -\frac{1}{5} & \frac{2}{5} & 0 \\ \frac{1}{2} & \frac{1}{2} & -1 & -1 \\ \frac{4}{15} & \frac{4}{15} & -\frac{8}{15} & \frac{4}{5} \end{pmatrix}}, \begin{pmatrix} 1 & 0 & 0 & 0 \\ -\frac{2}{5} & 1 & 0 & 0 \\ \frac{5}{6} & \frac{1}{2} & 1 & 0 \\ 0 & -\frac{1}{5} & 1 & 1 \end{pmatrix}$$

G

L

lower +

If you need orthonormal output of Gram-Schmidt

use $A = \text{matrix}(\text{QQbar}, \dots)$

RDF - real decimal field
QQbar

```
A=matrix(QQbar, 4, 4, [0,2,1,0, 1, -1, 0, 0, 1, 2, 0, -1, 1, 0, 0, 1])
```

A

$$\begin{pmatrix} 0 & 2 & 1 & 0 \\ 1 & -1 & 0 & 0 \\ 1 & 2 & 0 & -1 \\ 1 & 0 & 0 & 1 \end{pmatrix}$$

```
A.gram_schmidt(orthonormal=True)
```

$$\left(\begin{array}{cccc} 0 & 0.8944271909999159? & 0.4472135954999580? & 0 \\ 0.912870929175277? & -0.1825741858350554? & 0.3651483716701108? & 0 \\ 0.3162277660168379? & 0.3162277660168379? & -0.6324555320336758? & -0.6324555320336758? \\ 0.2581988897471611? & 0.2581988897471611? & -0.516397779494323? & 0.774596669241484? \end{array} \right), \left(\begin{array}{cccc} 2.236067977499790? & 0 & 0 & 0 \\ -0.8944271909999159? & 1.095445115010333? & 1.788854381999832? & 0.5477225575051661? \\ 0 & 0 & 0 & 0 \\ 0.912870929175277? & 0 & 0 & 0 \end{array} \right)$$

```
G=A.gram_schmidt(orthonormal=True)[0]
```

G

$$\left(\begin{array}{cccc} 0 & 0.8944271909999159? & 0.4472135954999580? & 0 \\ 0.912870929175277? & -0.1825741858350554? & 0.3651483716701108? & 0 \\ 0.3162277660168379? & 0.3162277660168379? & -0.6324555320336758? & -0.6324555320336758? \\ 0.2581988897471611? & 0.2581988897471611? & -0.516397779494323? & 0.774596669241484? \end{array} \right)$$

G.T*G

T=transpose.
could also write

G.transpose()

$$\left(\begin{array}{cccc} 1.000000000000000? & 0.?e-17 & 0.?e-17 & 0.?e-17 \\ 0.?e-17 & 1.000000000000000? & 0.?e-17 & 0.?e-17 \\ 0.?e-17 & 0.?e-17 & 1.000000000000000? & 0.?e-17 \\ 0.?e-17 & 0.?e-17 & 0.?e-17 & 1.000000000000000? \end{array} \right)$$

$$10^{-17} \approx 0$$

question marks mean
decimal approximation

Prof Robert Lipshitz (U. Oregon) came up with the following "better gram-schmidt" function

Define a new function "better_gs" that takes a

- takes a list of vectors
- makes a matrix A whose rows are those vectors
- applies built-in gram-schmidt to get orthogonal rows
- scales each row by norm to get orthonormal rows

```
def better_gs(vectors):
    A=matrix(vectors)
    G=A.gram_schmidt()[0]
    return [v/v.norm() for v in G.rows()]

v1=vector([0, 2, 1, 0])
v2=vector([1, -1, 0, 0])
v3=vector([1, 2, 0, -1])
v4=vector([1, 0, 0, 1])
```

v1

$$(0, 2, 1, 0)$$

v1.norm()

$$\sqrt{5}$$

better_gs([v1, v2, v3, v4])

$$[(0, \frac{2}{5}\sqrt{5}, \frac{1}{5}\sqrt{5}, 0), (\frac{5}{6}\sqrt{\frac{6}{5}}, -\frac{1}{6}\sqrt{\frac{6}{5}}, \frac{1}{3}\sqrt{\frac{6}{5}}, 0), (\frac{1}{5}\sqrt{\frac{5}{2}}, \frac{1}{5}\sqrt{\frac{5}{2}}, -\frac{2}{5}\sqrt{\frac{5}{2}}, -\frac{2}{5}\sqrt{\frac{5}{2}}), (\sqrt{\frac{1}{15}}, \sqrt{\frac{1}{15}}, -2\sqrt{\frac{1}{15}}, 3\sqrt{\frac{1}{15}})]$$

QR factorization. This will follow from Gram-Schmidt

In recent years a numerical algorithm based on the Gram-Schmidt process, and known as **QR-decomposition**, has assumed growing importance as the mathematical foundation for a wide variety of numerical algorithms, including those for computing eigenvalues of large matrices. The technical aspects of such algorithms are discussed in books that specialize in the numerical aspects of linear algebra. However, we will discuss some of the underlying ideas here.

Anton
6.3

Given an $n \times n$ matrix A of rank n , we can write it as

$$A = Q R$$

↑
orthogonal matrix ← upper triangular matrix

An orthogonal matrix Q is a $n \times n$ matrix Q such that

$$Q^T Q = I_n, \text{ equivalently } [Q^T = Q^{-1}], \text{ and so for orthogonal}$$

This condition shows the columns (or rows) of an orthogonal matrix are not just orthogonal, but orthonormal.

Here's how to find the $A = QR$ factorization using Gram-Schmidt

If v_1, \dots, v_n are the columns of A , apply Gram-Schmidt to get an orthonormal (not just orthogonal) basis q_1, \dots, q_n .

Set Q to be the matrix whose columns are q_i .

$$Q = [q_1 \ q_2 \ \dots \ q_n]$$

For $i=1, 2, \dots, n$

key insight: v_1, v_2, \dots, v_i is in the span of q_1, q_2, \dots, q_i (think about the Gram Schmidt process)

$$v_1 = \langle v_1, q_1 \rangle q_1$$

$$v_2 = \langle v_2, q_1 \rangle q_1 + \langle v_2, q_2 \rangle q_2$$

$$v_3 = \langle v_3, q_1 \rangle q_1 + \langle v_3, q_2 \rangle q_2 + \langle v_3, q_3 \rangle q_3$$

so $A = QR$ where

$$\underbrace{\begin{bmatrix} v_1 & v_2 & \dots & v_n \end{bmatrix}}_{A} = \underbrace{\begin{bmatrix} q_1 & \dots & q_n \end{bmatrix}}_Q \left[\begin{array}{cccc} \langle v_1, q_1 \rangle & \langle v_2, q_1 \rangle & \dots & \langle v_n, q_1 \rangle \\ \langle v_1, q_2 \rangle & \langle v_2, q_2 \rangle & & \langle v_n, q_2 \rangle \\ \vdots & & & \langle v_n, q_n \rangle \end{array} \right] \underbrace{\begin{bmatrix} & & & \\ & & & \\ & & & \\ & & & \end{bmatrix}}_R$$

upper triangular

$$A = QR$$

$$A^+ = (QR)^T = R^T Q^T$$