





**Definition 6.13.**  $A \in M_{n(\times n)}(\mathbf{F})$  define  $A^* = \overline{A^T}$  adjoint matrix

$$(A^*)_{ij} = \overline{A_{ji}}$$

Ex.  $A = \begin{pmatrix} 1 & 1 + 2i \\ 2 & 2 - 3i \end{pmatrix}$   $A^* = \begin{pmatrix} 1 & 2 \\ 1 - 2i & 2 + 3i \end{pmatrix}$

**Example 6.14.** Frobenius inner product on  $M_n(\mathbf{F}) = V$

$$\langle A, B \rangle = \text{tr}(B^* A)$$

$$\begin{aligned} \langle A + B, C \rangle &= \text{tr}(C^*(A + B)) \\ &= \text{tr}(C^* A + C^* B) \stackrel{\text{tr additive; triv.}}{=} \text{tr}(C^* A) + \text{tr}(C^* B) \\ &= \langle A, C \rangle + \langle B, C \rangle \end{aligned}$$

$$\begin{aligned} \langle A, A \rangle &= \text{tr}(A^* A) = \sum_{i=1}^n (A^* A)_{ii} = \\ &= \sum_{i=1}^n \sum_{k=1}^n (A^*)_{ik} A_{ki} = \sum_{i=1}^n \sum_{k=1}^n \overline{A_{ki}} A_{ki} \\ &= \sum_{i=1}^n \sum_{k=1}^n |A_{ki}|^2 \iff \exists k, i : A_{k,i} \neq 0 \\ &\iff A \neq 0 \end{aligned}$$

**Theorem 6.15.** If  $\langle \cdot, \cdot \rangle$  inner product, then  $\|\mathbf{v}\| := \sqrt{\langle \mathbf{v}, \mathbf{v} \rangle}$  norm.

moreover it satisfies the

Cauchy-Schwarz inequality  $|\langle \mathbf{x}, \mathbf{y} \rangle| \leq \|\mathbf{x}\| \cdot \|\mathbf{y}\|$

parallelogram identity  $\|\mathbf{a} + \mathbf{b}\|^2 + \|\mathbf{a} - \mathbf{b}\|^2 = 2(\|\mathbf{a}\|^2 + \|\mathbf{b}\|^2)$   
 $\sim$  test normed space is not an inner product space

**Definition 6.16.** If  $(V, \langle \cdot, \cdot \rangle)$  is a complete metric space with

$$\|\cdot\| = \sqrt{\langle \cdot, \cdot \rangle}, \text{ then it is a Hilbert space .}$$

$\left( \begin{array}{c} \text{inner product} \\ \text{inner product space} \end{array} \right)$	$\rightsquigarrow$	$\left( \begin{array}{c} \text{norm} \\ \text{normed space} \end{array} \right)$	$\rightsquigarrow$	$\left( \begin{array}{c} \text{distance} \\ \text{metric space} \end{array} \right)$
$\updownarrow$		$\updownarrow$		$\updownarrow$
Hilbert space	$\rightsquigarrow$	Banach space	$\rightsquigarrow$	complete metric space
<u>Ex.</u> $L^2(X)$		$L^p(X) \quad p \geq 1$		

**Theorem 6.17.** An inner product has the following properties:

- 1) sesquilinear ( $\mathbf{F} = \mathbb{C}$ ) or bilinear ( $\mathbf{F} = \mathbb{R}$ )
- 2)  $\langle \mathbf{x}, \mathbf{0} \rangle = \langle \mathbf{0}, \mathbf{x} \rangle = 0$
- 3)  $\langle \mathbf{x}, \mathbf{x} \rangle = 0 \iff \mathbf{x} = \mathbf{0}$  (positive definiteness)
- 4) if  $\langle \mathbf{x}, \mathbf{y} \rangle = \langle \mathbf{x}, \mathbf{z} \rangle \quad \forall \mathbf{x} \Rightarrow \mathbf{y} = \mathbf{z}$  (non-degeneracy)

**Example 6.18.**  $V = \left\{ f : [0, 2\pi] \rightarrow \mathbb{C} \quad \langle f, g \rangle = \frac{1}{2\pi} \int_0^{2\pi} f(x) \overline{g(x)} dx \right.$   
 $\left. \int_0^{2\pi} |f|^2 dx < \infty \right\}$   
 $= L^2([0, 2\pi]) \cong L^2(S^1)$  (Fourier calculus)

Consider  $f_n(t) = e^{int} \quad n \in \mathbb{Z}$

( $f_0 \equiv 1$ )

$$\begin{aligned} \langle f_m, f_n \rangle &= \frac{1}{2\pi} \int_0^{2\pi} e^{imt} \overline{e^{int}} dt && (m \neq n) \\ &= \frac{1}{2\pi} \int_0^{2\pi} e^{i(m-n)t} dt = \frac{1}{2\pi i(m-n)} e^{i(m-n)t} \Big|_0^{2\pi} = 0 \\ \|f_n\|^2 = \langle f_n, f_n \rangle &= \frac{1}{2\pi} \int_0^{2\pi} e^{int} \overline{e^{int}} dt \\ &= \frac{1}{2\pi} \int_0^{2\pi} \underbrace{|e^{int}|^2}_1 dt = \frac{1}{2\pi} \cdot 2\pi = 1. \end{aligned}$$

$$\langle f_m, f_n \rangle = \delta_{mn}$$

**Definition 6.19.** If  $\langle \mathbf{v}, \mathbf{w} \rangle = 0$ , say  $\mathbf{v}$  is orthogonal

(perpendicular) to  $\mathbf{w}$  and write  $\mathbf{v} \perp \mathbf{w}$ .

$S \subset V$  is an orthogonal set if  $\mathbf{v} \perp \mathbf{w} \quad \forall \mathbf{v} \neq \mathbf{w} \in S$

$S$  is orthonormal if  $S$  orthogonal &  $\|\mathbf{v}\| = 1 \quad \forall \mathbf{v} \in S$

$S \subset V$  is a complete orthonormal if  $S$  orthonormal and

$$\forall \mathbf{v} \in V \setminus \{\mathbf{0}\} : \exists \mathbf{v}' \in S : \langle \mathbf{v}, \mathbf{v}' \rangle \neq 0$$

**Remark 6.20.**  $S = \{\mathbf{v}_1, \dots, \mathbf{v}_n\} \neq \mathbf{0}$  orthogonal  $\mathbf{v}_i \rightsquigarrow \frac{\mathbf{v}_i}{\|\mathbf{v}_i\|}$  normalization  $\rightarrow S' = \{\mathbf{v}_i / \|\mathbf{v}_i\|\}$   
 orthonormal set

**Theorem 6.21.** (Riesz-Fischer)  $\left\{ e^{int} \right\}_{n \in \mathbb{Z}}$  is a complete orthonormal set  
 on  $V = L^2(S^1)$

## 6.2. Gram-Schmidt and orthogonal complements.

**Definition 6.22.**  $S \subset V$  inner product space is orthonormal basis if  
 $S$  is orthonormal set and a basis

**Example 6.23.**  $\left\{ \frac{1}{\sqrt{5}}(2, 1), \frac{1}{\sqrt{5}}(1, -2) \right\}$  is orthonormal basis  
 of  $\mathbb{R}^2$

**Example 6.24.**  $L^2(S^1)$  has a complete orthonormal set  $S$   
 which is not a basis. (It is not clear how to find an orthonormal basis.)

(by Gram-Schmidt)



We will later see that in finite dim.

complete orthonormal set  $\simeq$  orthonormal basis

**Theorem 6.25.** *If  $V$  inner product space  $\mathcal{E} S = \{\mathbf{v}_1, \dots, \mathbf{v}_k\}$  orthogonal set ( $\mathbf{v}_i \neq \mathbf{0}$ )  $\mathcal{E} \mathbf{y} \in \text{span}(S)$  then*

$$\mathbf{y} = \sum_{i=1}^k \frac{\langle \mathbf{y}, \mathbf{v}_i \rangle}{\|\mathbf{v}_i\|^2} \mathbf{v}_i \tag{1}$$

*in particular if  $S$  is orthonormal*

$$\mathbf{y} = \sum_{i=1}^k \langle \mathbf{y}, \mathbf{v}_i \rangle \mathbf{v}_i \tag{2}$$

*Proof.* (2)  $\implies$  (1) by normalization, so prove (2).

$$\text{Let } \mathbf{y} = \sum a_i \mathbf{v}_i \iff \mathbf{y} \in \text{span}(S)$$

Then

$$\begin{aligned} \langle \mathbf{y}, \mathbf{v}_j \rangle &= \langle \sum a_i \mathbf{v}_i, \mathbf{v}_j \rangle = \sum a_i \langle \mathbf{v}_i, \mathbf{v}_j \rangle \\ &= \sum a_i \delta_{ij} = a_j \quad \square \end{aligned}$$

**Remark 6.26.** (1) and (2) are called orthogonal decomposition of  $\mathbf{y}$  (w.r.t.  $S$ )

**Corollary 6.27.**  $V$  inner product space  $S \not\ni \mathbf{0}_V$  orthogonal ( $\iff$  orthonormal)  $\implies S$  linearly independent

*Proof.* if  $\sum a_i \mathbf{v}_i = \mathbf{0}_V$

$$a_j = \langle \sum a_i \mathbf{v}_i, \mathbf{v}_j \rangle = \langle \mathbf{0}_V, \mathbf{v}_j \rangle = 0 \quad \forall j \quad \square$$

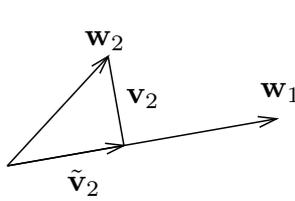
**Corollary 6.28.**

$$\|\mathbf{y}\|^2 = \sum_{i=1}^k |\langle \mathbf{y}, \mathbf{v}_i \rangle|^2 \quad \{\mathbf{v}_i\} \text{ orthonormal set}$$

$$\begin{aligned} \text{Proof. } \langle \mathbf{y}, \mathbf{y} \rangle &= \left\langle \sum_{i=1}^k \overbrace{\langle \mathbf{y}, \mathbf{v}_i \rangle}^{a_i} \mathbf{v}_i, \sum_{j=1}^k \overbrace{\langle \mathbf{y}, \mathbf{v}_j \rangle}^{a_j} \mathbf{v}_j \right\rangle \\ &= \sum_{i=1}^k \sum_{j=1}^k a_i \overline{a_j} \underbrace{\langle \mathbf{v}_i, \mathbf{v}_j \rangle}_{\delta_{ij}} \\ &= \sum_{i=1}^k a_i \overline{a_i} = \sum_{i=1}^k |a_i|^2 \quad \square \end{aligned}$$

How to get orthonormal basis?

orthogonal projection  $\mathbf{w}_1, \mathbf{w}_2 \in \mathbb{R}^2$  linearly independent



$$\mathbf{w}_2 - \tilde{\mathbf{v}}_2 \perp \tilde{\mathbf{v}}_2$$

$$\mathbf{v}_2 = \mathbf{w}_2 - c\mathbf{w}_1 \perp \mathbf{w}_1$$

$$\langle \mathbf{w}_2, \mathbf{w}_1 \rangle = \langle \mathbf{w}_2 - c\mathbf{w}_1, \mathbf{w}_1 \rangle + c\langle \mathbf{w}_1, \mathbf{w}_1 \rangle = c\|\mathbf{w}_1\|^2$$

$$\rightsquigarrow c = \frac{\langle \mathbf{w}_2, \mathbf{w}_1 \rangle}{\|\mathbf{w}_1\|^2}$$

$$\tilde{\mathbf{v}}_2 = \frac{\langle \mathbf{w}_2, \mathbf{w}_1 \rangle}{\|\mathbf{w}_1\|^2} \mathbf{w}_1 \quad \text{orthogonal projection of } \mathbf{w}_2 \text{ onto } \mathbf{w}_1$$

$$\mathbf{v}_2 = \mathbf{w}_2 - \tilde{\mathbf{v}}_2 = \mathbf{w}_2 - \frac{\langle \mathbf{w}_2, \mathbf{w}_1 \rangle}{\|\mathbf{w}_1\|^2} \mathbf{w}_1 \quad \text{orthogonalization of } \mathbf{w}_2 \text{ w.r.t. } \mathbf{w}_1$$

**Theorem 6.29.** Let  $S = \{\mathbf{w}_1, \dots, \mathbf{w}_n\}$  be linear independent set of inner product space  $V$ . Then

$$\mathbf{v}_k = \frac{\mathbf{w}_k - \sum_{j=1}^{k-1} \langle \mathbf{w}_k, \mathbf{v}_j \rangle \mathbf{v}_j}{\left\| \mathbf{w}_k - \sum_{j=1}^{k-1} \langle \mathbf{w}_k, \mathbf{v}_j \rangle \mathbf{v}_j \right\|}$$

*gives an orthonormal set  $S' = \{\mathbf{v}_k\}$  with  $\text{span}(S') = \text{span}(S)$*

Gram-Schmidt orthonormalization

**Example 6.30.**  $\mathbb{R}^4$   $\mathbf{w}_1 = (1, 0, 1, 0)$   $\mathbf{w}_2 = (1, 1, 1, 1)$   $\mathbf{w}_3 = (0, 1, 2, 1)$

$$\mathbf{v}_1 = \frac{\mathbf{w}_1}{\|\mathbf{w}_1\|} = \frac{1}{\sqrt{2}}(1, 0, 1, 0)$$

$$\begin{aligned} \mathbf{v}'_2 &= \mathbf{w}_2 - \langle \mathbf{w}_2, \mathbf{v}_1 \rangle \mathbf{v}_1 = (1, 1, 1, 1) - \frac{1}{\sqrt{2}} \cdot 2 \cdot \frac{1}{\sqrt{2}}(1, 0, 1, 0) \\ &= (0, 1, 0, 1) \end{aligned}$$

$$\mathbf{v}_2 = \frac{(0, 1, 0, 1)}{\|(0, 1, 0, 1)\|} = \frac{1}{\sqrt{2}}(0, 1, 0, 1)$$

$$\begin{aligned} \mathbf{v}'_3 &= \mathbf{w}_3 - \langle \mathbf{w}_3, \mathbf{v}_1 \rangle \mathbf{v}_1 - \langle \mathbf{w}_3, \mathbf{v}_2 \rangle \mathbf{v}_2 \\ &= (0, 1, 2, 1) - \frac{1}{\sqrt{2}} \cdot 2 \cdot \frac{1}{\sqrt{2}}(1, 0, 1, 0) - \frac{1}{\sqrt{2}} \cdot 2 \cdot \frac{1}{\sqrt{2}}(0, 1, 0, 1) \\ &= (-1, 0, 1, 0) \end{aligned}$$

$$\mathbf{v}_3 = \frac{\mathbf{v}_3}{\|\mathbf{v}_3\|} = \frac{1}{\sqrt{2}}(-1, 0, 1, 0)$$

**Example 6.31.**  $V = \mathcal{P}(\mathbb{R})$      $\langle f, g \rangle = \int_{-1}^1 f \cdot g$

$$S = \{1, x, x^2\}$$

$$\tilde{\mathbf{v}}_1 = \mathbf{w}_1 = 1$$

$$\|\mathbf{w}_1\|^2 = \int_{-1}^1 1^2 dx = 2 \quad \mathbf{v}_1 = \frac{\mathbf{w}_1}{\|\mathbf{w}_1\|} = \underline{\underline{\frac{1}{\sqrt{2}}}}$$

$$\begin{aligned} \mathbf{w}_2 = x \quad \tilde{\mathbf{v}}_2 &= \mathbf{w}_2 - \langle \mathbf{w}_2, \mathbf{v}_1 \rangle \mathbf{v}_1 & \langle \mathbf{w}_2, \mathbf{v}_1 \rangle &= \frac{1}{\sqrt{2}} \int_{-1}^1 1 \cdot x dx = 0 \\ &= \mathbf{w}_2 = x \end{aligned}$$

$$\mathbf{v}_2 = \frac{\tilde{\mathbf{v}}_2}{\|\tilde{\mathbf{v}}_2\|} = \underline{\underline{\sqrt{\frac{3}{2}}x}} \quad \|\tilde{\mathbf{v}}_2\|^2 = \int_{-1}^1 x^2 dx = \frac{2}{3}$$

$$\begin{aligned} \mathbf{w}_3 = x^2 \quad \tilde{\mathbf{v}}_3 &= \mathbf{w}_3 - \langle \mathbf{w}_3, \mathbf{v}_1 \rangle \mathbf{v}_1 - \\ &\quad - \langle \mathbf{w}_3, \mathbf{v}_2 \rangle \mathbf{v}_2 \end{aligned} \quad \langle \mathbf{w}_3, \mathbf{v}_1 \rangle = \frac{1}{\sqrt{2}} \int_{-1}^1 x^2 dx = \frac{\sqrt{2}}{3}$$

$$= \mathbf{w}_3 - \frac{\sqrt{2}}{3} \cdot \frac{1}{\sqrt{2}} \quad \langle \mathbf{w}_3, \mathbf{v}_2 \rangle = \sqrt{\frac{3}{2}} \int_{-1}^1 x^3 dx = 0$$

$$= x^2 - \frac{1}{3} \quad \|\tilde{\mathbf{v}}_3\|^2 = \int_{-1}^1 x^4 - \frac{2}{3}x^2 + \frac{1}{9} dx$$

$$\begin{aligned} \mathbf{v}_3 = \frac{\tilde{\mathbf{v}}_3}{\|\tilde{\mathbf{v}}_3\|} &= \sqrt{\frac{45}{8}} \left( x^2 - \frac{1}{3} \right) = \underline{\underline{\sqrt{\frac{5}{8}}(3x^2 - 1)}} & &= \frac{2}{5} - \frac{4}{9} + \frac{2}{9} \\ & & &= \frac{8}{45} \end{aligned}$$

**Example 6.32.** (continued)

express  $f = 1 + 2x + 3x^2$  as a linear combination of  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$

$f = \sum a_i \mathbf{v}_i$      $a_i = \langle f, \mathbf{v}_i \rangle$     calculate inner product instead  
of solving a linear eqn system !

$$a_1 = \int_{-1}^1 \frac{1}{\sqrt{2}} (1 + 2x + 3x^2) dx = \frac{2}{\sqrt{2}} + \frac{1}{\sqrt{2}} \left( x^2 + x^3 \Big|_{-1}^1 \right) = \frac{4}{\sqrt{2}} = 2\sqrt{2}$$

$$\begin{aligned} a_2 &= \int_{-1}^1 \sqrt{\frac{3}{2}} x (1 + 2x + 3x^2) dx = \sqrt{\frac{3}{2}} \left( \frac{x^2}{2} + \frac{2}{3}x^3 + \frac{3}{4}x^4 \right) \Big|_{-1}^1 \\ &= \frac{4}{3} \sqrt{\frac{3}{2}} = \frac{2\sqrt{6}}{3} \end{aligned}$$

$$a_3 = \int_{-1}^1 \sqrt{\frac{5}{8}} (3x^2 - 1)(1 + 2x + 3x^2) dx = \dots = \sqrt{\frac{8}{5}}$$

fin. dim. space $\{\mathbf{v}_i\}$ orthonormal basis $\mathbf{x} = \sum_{i=1}^n \langle \mathbf{x}, \mathbf{v}_i \rangle \mathbf{v}_i$ (ortho. decomp.) $\ \mathbf{x}\ ^2 = \sum_{i=1}^n  \langle \mathbf{x}, \mathbf{v}_i \rangle ^2$	$\rightsquigarrow$	$\infty$ dim. (Hilbert) space $\{\mathbf{v}_i\}$ (countable) complete orth. set $\mathbf{x} = \sum_{i=1}^{\infty} \langle \mathbf{x}, \mathbf{v}_i \rangle \mathbf{v}_i$ converges in norm $\left\  \mathbf{x} - \sum_{i=1}^n \langle \mathbf{x}, \mathbf{v}_i \rangle \mathbf{v}_i \right\  \xrightarrow{n \rightarrow \infty} 0$ $\ \mathbf{x}\ ^2 = \sum_{i=1}^{\infty}  \langle \mathbf{x}, \mathbf{v}_i \rangle ^2$
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When  $V = L^2([0, 2\pi])$  and  $S = \{e^{int}\}$ ,  
 the orthogonal decomposition

$$f(t) = \sum_{n=-\infty}^{\infty} \langle f, e^{int} \rangle e^{int}$$

is called Fourier series (and  $\langle f, e^{int} \rangle$   
Fourier coefficients) of  $f$ .

**Example 6.33.**  $L^2([0, 2\pi]) \supset S = \{ \underbrace{e^{int}}_{f_n} : n \in \mathbb{Z} \}$   
 $f(t) = t$

$$\|f\|^2 = \frac{1}{2\pi} \int_0^{2\pi} t \cdot \bar{t} dt = \frac{1}{2\pi} \int_0^{2\pi} t^2 dt = \frac{4\pi^2}{3}.$$

$$\langle f, f_0 \rangle = \frac{1}{2\pi} \int_0^{2\pi} t \cdot \bar{1} dt = \frac{2\pi^2}{2\pi} = \pi \quad |\langle f, f_0 \rangle|^2 = \pi^2$$

$$\langle f, f_n \rangle = \frac{1}{2\pi} \int_0^{2\pi} t \cdot \overline{e^{int}} dt = \frac{1}{2\pi} \int_0^{2\pi} t \cdot e^{-int} dt$$

$$= \frac{1}{2\pi} \left( \underbrace{\frac{t}{-in} e^{-int} \Big|_0^{2\pi}}_{\frac{2\pi}{-in}} - \underbrace{\frac{1}{(-in)^2} e^{-int} \Big|_0^{2\pi}}_0 \right) \int te^{-int} dt = \frac{t}{-in} e^{-int} - \frac{1}{(-in)^2} e^{-int} + C$$

$$= -\frac{1}{in} = \frac{i}{n} \quad (i = \sqrt{-1})$$

$$|\langle f, f_n \rangle|^2 = \frac{1}{n^2} \quad n \in \mathbb{Z} \setminus \{0\}$$

Thus

$$\frac{4\pi^2}{3} = \|f\|^2 = \sum_{n=-\infty}^{\infty} |\langle f, e^{int} \rangle|^2 = \pi^2 + 2 \sum_{n=1}^{\infty} \frac{1}{n^2},$$

$$\zeta(2) = \sum_{n=1}^{\infty} \frac{1}{n^2} = \frac{\pi^2}{6} \quad \text{even Riemann zeta values}$$

$$\zeta(z) = \sum_{n=1}^{\infty} \frac{1}{n^z} \text{ converges absolutely for } z \in \mathbb{C} \text{ if } \Re z > 1$$

Riemann zeta function

( $\rightsquigarrow$  Riemann hypothesis etc.)

similarly it is known

$$\zeta(2k) = \pi^{2k} \cdot \langle \text{something rational} \rangle$$

but odd  $\zeta(2k+1)$  ?? odd Riemann zeta values?

$\zeta(3) \notin \mathbb{Q}$  is known, and

$$\dim_{\mathbb{Q}} \text{span}_{\mathbb{Q}} \{ \zeta(2k+1) : k \geq 1 \} = \infty \quad \begin{array}{l} \text{latest cry} \\ \text{of research} \end{array} !$$

$\downarrow$   
(so  $\infty$  many are irrational, but which?)

**Example 6.34.**  $\sum_{n=0}^{\infty} \frac{1}{n^2 + n + 1} \quad \left(n + \frac{1}{2}\right)^2 + \frac{3}{4}$

$$\langle f, g \rangle = \frac{1}{2\pi} \int_0^{2\pi} f \bar{g}$$

$$f = e^{-it/2 + \frac{\sqrt{3}}{2}t} \in L^2([0, 2\pi])$$

$$\|f\|^2 = \frac{1}{2\pi} \int_0^{2\pi} e^{\sqrt{3}t} dt = \frac{1}{2\sqrt{3}\pi} (e^{2\sqrt{3}\pi} - 1)$$

$$\langle f, e^{int} \rangle = \frac{1}{2\pi} \int_0^{2\pi} e^{t(\frac{\sqrt{3}}{2} + (-1/2 - n)i)} dt$$

$$= \frac{1}{2\pi \left(\frac{\sqrt{3}}{2} - (n + 1/2)i\right)} \left( -e^{\sqrt{3}\pi} - 1 \right) e^{-(2n+1)\pi i} = -1$$

$$|\langle f, e^{int} \rangle|^2 = \frac{1}{(2\pi)^2 \left(\frac{3}{4} + (n + 1/2)^2\right)} (e^{\sqrt{3}\pi} + 1)^2$$

$$n^2 + n + 1$$

$$\|f\|^2 = \sum_{n=-\infty}^{\infty} |\langle f, e^{int} \rangle|^2$$

$$\begin{aligned} \frac{1}{2\sqrt{3}\pi} \left( e^{2\sqrt{3}\pi} - 1 \right) &= \sum_{n=-\infty}^{\infty} \frac{1}{(2\pi)^2(n^2 + n + 1)} \left( e^{\sqrt{3}\pi} + 1 \right)^2 \\ \frac{2\pi}{\sqrt{3}} \cdot \frac{e^{\sqrt{3}\pi} - 1}{e^{\sqrt{3}\pi} + 1} &= \sum_{n=-\infty}^{\infty} \frac{1}{n^2 + n + 1} \\ &= \sum_{n=0}^{\infty} \frac{1}{n^2 + n + 1} + \sum_{n=0}^{\infty} \frac{1}{(-n-1)^2 + (-n-1) + 1} = 2 \sum_{n=0}^{\infty} \frac{1}{n^2 + n + 1} \\ 1.798 \approx \frac{\pi}{\sqrt{3}} \frac{e^{\sqrt{3}\pi} - 1}{e^{\sqrt{3}\pi} + 1} &= \sum_{n=0}^{\infty} \frac{1}{n^2 + n + 1} \end{aligned}$$

### orthogonal complements

**Definition 6.35.** set  $S \subset V$  inner prod. space

$$S^\perp = \{ \mathbf{x} \in V : \forall s \in S \quad \langle \mathbf{x}, \mathbf{s} \rangle = 0 \}$$

orthogonal complement of  $S$

**Theorem 6.36.**  $S^\perp$  is a linear subspace of  $V$ .

**Example 6.37.**  $\{0\}^\perp = V \quad V^\perp = \{0\}$

$$\begin{aligned} &\updownarrow \\ \forall \mathbf{v} \in V \quad \langle \mathbf{x}, \mathbf{v} \rangle = 0 &\stackrel{\mathbf{x}=\mathbf{v}}{\implies} \langle \mathbf{x}, \mathbf{x} \rangle = 0 \\ &\parallel \\ &\|\mathbf{x}\|^2 \\ &\implies \|\mathbf{x}\| = 0 \implies \mathbf{x} = \mathbf{0} \end{aligned}$$

**Example 6.38.** Let  $V = L^2([0, 2\pi]) \quad S = \underbrace{\text{span}\{ e^{int} \}}_{S'}$

$$S'^\perp = S^\perp = \{0\} \text{ although } S \neq V$$

↑  
 $S'$  complete orth. system

so  $V$  has a proper subspace with trivial orthogonal complement (this does not happen in finite dim.)

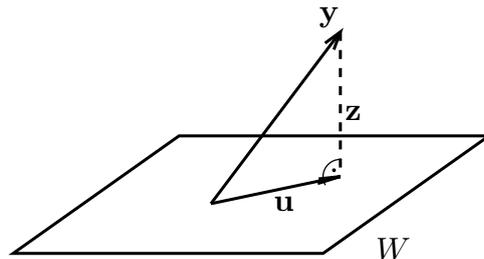
**Theorem 6.39.** (orthogonal decomposition)

Let  $V \supset W$  f.d. subspace  $\mathbf{y} \in V$ .

Then  $\mathbf{y} = \underbrace{\mathbf{u}}_W + \underbrace{\mathbf{z}}_{W^\perp}$  unique decomposition

and  $\mathbf{u}$  'closest' to  $\mathbf{y}$  on  $W$ :

$$\|\mathbf{y} - \mathbf{u}\| \leq \|\mathbf{y} - \mathbf{x}\| \quad \forall \mathbf{x} \in W$$



*Proof.*  $\mathbf{u} := \sum \langle \mathbf{y}, \mathbf{v}_i \rangle \mathbf{v}_i$  for some orthonormal basis

of  $W$

$$\mathbf{z} := \mathbf{y} - \mathbf{u}$$

$$\begin{aligned} \|\mathbf{y} - \mathbf{x}\|^2 &= \|\mathbf{u} + \mathbf{z} - \mathbf{x}\|^2 = \|(\mathbf{u} - \mathbf{x}) + \mathbf{z}\|^2 \\ &= \|\mathbf{u} - \mathbf{x}\|^2 + \|\mathbf{z}\|^2 \geq \|\mathbf{z}\|^2 = \|\mathbf{y} - \mathbf{u}\|^2. \end{aligned}$$

□

$$\begin{array}{l} \mathbf{u} - \mathbf{x} \in W \\ \mathbf{z} \in W^\perp \end{array} \quad \langle \mathbf{u} - \mathbf{x}, \mathbf{z} \rangle = 0$$

**Definition 6.40.**  $\mathbf{u}$  orthogonal projection of  $\mathbf{y}$  on  $W$   
 $\mathbf{z}$  orthogonalization of  $\mathbf{y}$  w.r.t.  $W$

**Theorem 6.41.**  $V$  fin. dim. VS  $W$  subspace

$$V = W \oplus W^\perp \quad \dim(V) = \dim(W) + \dim(W^\perp)$$

### 6.3. The adjoint of a linear operator.

$$A \in M_n(\mathbf{F}) \rightsquigarrow A^* = \overline{A^T}$$

now we define the corresponding linear operator

**Theorem 6.42.**  $V$  fin. dim. inner product space /  $\mathbf{F}$

$$\forall g : V \rightarrow \mathbf{F} \text{ linear} \quad \exists! \mathbf{x} \in V : \forall \mathbf{y} \in V \quad g(\mathbf{y}) = \langle \mathbf{y}, \mathbf{x} \rangle$$

*I can represent a linear map to  $\mathbf{F}$  as an inner product with a vector  $\mathbf{x}$ .*

*Proof.*  $\{\mathbf{v}_i\}$  orthonormal basis

$$\mathbf{x} = \sum \overline{g(\mathbf{v}_i)} \mathbf{v}_i$$

$\mathbf{x}$  unique because  $\langle \cdot, \cdot \rangle$  non-degenerate. □

Now let

$$T : V \rightarrow V \text{ linear} \quad \mathbf{y} \in V$$

$$\begin{array}{ccc} \mathbf{x} & \longmapsto & \langle T(\mathbf{x}), \mathbf{y} \rangle \\ \cap & & \cap \\ V & & \mathbf{F} \end{array} \quad \text{is map to } \mathbf{F} \text{ linear in } \mathbf{x}$$

$$\overset{\text{Theorem}}{\rightsquigarrow} \exists (!) \mathbf{y}' \in V \quad \langle T(\mathbf{x}), \mathbf{y} \rangle = \langle \mathbf{x}, \mathbf{y}' \rangle$$

map  $\mathbf{y} \mapsto \mathbf{y}'$  is called adjoint map to  $T$

**Definition 6.43.**  $T : V \rightarrow V$  Define  $T^* : V \rightarrow V$  by

$$\langle T(\mathbf{x}), \mathbf{y} \rangle = \langle \mathbf{x}, T^*(\mathbf{y}) \rangle \quad \forall \mathbf{x}, \mathbf{y} \in V$$

**Theorem 6.44.**  $T^*$  is linear and if  $\beta$  is orthonormal basis of  $V$  then

$$[T^*]_\beta = [T]_\beta^* \quad \left( = \overline{[T]_\beta^T} \right)$$

(Proof: book)

**Corollary 6.45.**  $A \in M_n(\mathbf{F}) \quad L_{A^*} = (L_A)^*$

$$(T^*)^* = T$$

6.4. Self-adjoint operators.

**Definition 6.46.**

$$\begin{aligned}
 T : V \rightarrow V & \text{ self-adjoint (Hermitian) if } T = T^* \\
 A \in M_n(\mathbf{F}) & \text{ self-adjoint if } L_A = L_A^* \iff A = A^* \\
 \hookrightarrow \mathbf{F} = \mathbb{R} & \quad A = A^T \text{ symmetric matrix} \\
 \mathbf{F} = \mathbb{C} & \quad A = \overline{A}^T \text{ Hermitian matrix}
 \end{aligned}$$

$$\langle T(\mathbf{x}), \mathbf{y} \rangle = \langle \mathbf{x}, T(\mathbf{y}) \rangle \quad \forall \mathbf{x}, \mathbf{y}.$$

**Theorem 6.47.** a)  $\lambda$  EV  $\implies \lambda$  real      b) eigenvectors to different EVs are  $\perp$

*Proof.* a)  $\langle \lambda \mathbf{x}, \mathbf{x} \rangle = \langle T(\mathbf{x}), \mathbf{x} \rangle = \langle \mathbf{x}, T(\mathbf{x}) \rangle = \langle \mathbf{x}, \lambda \mathbf{x} \rangle = \overline{\lambda} \langle \mathbf{x}, \mathbf{x} \rangle$   
 $\parallel$   
 $\lambda \langle \mathbf{x}, \mathbf{x} \rangle \qquad \langle \mathbf{x}, \mathbf{x} \rangle \neq 0 \quad \rightarrow \quad \underline{\lambda = \overline{\lambda}}$

b)  $\langle T(\mathbf{x}), \mathbf{y} \rangle = \lambda_1 \langle \mathbf{x}, \mathbf{y} \rangle$        $T(\mathbf{x}) = \lambda_1 \mathbf{x}$   
 $\parallel$        $T(\mathbf{y}) = \lambda_2 \mathbf{y}$   
 $\langle \mathbf{x}, T(\mathbf{y}) \rangle = \overline{\lambda_2} \langle \mathbf{x}, \mathbf{y} \rangle = \lambda_2 \langle \mathbf{x}, \mathbf{y} \rangle$        $\frac{(\lambda_1 - \lambda_2) \langle \mathbf{x}, \mathbf{y} \rangle}{\neq 0} = 0 \quad \square$   
↑  
a)

**Theorem 6.48.**  $T$  self-adjoint  $\implies \chi_T$  splits.

*Proof.* clearly over  $\mathbf{F} = \mathbb{C}$ . Over  $\mathbf{F} = \mathbb{R}$  regard  $T$  as an operator over  $\mathbb{C}$ , then  $\chi_T$  splits over  $\mathbb{C}$ , but all EV are real, thus  $\chi_T$  also splits over  $\mathbb{R}$ . □

**Theorem 6.49.** (Schur)  $T$  self-adjoint  $\implies T$  is diagonalizable and has an orthonormal eigenbasis

**Definition 6.50.**  $T : V \rightarrow V$  self-adjoint is called positive/negative definite semidefinite

$$\begin{aligned}
 \text{if } \langle T(\mathbf{x}), \mathbf{x} \rangle & > 0 \quad \forall \mathbf{x} \neq \mathbf{0}_V \\
 & < \\
 & \geq \\
 & \leq
 \end{aligned}$$

$$A \text{ positive/negative (semi)definite} \iff L_A \text{ is so}$$

**Theorem 6.51.**  $T, A$  positive def.  $\iff$  all EV  $\lambda > 0$   
 negative def.  $<$   
 positive semidef.  $\geq$   
 negative semidef.  $\leq$

**Example 6.52.**  $f(x_1, \dots, x_n) \in \mathbb{R} \quad f : \mathbb{R}^n \rightarrow \mathbb{R}$

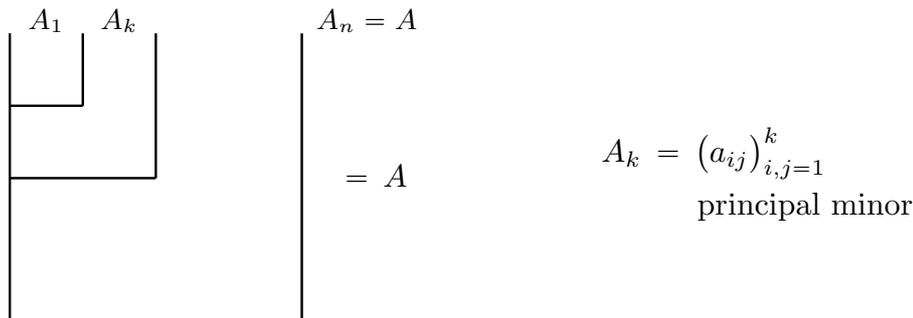
$\nabla f(x_1, \dots, x_n) = 0$       critical point

gradient  $\nearrow$

$$\text{Hess}_f(\mathbf{x}_0) = \left( \frac{\partial^2 f}{\partial x_i \partial x_j}(\mathbf{x}_0) \right)_{i,j=1}^n$$

**Theorem 6.53.**  $\mathbf{x}_0$  critical point of  $f \quad \nabla f(\mathbf{x}_0) = 0$

Hess $_f(\mathbf{x}_0)$  positive def.  $\implies \mathbf{x}_0$  loc. minimum  $\implies$  Hess $_f(\mathbf{x}_0)$  pos. semidef.  
 negative  maximum  neg.



**Theorem 6.54.** (principal minor test)  $A \in M_n(\mathbf{F}) \quad A = A^*$

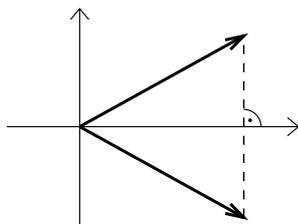
$$\left. \begin{array}{l} A \text{ pos def.} \iff \det A_k > 0 \\ \text{neg def.} \iff (-1)^k \det A_k > 0 \\ \text{pos semid.} \iff \det A_k \geq 0 \\ \text{neg semid.} \iff (-1)^k \det A_k \geq 0 \end{array} \right\} \forall k = 1, \dots, n$$

**6.5. Unitary and orthogonal operators and their matrices.**

**Definition 6.55.**  $T : V \rightarrow V$  is unitary ( $\mathbf{F} = \mathbb{C}$ ) or orthogonal ( $\mathbf{F} = \mathbb{R}$ ) if  $\|T(\mathbf{x})\| = \|\mathbf{x}\| \quad \forall \mathbf{x} \in V$ .  
 preserves length  $A \text{ o/u} : \iff L_A \text{ o/u}$

Ex. reflection in  $\mathbb{R}^2$

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

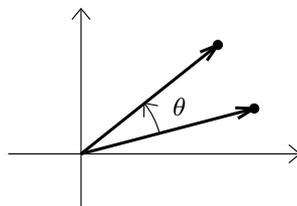


Ex.  $\pm Id$  ,  $\lambda Id$

$$\mathbf{F} = \mathbb{R} \quad \begin{array}{l} |\lambda| = 1 \\ \mathbf{F} = \mathbb{C} \end{array}$$

rotation  $\theta \in (0, 2\pi)$

$$R_\theta = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$



**Theorem 6.56.**

$$\begin{array}{l} \dim V < \infty \\ T \text{ orthogonal/unitary} \iff TT^* = T^*T = Id \\ \iff T \text{ invertible and } T^{-1} = T^* \end{array}$$

**Remark 6.57.**  $T \text{ o/u} \implies T$  injective. But if  $\dim V = \infty$  not always surjective. For example, on

$$V = l^2 = \{ (x_1, \dots, x_n, \dots) : \sum_{i=1}^{\infty} |x_i|^2 < \infty \} \quad \langle \mathbf{x}, \mathbf{y} \rangle = \sum_{i=1}^{\infty} x_i \overline{y_i}$$

consider

$$T((x_1, \dots, x_n, \dots)) = (0, x_1, \dots, x_n, \dots).$$

$$\begin{aligned} \text{Lemma 6.58.} \quad \langle \mathbf{x}, \mathbf{y} \rangle &= \frac{\|\mathbf{x} + \mathbf{y}\|^2 - \|\mathbf{x}\|^2 - \|\mathbf{y}\|^2}{2} & \mathbf{F} = \mathbb{R} \\ \Re \langle \mathbf{x}, \mathbf{y} \rangle &= (same) & \left. \vphantom{\langle \mathbf{x}, \mathbf{y} \rangle} \right\} \mathbf{F} = \mathbb{C} \\ \Im \langle \mathbf{x}, \mathbf{y} \rangle &= \Re \langle \mathbf{x}, i\mathbf{y} \rangle \end{aligned}$$

**Lemma 6.59.**  $T$  is orthogonal/unitary  $\iff T$  preserves  $\langle \cdot, \cdot \rangle$ , i.e.,

$$\langle T(\mathbf{x}), T(\mathbf{y}) \rangle = \langle \mathbf{x}, \mathbf{y} \rangle \quad \forall \mathbf{x}, \mathbf{y} \in V$$

*Proof.* “ $\implies$ ” e.g.,  $\mathbf{F} = \mathbb{C}$

$$\begin{aligned} \Im \langle T(\mathbf{x}), T(\mathbf{y}) \rangle &= \Re \langle T(\mathbf{x}), iT(\mathbf{y}) \rangle \\ &= \Re \langle T(\mathbf{x}), T(i\mathbf{y}) \rangle \\ &= \frac{\|T(\mathbf{x}) + iT(\mathbf{y})\|^2 - \|T(\mathbf{x})\|^2 - \|T(i\mathbf{y})\|^2}{2} \\ &= \frac{\|T(\mathbf{x} + i\mathbf{y})\|^2 - \|T(\mathbf{x})\|^2 - \|T(i\mathbf{y})\|^2}{2} \\ &\stackrel{T \text{ unitary}}{=} \frac{\|\mathbf{x} + i\mathbf{y}\|^2 - \|\mathbf{x}\|^2 - \|i\mathbf{y}\|^2}{2} \\ &= \Re \langle \mathbf{x}, i\mathbf{y} \rangle = \Im \langle \mathbf{x}, \mathbf{y} \rangle \quad \square \end{aligned}$$

Proof of theorem 6.56.

$$\begin{aligned} \langle T(\mathbf{x}), T(\mathbf{y}) \rangle &= \langle \mathbf{x}, \mathbf{y} \rangle \quad \forall \mathbf{x}, \mathbf{y} \in V \\ &\parallel \\ \langle T^*T(\mathbf{x}), \mathbf{y} \rangle & \\ \langle \cdot, \cdot \rangle & \\ \text{non-degenerate} & \\ \implies \mathbf{x} = T^*T(\mathbf{x}) &\implies T^*T = Id \quad \square \end{aligned}$$

**Corollary 6.60.**  $|\det(T)| = 1$ .

$$\begin{aligned} \text{Proof. } \det(T^*T) &= \det(T^*) \det(T) = \\ &= \overline{\det(T)} \det(T) = |\det(T)|^2 = 1 \quad \square \end{aligned}$$

**Theorem 6.61.**  $T$  unitary/orth.  $\lambda$  EV  $\implies |\lambda| = 1$ .

$$\text{Proof. } \|\mathbf{x}\| = \|T(\mathbf{x})\| = \|\lambda\mathbf{x}\| = |\lambda| \|\mathbf{x}\|. \quad \square$$

**Example 6.62.** (!! ) rotation shows that  $T$  may not have EV (over  $\mathbf{F} = \mathbb{R}$ ) in particular not diagonalizable!

(Below we'll see this doesn't occur for  $\mathbf{F} = \mathbb{C}$ .)

**Theorem 6.63.**  $T$  unitary/orth. eigenvectors to different EV are  $\perp$ .

*Proof.*  $T(\mathbf{x}) = \lambda_1 \mathbf{x}$   
 $T(\mathbf{y}) = \lambda_2 \mathbf{y}$

$$\langle \mathbf{x}, \mathbf{y} \rangle = \langle T(\mathbf{x}), T(\mathbf{y}) \rangle \\ = \langle \lambda_1 \mathbf{x}, \lambda_2 \mathbf{y} \rangle = \lambda_1 \bar{\lambda}_2 \langle \mathbf{x}, \mathbf{y} \rangle. \text{ If } \langle \mathbf{x}, \mathbf{y} \rangle \neq 0,$$

then  $\lambda_1 \bar{\lambda}_2 = 1$ , but  $|\lambda_1|^2 = \lambda_1 \bar{\lambda}_1 = 1$   
 $\implies \lambda_1 = \lambda_2. \quad \square$

**Example 6.64.** orthogonal maps of  $\mathbb{R}^2$

/	rotation $R_\theta$	det = 1
\	$R_\theta \cdot \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$	det = -1

**Theorem 6.65.**  $T$  orthogonal / unitary  $V \rightarrow V$ ,  $\dim V < \infty$ ,  $W \subset V$  subspace  
 $(T(W))^\perp = T(W^\perp)$

*Proof.*  $\mathbf{x} \in (T(W))^\perp \iff \forall \mathbf{w} \in W \langle \mathbf{x}, T(\mathbf{w}) \rangle = 0$

Let  $\mathbf{y} \in V$ . Then  $T(\mathbf{y}) = \mathbf{x} \in (T(W))^\perp$

$\langle T(\mathbf{y}), T(\mathbf{w}) \rangle = 0 \quad \forall \mathbf{w} \in W$
$\iff \langle \mathbf{y}, \mathbf{w} \rangle = 0 \quad \forall \mathbf{w} \in W$
$\iff \mathbf{y} \in W^\perp$

$T^{-1}((T(W))^\perp) = W^\perp \quad \text{apply } T \quad \square$

**Corollary 6.66.**  $T : V \rightarrow V$  over  $\mathbf{F} = \mathbb{C}$ .  $\dim V = n < \infty$   
unitary  $\implies T$  diagonalizable

*Proof.* Induction over  $n = \dim V$ .

$\chi_T$  has a root  $\lambda \in \mathbb{C}$  (because  $\mathbb{C}$  is algebraically closed)

Let  $\mathbf{v} \in V$  be eigenvector to EV  $\lambda$ .

$S = \mathbf{F}\mathbf{v} = \text{span}(\{\mathbf{v}\})$ .

Then  $T(S) = S$ .

Thus  $T(S^\perp) = T(S)^\perp = S^\perp$ ,

i.e.,  $S^\perp$  is an invariant subspace of  $V$  of dimension  $n - 1$

Consider  $T' = T|_{S^\perp} : S^\perp \rightarrow S^\perp$  unitary

By induction  $T' = T|_{S^\perp}$  is diagonalizable  
 $\exists \beta' \subset S^\perp$  OB

$$\beta' = \{\mathbf{v}_1, \dots, \mathbf{v}_{n-1}\} \quad [T']_{\beta'} = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_{n-1} \end{bmatrix}.$$

Consider OB  $\beta = \{\mathbf{v}, \mathbf{v}_1, \dots, \mathbf{v}_{n-1}\}$  of  $V$ .

$$\text{Then } [T]_{\beta} = \left[ \begin{array}{c|c} \lambda & 0 \\ \hline 0 & [T']_{\beta'} \end{array} \right] = \left[ \begin{array}{c|c} \lambda & 0 \\ \hline 0 & \begin{matrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_{n-1} \end{matrix} \end{array} \right]$$

$\implies T$  diagonalizable.

□

7. CONICS (원뿔곡선)



cone  
원뿔

cone  $\cap$  plane (except if going through the vertex)  
is a circle



ellipse



parabola



쌍곡선 hyperbola



conics  
원뿔곡선

**Example 7.1.** shed light with a flashlight on a wall (all types occur)

planet orbits are ellipses

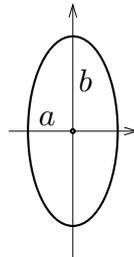
path of water in a sprinkle is a parabola

if you sharpen a pencil, you get a hyperbola

7.1. The general conic.

all  $\underset{\substack{\cap \\ \mathbb{R}^2}}{\mathbf{x}} = (x_1, x_2)$  with  $x_1^2 + x_2^2 = r^2$  form a circle of radius  $r$  at  $(0, 0)$ .

(3)  $\lambda_1, \lambda_2 > 0$   
 $\lambda_1 x_1^2 + \lambda_2 x_2^2 = c$  gives an ellipse



$$a = \sqrt{\frac{c}{\lambda_1}}$$

$$b = \sqrt{\frac{c}{\lambda_2}}$$

Let us rewrite (3) as

(4)  $\mathbf{x}^T D \mathbf{x} - c = 0$      $\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$      $D = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$

If we rotate (around the origin) the ellipse by  $\alpha$ ,  
we get that  $\tilde{\mathbf{x}} = R_{-\alpha} \mathbf{x}$  must satisfy (4).

$$(R_{-\alpha}^T = R_{\alpha} =: R)$$

$$(5) \quad \mathbf{x}^T RDR^T \mathbf{x} - c = 0$$

equation for rotated ellipse at origin

If we rotate and translate the ellipse by  $\mathbf{v}$  then

$\tilde{\mathbf{x}} = \mathbf{x} - \mathbf{v}$  must satisfy (5).

$$(\mathbf{x}^T - \mathbf{v}^T)RDR^T(\mathbf{x} - \mathbf{v}) - c = 0$$

equation for ellipse in general position

Let  $A = RDR^T$

$$\mathbf{b} = A\mathbf{v}, \quad d = \mathbf{v}^T A\mathbf{v} - c$$

$$\mathbf{x}^T A\mathbf{x} - 2\mathbf{x}^T \mathbf{b} + d = 0$$

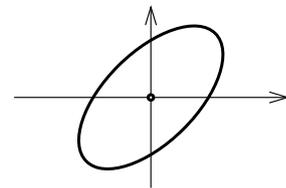
**Example 7.2.** Start with ellipse  $2x_1^2 + 4x_2^2 = 1$

$$\mathbf{x}^T \underbrace{\begin{pmatrix} 2 & 0 \\ 0 & 4 \end{pmatrix}}_D \mathbf{x} - 1 = 0 \quad (6)$$

What equation do we get when we rotate by  $\frac{\pi}{4} = 45^\circ$ ?

$$R_{\pi/4} = \begin{pmatrix} 1/\sqrt{2} & -1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \end{pmatrix} \longleftarrow \begin{pmatrix} \cos(\pi/4) & -\sin(\pi/4) \\ \sin(\pi/4) & \cos(\pi/4) \end{pmatrix} \quad \begin{array}{l} \sin(\pi/4) = \\ \cos(\pi/4) = \frac{1}{\sqrt{2}} \end{array}$$

$$A = RDR^T = \begin{pmatrix} 3 & -1 \\ -1 & 3 \end{pmatrix}$$



$$\mathbf{x}^T A\mathbf{x} - c = 0$$

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad \mathbf{x}^T A\mathbf{x} = 3x_1^2 + 3x_2^2 - 2x_1x_2$$

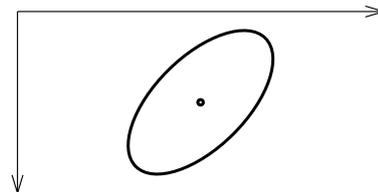
회전된

↪ equation for rotated ellipse

$$3x_1^2 + 3x_2^2 - 2x_1x_2 - 1 = 0$$

Now we translate by a vector  $\mathbf{v} = \begin{pmatrix} 2 \\ -1 \end{pmatrix}$

translated 옮겨진



$$\mathbf{b} = A\mathbf{v} = \begin{pmatrix} 3 & -1 \\ -1 & 3 \end{pmatrix} \cdot \begin{pmatrix} 2 \\ -1 \end{pmatrix} = \underline{\underline{\begin{pmatrix} 7 \\ -5 \end{pmatrix}}}$$

$$d = \begin{pmatrix} 2 & -1 \\ \uparrow v_1 & \uparrow v_2 \end{pmatrix} \begin{pmatrix} 3 & -1 \\ -1 & 3 \end{pmatrix} \begin{pmatrix} 2 \\ -1 \end{pmatrix} - 1 = 3v_1^2 + 3v_2^2 + 2 - 2v_1v_2 - 1 \\ = 3 \cdot 2^2 + 3 \cdot (-1)^2 - 2 \cdot 2 \cdot (-1) \\ = 12 + 3 + 4 - 1 = 18$$

$$\mathbf{x}^T A\mathbf{x} - 2\mathbf{x}^T \mathbf{b} + d = 0$$

$$\mathbf{x}^T \begin{pmatrix} 3 & -1 \\ -1 & 3 \end{pmatrix} \mathbf{x} - 2\mathbf{x}^T \begin{pmatrix} 7 \\ -5 \end{pmatrix} + 18 = 0 \quad \mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

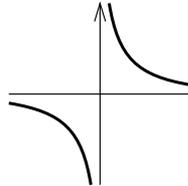
$$3x_1^2 + 3x_2^2 - 2x_1x_2 - 14x_1 + 10x_2 + 18 = 0$$

(end of Ex.)

Similar forms work for every conic.

**Example 7.3.** hyperbola

$$x_2 = \frac{1}{x_1}$$



$$x_1x_2 = 1$$

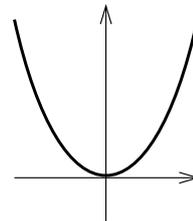
$$\mathbf{x}^T \begin{pmatrix} 0 & 1/2 \\ 1/2 & 0 \end{pmatrix} \mathbf{x} - 1 = 0 \quad \mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

**Example 7.4.** parabola

$$x_2 = x_1^2$$

$$x_1^2 - x_2 = 0$$

$$\mathbf{x}^T \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \mathbf{x} + \mathbf{x}^T \begin{pmatrix} 0 \\ -1 \end{pmatrix} = 0$$



General form of a conic

$$c_1x_1^2 + c_2x_2^2 + c_3x_1x_2 + c_4x_1 + c_5x_2 + c_6 = 0 \quad (7)$$



$$\underbrace{\mathbf{x}^T \begin{bmatrix} c_1 & 1/2c_3 \\ 1/2c_3 & c_2 \end{bmatrix} \mathbf{x}}_A - 2\mathbf{x}^T \begin{pmatrix} -1/2c_4 \\ -1/2c_5 \end{pmatrix} + c_6 = 0$$

7.2. **Analyzing conics.** Question: given (7), what type of conic is it?

first analyze  $A$

$A$  is diagonalizable; let  $\lambda_1, \lambda_2$  be eigenvalues

write  $|A| = \det A$ .

1. if  $\lambda_1, \lambda_2 < 0$  or  $\lambda_1, \lambda_2 > 0$ , then



$$\lambda_1\lambda_2 = |A| > 0$$

conic is an ellipse or circle



$$\lambda_1 \neq \lambda_2 \quad \lambda_1 = \lambda_2 \iff A \text{ scalar}$$

(elliptic case)

or it is a degenerate case: a point or empty

$$\text{e.g. } x_1^2 + x_2^2 = 0 \quad x_1^2 + x_2^2 + 1 = 0$$

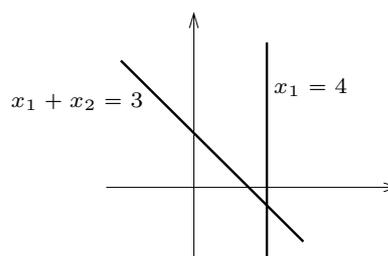
2.  $|A| = 0 \rightarrow$  w.l.o.g.  $\lambda_1 = 0$  (parabolic case)

if  $\lambda_2 = 0 \rightarrow A = 0$  degenerate (line for  $c_4 \neq 0$  or  $c_5 \neq 0$   
 $\mathbb{R}^2$  or  $\emptyset$  for  $c_4 = c_5 = 0$ )

$\lambda_2 \neq 0 \rightarrow$  conic is a parabola  
 or degenerate (2 parallel lines, 1 line or empty)

3.  $|A| < 0$  ( $\lambda_1, \lambda_2$  have opposite sign) (hyperbolic case)

퇴화  
 conic is a hyperbola or degenerate case:  
 2 lines  
 E.g.  $(x_1 + x_2 - 3)(x_1 - 4) = 0$   
 $x_1^2 + x_1x_2 - 7x_1 - 4x_2 + 12 = 0$



How do we see if some conic is degenerate?

We will check it using the center.

7.3. Position of a conic and degeneracy test. Every non-parabolic conic (incl degenerate) is obtained from one in standard form

$$\mathbf{x}^T D \mathbf{x} - c = 0 \quad \left( \text{center } \begin{pmatrix} 0 \\ 0 \end{pmatrix} \right)$$

$$D = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \quad (8) \text{ by } \begin{matrix} \text{rotation} \\ \text{translation} \end{matrix} \quad \begin{matrix} \textcircled{B} & \textcircled{A} \end{matrix}$$

Write conic in form

$$\mathbf{x}^T A \mathbf{x} - 2\mathbf{x}^T \mathbf{b} + d = 0 \quad (9)$$

$\textcircled{A}$  translation vector  $\mathbf{v}$  satisfies  $A\mathbf{v} = \mathbf{b}$   
 you know  $A$  and  $\mathbf{b}$ ; solve for  $\mathbf{v}$ . if  $|A| \neq 0$   
 $\mathbf{v} \in \mathbb{R}^2$  is center (we regard it as a point now)

Degeneracy test for non-parabolic conics ( $|A| \neq 0$ )

1) if center lies on the conic, then the conic degenerates

정칙

2) Otherwise the conic is either regular, or

$|A| > 0$  and the conic is empty.

To finish degeneracy test, we must know how to test for empty conic. We assume  $|A| > 0$ .

Evaluate the r.h.s. of (9) for  $\mathbf{x} = \mathbf{v}$

get some number  $t \in \mathbb{R}$ :

$$t = \mathbf{v}^T A \mathbf{v} - 2\mathbf{v}^T \mathbf{b} + d. \quad (\text{If } t = 0, \text{ conic is a point.})$$

$$A = \begin{pmatrix} a & c \\ c & b \end{pmatrix} \quad \left. \begin{array}{l} \text{If } \lambda_1, \lambda_2 > 0 \text{ (} \iff a > 0 \text{) and } t > 0, \text{ or} \\ \lambda_1, \lambda_2 < 0 \text{ (} \iff a < 0 \text{) and } t < 0 \end{array} \right\} \iff a \cdot t > 0$$

then conic is empty, otherwise it is a circle or ellipse (regular).

In that case, up to translation and rotation we have the basic form (3) with  $c = -t$ , thus

$$a = \sqrt{-\frac{t}{\lambda_1}}, \quad b = \sqrt{-\frac{t}{\lambda_2}}, \quad \text{area} = \pi \cdot a \cdot b = \frac{|t|\pi}{\sqrt{\lambda_1 \lambda_2}} = \frac{|t|\pi}{\sqrt{|A|}}$$

**Example 7.5.**  $x_1^2 + x_1 x_2 - 7x_1 - 4x_2 + 12 = 0$  (from p.19)

$$\mathbf{x}^T \begin{pmatrix} 1 & 1/2 \\ 1/2 & 0 \end{pmatrix} \mathbf{x} - 2\mathbf{x}^T \begin{pmatrix} 7/2 \\ 2 \end{pmatrix} + 12 = 0$$

$A\mathbf{v} = \mathbf{b}$

can see it from the picture on p.19

$$\begin{pmatrix} 1 & 1/2 \\ 1/2 & 0 \end{pmatrix} \mathbf{v} = \begin{pmatrix} 7/2 \\ 2 \end{pmatrix} \quad \longrightarrow \quad \mathbf{v} = \begin{pmatrix} 4 \\ -1 \end{pmatrix}$$

put into the conic equation  $\mathbf{x} = \begin{pmatrix} 4 \\ -1 \end{pmatrix}$   $x_1 = 4, \quad x_2 = -1$

$$4^2 + 4 \cdot (-1) - 7 \cdot 4 - 4 \cdot (-1) + 12 = 16 - 4 - 28 + 4 + 12 = 0$$

center on conic  
 $\rightsquigarrow$  conic degenerate

**Example 7.6.** Let's test this on a complicated empty conic

$$\underbrace{(x_1 + x_2 - 3)^2}_{[1]} + \underbrace{(x_1 - 4)^2}_{[2]} + 1 = 0 \quad (10)$$

$$2x_1^2 + 2x_1 x_2 + x_2^2 - 14x_1 - 6x_2 + 26 = 0 \quad (11)$$

Now forget first equation; how do you see conic is empty?

$$\mathbf{x}^T \underbrace{\begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}}_A \mathbf{x} - 2\mathbf{x}^T \underbrace{\begin{pmatrix} 7 \\ 3 \end{pmatrix}}_b + 26 = 0$$

Let's first check  $|A| = 2 \cdot 1 - 1 \cdot 1 = 1 > 0$  (ok)

$$\text{Solve} \quad \underbrace{\begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}}_A \mathbf{v} = \underbrace{\begin{pmatrix} 7 \\ 3 \end{pmatrix}}_b \quad \longrightarrow \quad \mathbf{v} = \begin{pmatrix} 4 \\ -1 \end{pmatrix}$$

Test l.h.s. of (11) for  $\mathbf{x} = \mathbf{v} = \begin{pmatrix} 4 \\ -1 \end{pmatrix}$  can see this also in (10) because the two lines given by  $[1] = 0$  and  $[2] = 0$  intersect at  $(4, -1)$

$$x_1 = 4, \quad x_2 = -1$$

$$t = 2 \cdot 4^2 + 2 \cdot 4 \cdot (-1) + (-1)^2 - 14 \cdot 4 - 6 \cdot (-1) + 26 = \begin{matrix} \text{clear from} \\ (10) \\ \downarrow \\ 1 \end{matrix}$$

$$32 \quad -8 \quad +1 \quad -56 \quad +6 \quad +26 = 1$$

$$A = \begin{pmatrix} a & c \\ c & b \end{pmatrix} = \begin{pmatrix} \textcircled{2} & 1 \\ 1 & 1 \end{pmatrix} \quad a = 2 > 0$$

$$a \cdot t = 2 \cdot 1 > 0 \rightsquigarrow \text{conic is empty} \quad (\text{End Example 7.6})$$

(B) To determine rotation, we need to find eigenvectors of  $A$  in (9) (rotation does not depend on  $\mathbf{b}$ ,  $d$ )

- 1) If EV  $\lambda_1 = \lambda_2$  (and conic is not degenerate) then conic is a circle, and rotation does not matter

$$\lambda_1 = \lambda_2 (=:\lambda) \iff A = \lambda \cdot Id = \begin{pmatrix} \lambda & 0 \\ 0 & \lambda \end{pmatrix} \text{ scalar}$$

- 2) So assume  $\lambda_1 \neq \lambda_2$

$$\|\mathbf{v}_i\| = 1$$

Let  $\mathbf{v}_1, \mathbf{v}_2$  be normalized eigenvectors to  $\lambda_1, \lambda_2$ .

$$A\mathbf{v}_i = \lambda_i\mathbf{v}_i. \text{ We know } \mathbf{v}_1^T \mathbf{v}_2 = 0 \quad \|\mathbf{v}_i\| = 1 \text{ determines } \mathbf{v}_i \text{ up to } \pm 1$$

$$\text{as before left } R = \begin{pmatrix} \mathbf{v}_1 & \mathbf{v}_2 \end{pmatrix}.$$

We know  $R$  is orthogonal. Thus  $|R| = 1$  or  $|R| = -1$ . If  $|R| = -1$ , replace  $\mathbf{v}_1$  by  $-\mathbf{v}_1$ .

Thus assume  $|R| = 1 \rightsquigarrow R = R_{\alpha}$  (rotation matrix).

We have

$$AR = R\Lambda$$

$$A = R\Lambda R^T \rightsquigarrow R \text{ is the applied rotation}$$

compare to

(5) on p.17 with  $D = \Lambda$

$$\Lambda = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$$

$\alpha$  is the angle of rotation

(up to multiples of  $180^\circ$ )

**Example 7.7.** We consider the matrix  $A = \begin{pmatrix} 3 & -1 \\ -1 & 3 \end{pmatrix}$  of p.17.

$$\det(A - \lambda Id) = (3 - \lambda)^2 - 1 = \lambda^2 - 6\lambda + 8$$

$$\lambda_{1,2} = 3 \pm \sqrt{9 - 8} \quad \lambda_1 = 4 \quad \lambda_2 = 2$$

$$\underline{\lambda_1 = 4} \quad A_1 = A - \lambda_1 Id = \begin{pmatrix} -1 & -1 \\ -1 & -1 \end{pmatrix}$$

$$A_1 \mathbf{v}_1 = \mathbf{0} \quad \mathbf{v}_1 = \begin{pmatrix} v_{1,1} \\ v_{1,2} \end{pmatrix} \quad -v_{1,1} - v_{1,2} = 0 \quad \longrightarrow$$

$$v_{1,2} = t, v_{1,1} = -t \quad \longrightarrow \mathbf{v} = t \begin{pmatrix} -1 \\ +1 \end{pmatrix}$$

Determine  $t$  using  $\|\mathbf{v}_1\| = |t| \cdot \sqrt{2} = 1 \rightarrow t = \frac{1}{\sqrt{2}}$

$$\mathbf{v}_1 = \textcircled{\pm} \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

$$\underline{\lambda_2 = 2} \quad A_2 = A - \lambda_2 Id = \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix}$$

$$A_2 \mathbf{v}_2 = \mathbf{0} \quad \mathbf{v}_2 = t \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad \|\mathbf{v}_2\| = |t| \cdot \sqrt{2} = 1$$

$$\rightarrow \mathbf{v}_2 = \textcircled{\pm} \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

choose first '+' in both  $\textcircled{\pm}$ :

$$R = \begin{pmatrix} \mathbf{v}_1 & \mathbf{v}_2 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} -1 & 1 \\ +1 & 1 \end{pmatrix} \quad |R| = \frac{1}{2} \cdot (-2) = -1$$

change sign of  $\mathbf{v}_1 \rightarrow \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix} \rightarrow$  thus now sign correct

$$R = R_\alpha = \begin{pmatrix} \cos \alpha & -\sin \alpha \\ \sin \alpha & \cos \alpha \end{pmatrix} \quad \cos \alpha = \frac{1}{\sqrt{2}}$$

$$\sin \alpha = -\frac{1}{\sqrt{2}}$$

$$\alpha = -45^\circ$$

Why did we get  $-45^\circ$  and not  $45^\circ$ , as we did in example 7.2?  
 Because we used different order of EV  $\lambda_i$ . With (6), we put  
 in (8)  $\lambda_1 = 2, \lambda_2 = 4$ .  
 If we use order  $\lambda_1 = 2, \lambda_2 = 4$  of p.17 here, we have  $R = \begin{pmatrix} \mathbf{v}_2 & \mathbf{v}_1 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}$   
 $\rightsquigarrow \alpha = 45^\circ$ , as in that  
 example

determining (symmetry) axes (except for circle) and asymptotes of hyperbola

symmetry axes =  $\mathbf{v} +$  multiples of eigenvectors of  $A$  (= cols of  $R$ ) =  
 $= \mathbf{v} + t_1 \mathbf{v}_1$ , and  $\mathbf{v} + t_2 \mathbf{v}_2$  (for  $t_{1,2} \in \mathbb{R}$  independent variables)

$$\text{asymptotes (of hyperbola)} = \mathbf{v} + t_{\pm} \cdot \underbrace{R \cdot \begin{pmatrix} \pm\sqrt{|\lambda_2|} \\ \sqrt{|\lambda_1|} \end{pmatrix}}_{\mathbf{v}_{\pm}}$$

$$(t_+, t_- \in \mathbb{R} \text{ independent, } D = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \text{ with } \lambda_1 \lambda_2 < 0).$$

### Example

$$x_1^2 + x_1 x_2 - 7x_1 - 4x_2 + 12 = 0 \text{ (from p.19)}$$

$$\mathbf{x}^T \begin{pmatrix} 1 & 1/2 \\ 1/2 & 0 \end{pmatrix} \mathbf{x} - 2\mathbf{x}^T \begin{pmatrix} 7/2 \\ 2 \end{pmatrix} \dots$$

$$\mathbf{v} = \begin{pmatrix} 4 \\ -1 \end{pmatrix}$$

$$\lambda^2 - \lambda - 1/4 = 0 \quad \lambda = \frac{1}{2} \pm \sqrt{\frac{1}{4} + \frac{1}{4}} = \frac{1 \pm \sqrt{2}}{2}$$

In calculating asymptote vecs can use multiples, so write  $\mathbf{w} \doteq \mathbf{u}$  for  $\mathbf{w} = c \cdot \mathbf{u}$ ,  $c \neq 0$ .

$$\lambda_1 = \frac{1 + \sqrt{2}}{2} \quad A - \lambda_1 Id = \begin{pmatrix} \frac{1 - \sqrt{2}}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{-1 - \sqrt{2}}{2} \end{pmatrix} \rightsquigarrow \mathbf{v}_1 \doteq \begin{pmatrix} 1 \\ -1 + \sqrt{2} \end{pmatrix}$$

$$\lambda_2 = \frac{1 - \sqrt{2}}{2} \rightsquigarrow \mathbf{v}_2 \doteq \begin{pmatrix} 1 - \sqrt{2} \\ 1 \end{pmatrix} \text{ (by Lity } \because \lambda_1 \neq \lambda_2 \text{ no circle!)}$$

$$\frac{1}{\sqrt{4 - \sqrt{8}}} \begin{pmatrix} 1 & 1 - \sqrt{2} \\ -1 + \sqrt{2} & 1 \end{pmatrix} = R$$

$$\mathbf{v}_{\pm} \doteq \frac{1}{\sqrt{4 - \sqrt{8}}} \begin{pmatrix} 1 & 1 - \sqrt{2} \\ -1 + \sqrt{2} & 1 \end{pmatrix} \cdot \begin{pmatrix} \pm\sqrt{\frac{\sqrt{2} - 1}{2}} \\ \sqrt{\frac{\sqrt{2} + 1}{2}} \end{pmatrix} \doteq \begin{pmatrix} 1 & 1 - \sqrt{2} \\ -1 + \sqrt{2} & 1 \end{pmatrix} \cdot \begin{pmatrix} \pm 1 \\ \sqrt{2} + 1 \end{pmatrix}$$

$$= \begin{pmatrix} \pm 1 - 1 \\ \mp 1 \pm \sqrt{2} + \sqrt{2} + 1 \end{pmatrix} = \begin{pmatrix} -2 \\ 2 \end{pmatrix}, \begin{pmatrix} 0 \\ 2\sqrt{2} \end{pmatrix} \doteq \begin{pmatrix} -1 \\ 1 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix},$$

$$\text{giving } \begin{pmatrix} 4 \\ -1 \end{pmatrix} + \tilde{t}_- \cdot \begin{pmatrix} -1 \\ 1 \end{pmatrix}, \quad \begin{pmatrix} 4 \\ -1 \end{pmatrix} + \tilde{t}_+ \cdot \begin{pmatrix} 0 \\ 1 \end{pmatrix} \quad (\tilde{t}_{\pm} \doteq t_{\pm}),$$

the lines given by the two factors  $(x_1 + x_2 - 3)(x_1 - 4) = 0$

(the hyp. case degenerates into its asymptotes)

### Degeneracy test for parabolic conics ( $|A| = 0$ )

We consider now (7)

$$(7) \quad c_1 x_1^2 + c_2 x_2^2 + c_3 x_1 x_2 + c_4 x_1 + c_5 x_2 + c_6 = 0$$

and we assume that

$$A = \begin{pmatrix} c_1 & c_3/2 \\ c_3/2 & c_2 \end{pmatrix} \quad \text{has } |A| = 0, \quad \text{but } A \neq 0$$

This parabolic case determines either

a parabola	(regular case)	
2 parallel lines	}	degenerate cases
1 line		
$\emptyset$		

$$|A| = c_1 c_2 - \frac{c_3^2}{4} = 0 \quad \longrightarrow \quad c_1 c_2 \geq 0 \rightarrow c_1, c_2 \text{ have same sign (or one is 0)}$$

w.l.o.g. multiply (7) by  $-1$  so that  $c_1, c_2 \geq 0$ .

$$\text{Let } \varepsilon = \begin{cases} \text{sgn}(c_3) & \text{if } c_3 \neq 0 \\ \text{or } 1 & \text{if } c_3 = 0 \end{cases} \implies c_3 = 2\varepsilon \cdot \sqrt{c_1 c_2}$$

$$\text{Then } c_1 x_1^2 + c_2 x_2^2 + c_3 x_1 x_2 = (\sqrt{c_1} x_1 + \varepsilon \sqrt{c_2} x_2)^2.$$

$$\text{Degeneracy test: } c_4 \varepsilon \sqrt{c_2} = \sqrt{c_1} c_5 ? \iff \begin{matrix} c_4 x_1 + c_5 x_2 \\ \text{multiple of} \\ \sqrt{c_1} x_1 + \varepsilon \sqrt{c_2} x_2 \end{matrix}$$

If no  $\rightarrow$  conic is parabola (regular)

if yes, let  $h$  be so that

$$h \cdot (\sqrt{c_1} x_1 + \varepsilon \sqrt{c_2} x_2) = c_4 x_1 + c_5 x_2 \quad \left( \begin{matrix} h = \frac{c_4}{\sqrt{c_1}} \text{ or } h = \frac{c_5}{\varepsilon \sqrt{c_2}} \\ \text{whatever makes sense} \end{matrix} \right)$$

Test now the solvability (and number of solutions) of the equation

$$y^2 + hy + c_6 = 0$$

(Here  $y = \sqrt{c_1} x_1 + \varepsilon \sqrt{c_2} x_2$ .)

$$\begin{aligned} \frac{h^2}{4} &> c_6 && \text{two solutions} \longrightarrow \text{two parallel lines} \\ \frac{h^2}{4} &= c_6 && \text{one solution} \longrightarrow \text{one line} \\ \frac{h^2}{4} &< c_6 && \text{no solution} \longrightarrow \text{empty} \end{aligned}$$

**Example 7.8.**  $-4x_1^2 - x_2^2 - 4x_1 x_2 + 10x_1 + 5x_2 - 6 = 0$

$$A = \begin{pmatrix} -4 & -2 \\ -2 & -1 \end{pmatrix} \quad |A| = 0$$

here  $c_1, c_2 < 0$  multiply by  $-1$

$$4x_1^2 + x_2^2 + 4x_1 x_2 - 10x_1 - 5x_2 + 6 = 0$$

$$A = \begin{pmatrix} 4 & 2 \\ 2 & 1 \end{pmatrix} \quad |A| = 0$$

$$\varepsilon = \text{sgn}(c_3) = 1$$

$$\sqrt{c_1} = 2, \sqrt{c_2} = 1 \quad \longrightarrow \quad 4x_1^2 + x_2^2 + 4x_1x_2 = (2x_1 + x_2)^2$$

$$\text{now apply degeneracy test:} \quad \begin{array}{ccc} -10 \cdot 1 \cdot 1 & = & 2 \cdot (-5) \quad ? \\ c_4 \quad \varepsilon \quad \sqrt{c_2} & & \sqrt{c_1} \quad c_5 \end{array}$$

conic  
yes  $\rightarrow$  degenerate

$$h = \frac{c_4}{\sqrt{c_1}} = -\frac{10}{2} = -5 \quad (\text{ok since } \sqrt{c_1} \neq 0)$$

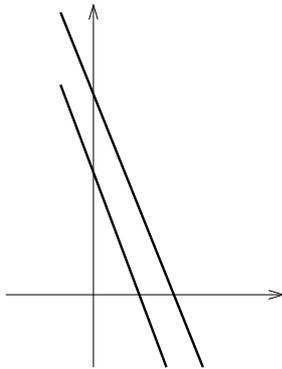
$$\begin{aligned} \text{Thus} \quad & 4x_1^2 + x_2^2 + 4x_1x_2 - 10x_1 - 5x_2 + 6 \\ & = y^2 - 5y + 6 \quad (\text{with } y = 2x_1 + x_2) \end{aligned}$$

⇐

Solutions of  $y^2 - 5y + 6 = 0$  are  $y = 2$  and  $y = 3$ .

(2 solutions because  $6 < (\frac{5}{2})^2$ )

Thus we have 2 parallel lines  $2x_1 + x_2 = 2$  and  $2x_1 + x_2 = 3$



**Example 7.9.**  $-4x_1^2 - x_2^2 - 4x_1x_2 + 10x_1 + 5x_2 - 6.25 = 0$

repeat calculation of previous example until

$$0 = y^2 - 5y + 6.25 \quad (\text{with } y = 2x_1 + x_2)$$

$$6.25 = \left(\frac{5}{2}\right)^2 \quad \longrightarrow \quad \text{one solution } y = 2.5$$

$$\text{conic} = \underline{\text{1 line}} \quad 2x_1 + x_2 = 2.5$$

**Example 7.10.**  $-4x_1^2 - x_2^2 + 4x_1x_2 - 5x_1 - 6x_2 + 6$

$$\rightarrow \quad 4x_1^2 + x_2^2 - 4x_1x_2 + 5x_1 + 6x_2 - 6$$

$$A = \begin{pmatrix} 4 & -2 \\ -2 & 1 \end{pmatrix} \quad \varepsilon = \text{sgn}(-2) = -1$$

$$\sqrt{c_1} = 2 \quad \sqrt{c_2} = 1$$

degeneracy test:

$$\begin{array}{ccc} 5 \cdot (-1) \cdot 1 & = & 6 \cdot 2 \\ c_4 \quad \varepsilon \quad \sqrt{c_2} & = & c_5 \quad \sqrt{c_1} \end{array} \quad \longrightarrow \quad \text{no} \quad \longrightarrow \quad \underline{\text{parabola}} \\ \text{(regular)}$$

**Example 7.11.**  $4x_2^2 - 4x_2 + 2 = 0$

$$A = \begin{pmatrix} 0 & 0 \\ 0 & 4 \end{pmatrix} \quad (c_1 = 0, c_2 \geq 0 \rightarrow \text{no need to change sign})$$

$$\text{sgn}(0) = 0 \rightarrow \varepsilon = 1$$

$$\sqrt{c_1} = 0 \quad \sqrt{c_2} = 2$$

$$c_4 = 0 \quad c_5 = -4$$

degeneracy test:

$$0 \cdot 1 \cdot 2 = -4 \cdot 0 \rightarrow \text{yes degenerate}$$

$$c_4 \quad \varepsilon \quad \sqrt{c_2} \quad c_5 \quad \sqrt{c_1}$$

$$h = \frac{c_4}{\sqrt{c_1}} = \frac{0}{0} \quad \times \quad h = \frac{c_5}{\varepsilon \sqrt{c_2}} = \frac{-4}{2} = -2$$

$$\underbrace{y^2 - 2y + 2}_{+hy} \quad \underbrace{\quad}_{c_6} \quad \left( y = \underbrace{\sqrt{c_1}}_0 x_1 + \varepsilon \underbrace{\sqrt{c_2}}_1 x_2 = x_2 \right)$$

$$c_6 > \frac{h^2}{4} \rightarrow \text{no solution } y \rightarrow \underline{\text{conic} = \emptyset}$$

determining vertex and (symmetry) axis of parabola

$$\mathbf{x}^T RDR^T \mathbf{x} - 2\mathbf{x}^T \mathbf{b} + d = 0$$

$$\begin{array}{c} R^T \\ \parallel \\ A = RDR^{-1} \end{array}$$

$$\begin{array}{c} RD = AR \\ \uparrow \\ \text{cols} = \text{eigenvecs of } A \end{array}$$

first col include Evec to  $\lambda \neq 0$ .

$$\text{So } D = \begin{pmatrix} \lambda & 0 \\ 0 & 0 \end{pmatrix}.$$

Let  $\mathbf{x} = R\mathbf{y}$ . Then

$$\underbrace{\mathbf{y}^T D \mathbf{y}}_{\lambda y_1^2} - 2\mathbf{y}^T R^T \mathbf{b} + d = 0$$

Solve for  $y_2$  and find vertex  $\begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$  and axis  $\begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} \cdot t$ .

(If you want vector pointing inside parabola,  $\dots + \begin{pmatrix} 0 \\ \text{sgn}(z_2 \cdot \lambda) \end{pmatrix} \cdot t$  for  $\begin{pmatrix} z_1 \\ z_2 \end{pmatrix} = R^T \mathbf{b}$ )

$$\text{Then orig vertex is } \mathbf{x} = R \cdot \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} \text{ and axis } R \cdot \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \overbrace{R \cdot \begin{pmatrix} 0 \\ 1 \end{pmatrix}}^{\text{Evec to } \lambda = 0} \cdot t.$$

Example

$$x_1^2 + 4x_1x_2 + 4x_2^2 + 5x_2 - 1 = 0.$$

$$A = \begin{pmatrix} 1 & 2 \\ 2 & 4 \end{pmatrix}, \quad \det A = 0.$$

$$(\lambda - 1)(\lambda - 4) = 4 \Rightarrow \lambda = 0, 5 \text{ EV.}$$

$$\lambda = 0 \quad \text{Evec} \begin{pmatrix} 2 \\ -1 \end{pmatrix} \cdot t$$

$$\lambda = 5 \quad \text{Evec} \begin{pmatrix} 1 \\ 2 \end{pmatrix} \cdot t \quad (\text{test: Evec} \perp)$$

$$R = \frac{1}{\sqrt{5}} \begin{pmatrix} 1 & 2 \\ 2 & -1 \end{pmatrix} \stackrel{\det=-1}{\rightsquigarrow} \frac{1}{\sqrt{5}} \begin{pmatrix} 1 & -2 \\ 2 & 1 \end{pmatrix} = R_\theta = R$$

$$\theta = \arccos\left(\frac{1}{\sqrt{5}}\right)$$

$$= \arctan(2)$$

$$\frac{1}{\sqrt{5}} \begin{pmatrix} 1 & 2 \\ -2 & 1 \end{pmatrix} = R^T = R^{-1}.$$

$$\text{rotate: } \mathbf{x} = R\mathbf{y} \qquad R^T \mathbf{b} = \frac{1}{\sqrt{5}} \begin{pmatrix} 1 & 2 \\ -2 & 1 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ -5/2 \end{pmatrix} = \frac{1}{\sqrt{5}} \begin{pmatrix} -5 \\ -5/2 \end{pmatrix}$$

$$\mathbf{y}^T \begin{pmatrix} 5 & 0 \\ 0 & 0 \end{pmatrix} \mathbf{y} - \frac{1}{\sqrt{5}} \mathbf{y}^T \begin{pmatrix} -10 \\ -5 \end{pmatrix} - 1 = 0.$$

$$5y_1^2 + \frac{10}{\sqrt{5}}y_1 + \frac{5}{\sqrt{5}}y_2 - 1 = 0.$$

$$y_2 = - \left( \sqrt{5}y_1^2 + 2y_1 - \frac{1}{\sqrt{5}} \right) \quad (\text{can solve for } y_2 \rightsquigarrow \text{is regular parabola})$$

$$- \left( \sqrt{5} \left( y_1 + \frac{1}{\sqrt{5}} \right)^2 - \frac{2}{\sqrt{5}} \right)$$

$$\text{vrt/axis} \left( -\frac{1}{\sqrt{5}}, +\frac{2}{\sqrt{5}} \right) + (0, 1) \cdot t$$

rotate back:

$$R \cdot \left[ \left( -\frac{1}{\sqrt{5}}, +\frac{2}{\sqrt{5}} \right) + (0, 1) \cdot t \right] = \frac{1}{5} \begin{pmatrix} 1 & -2 \\ 2 & 1 \end{pmatrix} \begin{pmatrix} -1 \\ 2 \end{pmatrix} + \begin{pmatrix} -2 \\ 1 \end{pmatrix} \cdot \tilde{t} \quad (\tilde{t} = t/\sqrt{5})$$

$$\frac{1}{5} \begin{pmatrix} -5 \\ 0 \end{pmatrix} + \begin{pmatrix} -2 \\ 1 \end{pmatrix} \cdot \tilde{t}$$

$$\begin{pmatrix} -1 \\ 0 \end{pmatrix} + \begin{pmatrix} -2 \\ 1 \end{pmatrix} \cdot \tilde{t}$$

