

Week 10: Unitary/Orthogonal Matrices, Bilinear Forms (textbook § 6.5, 6.8)Spectral Theorems Revisited

Recall: $Q \in M_{n \times n}(\mathbb{F})$ is

unitary / orthogonal
 $(\mathbb{F} = \mathbb{C}) \quad (\mathbb{F} = \mathbb{R})$

\iff

$$QQ^* = Q^*Q = I$$

These matrices arise naturally as change of coordinate matrices.

Lemma: Let β and γ be orthonormal bases of a finite dim'l inner product space $(V, \langle \cdot, \cdot \rangle)$. Then, the change of coordinate matrix $Q = [I]_{\beta}^{\gamma}$ is unitary/orthogonal.
 $(\mathbb{F} = \mathbb{C}) \quad (\mathbb{F} = \mathbb{R})$

"Proof:" We "explain" the proof by the example that

$$V = \mathbb{R}^n, \quad \gamma = \text{standard basis}$$

$$\beta = \{v_1, v_2, \dots, v_n\} \text{ O.N.B.}$$

In standard coordinates,

$$v_1 = \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{n1} \end{pmatrix}, \quad v_2 = \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{n2} \end{pmatrix}, \dots, \quad v_n = \begin{pmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{nn} \end{pmatrix}$$

then the change of basis matrix is

$$Q = [I]_{\beta}^{\gamma} = \left(\begin{array}{c|c|c|c} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{array} \right) \begin{matrix} \parallel & \parallel & \parallel \\ v_1 & v_2 & v_n \end{matrix}$$

Taking transpose,

$$Q^t = \begin{pmatrix} a_{11} & a_{21} & \dots & a_{n1} \\ a_{12} & a_{22} & \dots & a_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & \dots & a_{nn} \end{pmatrix} = \begin{matrix} v_1^t \\ v_2^t \\ \vdots \\ v_n^t \end{matrix}$$

Question: When is $Q^t Q = I$?

$$Q^t Q = \begin{pmatrix} a_{11} & a_{21} & \dots & a_{n1} \\ a_{12} & a_{22} & \dots & a_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & \dots & a_{nn} \end{pmatrix} \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix}$$

$$= \begin{pmatrix} \langle v_1, v_1 \rangle & \langle v_1, v_2 \rangle & \dots & \langle v_1, v_n \rangle \\ \langle v_2, v_1 \rangle & \langle v_2, v_2 \rangle & \dots & \langle v_2, v_n \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle v_n, v_1 \rangle & \langle v_n, v_2 \rangle & \dots & \langle v_n, v_n \rangle \end{pmatrix} \stackrel{\text{if } \beta \text{ is orthonormal}}{=} \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{pmatrix} = I$$

$\beta = \{v_1, v_2, \dots, v_n\}$
is orthonormal!!

Corollary: Suppose β, γ are O.N.B. for $(V, \langle \cdot, \cdot \rangle)$, and $T: V \rightarrow V$.

$$[T]_\beta = Q^{-1} [T]_\gamma Q = Q^* [T]_\gamma Q$$

where Q is unitary ($F = \mathbb{C}$) or orthogonal ($F = \mathbb{R}$).

Defⁿ: Two matrices $A, B \in M_{n \times n}(F)$ are $\underset{(F=\mathbb{C})}{\text{unitarily}}$ / $\underset{(F=\mathbb{R})}{\text{orthogonally}}$ equivalent if there exists a $\underset{(F=\mathbb{C})}{\text{unitary}}$ / $\underset{(F=\mathbb{R})}{\text{orthogonal}}$ matrix Q

s.t.

$$A = Q^* B Q$$

Therefore, A, B are unitarily/orthogonally equivalent iff they represent the same linear operator T under different orthonormal bases!

As a result, we can restate the Spectral Theorems and Schur's Lemma in matrix form.

Spectral Theorems: (Matrix Form)

$$(\mathbb{F} = \mathbb{C}) \quad (\mathbb{F} = \mathbb{R})$$

Let $A \in M_{n \times n}(\mathbb{F})$. Then, A is unitarily/orthogonally equivalent to a diagonal matrix if and only if A is normal ($\mathbb{F} = \mathbb{C}$) or self-adjoint ($\mathbb{F} = \mathbb{R}$).

Schur's Lemma: (Matrix Form)

Let $A \in M_{n \times n}(\mathbb{F})$. If the characteristic polynomial of A splits over \mathbb{F} , then A is $\begin{cases} \text{unitarily} & (\mathbb{F} = \mathbb{C}) \\ \text{orthogonally} & (\mathbb{F} = \mathbb{R}) \end{cases}$ equivalent to an upper triangular matrix.

Example: Let

$$A = \begin{pmatrix} 4 & 2 & 2 \\ 2 & 4 & 2 \\ 2 & 2 & 4 \end{pmatrix} \in M_{3 \times 3}(\mathbb{R})$$

$$\boxed{\mathbb{F} = \mathbb{R}}$$

Find an orthogonal matrix P st. $P^t A P$ is a diagonal matrix.

Solution: First of all, $A^t = A$ (i.e. A is self adjoint), so such a P exists. Finding P is equivalent to finding an orthonormal eigenbasis for A .

Finding eigenvalues: $\det(A - \lambda I) = 0$.

$$\begin{aligned}
 0 &= \det \begin{pmatrix} 4-\lambda & 2 & 2 \\ 2 & 4-\lambda & 2 \\ 2 & 2 & 4-\lambda \end{pmatrix} = (4-\lambda) \det \begin{pmatrix} 4-\lambda & 2 \\ 2 & 4-\lambda \end{pmatrix} - 2 \cdot \det \begin{pmatrix} 2 & 2 \\ 2 & 4-\lambda \end{pmatrix} \\
 &\quad + 2 \cdot \det \begin{pmatrix} 2 & 4-\lambda \\ 2 & 2 \end{pmatrix} \\
 &= (4-\lambda) [(4-\lambda)^2 - 4] - 2 [2(4-\lambda) - 4] \\
 &\quad + 2 [4 - 2(4-\lambda)] \\
 &= (4-\lambda)(\lambda^2 - 8\lambda + 12) - 8(2-\lambda) \\
 &\stackrel{(?)}{=} -(\lambda-4)(\lambda-6)(\lambda-2) + 8(\lambda-2) \\
 &= (\lambda-2)[-(\lambda^2 - 10\lambda + 24) + 8] \\
 &= (\lambda-2)[-(\lambda^2 - 10\lambda + 16)] \\
 &= -(\lambda-2)^2(\lambda-8).
 \end{aligned}$$

The eigenvalues and multiplicities are

$$\begin{array}{ll}
 \lambda_1 = 2, & m_1 = 2 \\
 \lambda_2 = 8, & m_2 = 1
 \end{array}$$

Finding orthonormal eigenvectors:

$$\boxed{\lambda_1 = 2} \quad E_{\lambda_1} = N(A - 2I) = N \begin{pmatrix} 2 & 2 & 2 \\ 2 & 2 & 2 \\ 2 & 2 & 2 \end{pmatrix} = \text{span} \left\{ \underbrace{\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix}}_{\text{NOT orthonormal!}} \right\}$$

extra step

Apply Gram-Schmidt!

$$\begin{aligned}
 v_1 &= \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \\
 v_2 &= \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix} - \frac{\begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}}{\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix} - \frac{1}{2} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} = \begin{pmatrix} -\frac{1}{2} \\ 1 \\ -\frac{1}{2} \end{pmatrix}
 \end{aligned}$$

Hence, $\left\{ \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \frac{1}{\sqrt{2}} \begin{pmatrix} -1 \\ 1 \\ -1 \end{pmatrix} \right\}$ is an O.N.B. for E_{λ_1} .

$$\boxed{\lambda_2 = 8} \quad E_{\lambda_2} = N(A - 8I) = N \begin{pmatrix} -4 & 2 & 2 \\ 2 & -4 & 2 \\ 2 & 2 & -4 \end{pmatrix} = \text{span} \left\{ \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \right\}.$$

Hence, $\left\{ \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \right\}$ is an O.N.B. for E_{λ_2} .

Putting all these together,

$\left\{ \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \frac{1}{\sqrt{6}} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix}, \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \right\}$ is an orthonormal eigenbasis for \mathbb{R}^3 .

The required orthogonal matrix P is thus given by

$$P = \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} \\ 0 & \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} \end{pmatrix}.$$

and $P^t A P = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 8 \end{pmatrix}$.

Midterm 2 up to here □

Bilinear Forms

Recall that an inner product was a 2-variable scalar-valued function

$$\langle \cdot, \cdot \rangle : V \times V \longrightarrow \mathbb{F}$$

which satisfies some properties (Ex: what are they?).

We know that $\langle \cdot, \cdot \rangle$ is linear in the 1st variable but only conjugate linear in the 2nd variable! (Note: There is no such different though when $\mathbb{F} = \mathbb{R}$.)

Now, we study 2-variable functions which are more "symmetric" in the two variables.

Defⁿ: Let V be a vector space over \mathbb{F} .

A bilinear form is a function

$$H: V \times V \longrightarrow \mathbb{F}$$

which is linear in each variable:

$$H(a_1x_1 + a_2x_2, y) = a_1 H(x_1, y) + a_2 H(x_2, y)$$

$$H(x, a_1y_1 + a_2y_2) = a_1 H(x, y_1) + a_2 H(x, y_2)$$

for all $a_1, a_2 \in \mathbb{F}$, $x_1, x_2, x, y_1, y_2, y \in V$.

Remark: Even though many interesting applications are concerned with general fields \mathbb{F} (e.g. \mathbb{Z}_2), for simplicity, we will mainly be focusing on the case $\mathbb{F} = \mathbb{C}$ or \mathbb{R} .

Example: If $\mathbb{F} = \mathbb{R}$, any inner product \langle , \rangle is a bilinear form.

Example: Given any $A \in M_{n \times n}(\mathbb{R})$, the function

$$H: \mathbb{R}^n \times \mathbb{R}^n \longrightarrow \mathbb{R}$$

defined by $H(x, y) = x^t A y$ is a bilinear form.

$$\begin{aligned} \text{e.g. } H\left(\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}\right) &= (x_1 \ x_2) \begin{pmatrix} 2 & 3 \\ 4 & -1 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} \\ &= 2x_1y_1 + 3x_1y_2 + 4x_2y_1 - x_2y_2 \end{aligned}$$

Given any matrix $A \in M_{n \times n}(\mathbb{F})$, I can associate to it a bilinear form H as above. Does all bilinear forms come from some matrix?



The Space of Bilinear Forms $B(V)$

Question: What can we do with bilinear forms?

- We can add two bilinear forms:

$$(H_1 + H_2)(x, y) := H_1(x, y) + H_2(x, y)$$

- We can scalar-multiply:

$$(\lambda H)(x, y) := \lambda H(x, y)$$

FACT: The space of bilinear forms $B(V)$ on a given vector space V forms a vector space (over the same field \mathbb{F} as V).

Question: What is $\dim B(V)$?

Answer: $\dim B(V) = n^2$ if $\dim_{\mathbb{F}} V = n$.

In fact, there exists a vector space isomorphism

$$B(V) \xrightarrow[\text{not canonical}]{\cong} M_{n \times n}(\mathbb{F}) \quad \text{if } \dim_{\mathbb{F}} V = n$$

! Does it look familiar? !

$$L(V) := \{T: V \rightarrow V \text{ linear}\} \xrightarrow[\text{not canonical}]{\cong} M_{n \times n}(\mathbb{F})$$

We have such a correspondence once we choose a basis β

(hence not entirely "natural")

Let $\beta = \{v_1, v_2, \dots, v_n\}$ be a basis for V .

Using bilinearity.

$$H\left(\underbrace{\sum_{i=1}^n a_i v_i}_{x}, \underbrace{\sum_{j=1}^n b_j v_j}_{y}\right) = \sum_{i,j=1}^n a_i b_j H(v_i, v_j)$$

these $n \times n = n^2$ "coefficients" determine H completely.

(8)

Defⁿ: Let $\beta = \{v_1, v_2, \dots, v_n\}$ be a basis for V . If $H \in \mathcal{B}(V)$,

$$\Psi_{\beta}(H) := \begin{pmatrix} H(v_1, v_1) & H(v_1, v_2) & \cdots & H(v_1, v_n) \\ H(v_2, v_1) & H(v_2, v_2) & \cdots & H(v_2, v_n) \\ \vdots & \vdots & \ddots & \vdots \\ H(v_n, v_1) & H(v_n, v_2) & \cdots & H(v_n, v_n) \end{pmatrix} = (H(v_i, v_j))$$

is the matrix representation of H w.r.t. β .

Example: Consider $H: \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$ to be the "determinant":

$$H\left(\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}\right) := \det \begin{pmatrix} x_1 & y_1 \\ x_2 & y_2 \end{pmatrix} = x_1 y_2 - x_2 y_1$$

By properties of det., H is a bilinear form.

Fix the standard basis $\beta = \{e_1, e_2\}$

$$H(e_1, e_1) = \det \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} = 0$$

$$H(e_1, e_2) = \det \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = 1$$

$$H(e_2, e_1) = \det \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = -1$$

$$H(e_2, e_2) = \det \begin{pmatrix} 0 & 0 \\ 1 & 1 \end{pmatrix} = 0$$

$$\boxed{\Psi_{\beta}(H) = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}}$$

skew-symmetric
" $A^t = -A$ "

By a very similar argument in establishing $\mathcal{L}(V) \cong M_{n \times n}(\mathbb{F})$,

we have:

Theorem: The map $\Psi_{\beta}: \mathcal{B}(V) \xrightarrow{\cong} M_{n \times n}(\mathbb{F})$ is an isomorphism between vector spaces over \mathbb{F} .

Proof: Exercise (see textbook Thm. 6.32 if you got stuck.)

Change of Basis Formula

Recall that for a linear operator $T: V \rightarrow V$, if we have two bases β and γ for V , then the matrices $[T]_{\beta}, [T]_{\gamma}$ are related by
$$(*)_T \quad [T]_{\gamma} = Q^{-1} [T]_{\beta} Q$$
 where Q is an invertible matrix.

? What about for bilinear forms ?

Theorem: Given a bilinear form $H \in \mathcal{B}(V)$, if β and γ are two bases for V , then the matrices $\psi_{\beta}(H)$ and $\psi_{\gamma}(H)$ are related by

$$(*)_H \quad \psi_{\gamma}(H) = Q^t \psi_{\beta}(H) Q$$

where $Q = [I]_{\gamma}^{\beta}$ is the invertible change of basis matrix from γ to β .

Caution! The matrix Q in $(*)_H$ may not be orthogonal, i.e. $Q^t \neq Q$ in general.

Note: Comparing $(*)_T$ and $(*)_H$, we see that although both linear operators T and bilinear forms H are represented by matrices, they transform differently when we change basis. Hence T and H are somewhat different in some fundamental way. In the language of "tensors", $T \in V \otimes V^*$ but $H \in V^* \otimes V^*$.

Proof of $(*_H)$:

Let Q be the change of coordinate matrix from γ to β ,

i.e. $(*) \quad [\mathbf{v}]_{\beta} = Q [\mathbf{v}]_{\gamma} \quad \text{for all } \mathbf{v} \in V.$

Now, by definition of matrix representation of H , we have

$$\begin{aligned} H(x, y) &= [x]_{\beta}^t \Psi_{\beta}(H) [y]_{\beta} \quad \& \quad H(x, y) = [x]_{\gamma}^t \Psi_{\gamma}(H) [y]_{\gamma} \\ &= [x]_{\gamma}^t [Q^t \Psi_{\beta}(H) Q] [y]_{\gamma} \end{aligned}$$

compare these expressions!

Because of $(*_H)$, we identify matrices which represent the same bilinear form H but in different bases.

Defⁿ: Two matrices $A, B \in M_{n \times n}(F)$ are **congruent**
if there exists invertible $Q \in M_{n \times n}(F)$ s.t. $B = Q^t A Q$

Diagonalizability

Since we can represent bilinear forms H as matrices, we have a similar notion of "diagonalizability":

Defⁿ: $H \in B(V)$ is **diagonalizable** iff there exists a basis β for V
s.t. $\Psi_{\beta}(H)$ is diagonal.

In terms of matrices, it is the same as asking:

Given $A \in M_{n \times n}(F)$, does there exist an invertible $Q \in M_{n \times n}(F)$
s.t. $Q^t A Q$ is diagonal?

! This is different from our usual notion of diagonalizing a matrix!

Symmetric Bilinear Forms

Fixing a basis β for V , we have a 1-1 correspondence

$$\begin{array}{ccc} \mathcal{B}(V) & \xleftrightarrow{\cong} & M_{n \times n}(\mathbb{F}) \\ \downarrow & & \downarrow \\ H & \longleftrightarrow & \psi_\beta(H) \end{array} \quad \dim V = n$$

As symmetric matrices are particularly nice in many aspects,

we ask : For which H is $\psi_\beta(H)$ a symmetric matrix ?

Defⁿ : $H \in \mathcal{B}(V)$ is symmetric iff $H(x, y) = H(y, x) \quad \forall x, y \in V$

FACT : $H \in \mathcal{B}(V)$ symmetric $\Leftrightarrow \psi_\beta(H)$ symmetric

Pf: Exercise!

The following is the most important result about symmetric bilinear forms :

Theorem: Let V be a finite dim'l vector space over $\mathbb{F} = \mathbb{R}$ or \mathbb{C} .

$H \in \mathcal{B}(V)$ is diagonalizable

\Leftrightarrow

H is a symmetric bilinear form

Caution: This is different from the Spectral Theorems we discussed since there is NO inner product $\langle \cdot, \cdot \rangle$ involved!

Proof: By induction on $n = \dim V$.

$n = 1$: trivial

Assume theorem is true for $n = k - 1$.

Suppose now $n = \dim V = k$. WLOG, assume $H \neq 0$.

Claim: $H(x, x) \neq 0$ for some $x \in V$

Proof of Claim: $H \neq 0 \Rightarrow H(u, v) \neq 0$ for some $u, v \in V$

Let $X = u + v$. Suppose $H(u, u) = H(v, v) = 0$. Otherwise we are done. Then

$$\begin{aligned} H(x, x) &= H(u+v, u+v) && H \text{ symmetric} \\ (\text{bilinearity}) \quad &= H(u, u) + \underline{H(u, v)} + \underline{H(v, u)} + H(v, v) \\ &= 0 + 2H(u, v) + 0 \neq 0. \end{aligned}$$

Fix some $0 \neq x \in V$ s.t. $H(x, x) \neq 0$.

Consider the linear map $L_x : V \rightarrow \mathbb{F}$ defined by

"fixing one of the variable to be x ", i.e.

$$L_x(y) := H(x, y)$$

Since $L_x(x) = H(x, x) \neq 0$ by our choice of x , L_x is onto.

By rank-nullity theorem,

$$\dim N(L_x) = \dim V - \dim R(L_x) = k - 1.$$

Therefore, $H : \underbrace{N(L_x) \times N(L_x)}_{\dim = k-1} \rightarrow \mathbb{F}$ is again a symmetric bilinear form

and is thus diagonalizable by some basis $\{v_1, \dots, v_{k-1}\}$ for $N(L_x)$.

Then, $\{v_1, \dots, v_{k-1}, x\}$ is a basis for V which diagonalize H .

Quadratic Forms

Recall that an inner product induces a norm :

$$\langle \cdot, \cdot \rangle \longrightarrow \|x\|^2 := \langle x, x \rangle$$

Note: But NOT vice versa, unless we have a "nice" norm
 (see Exercise #27 in 6.1)

We have something similar for bilinear forms which are symmetric :

Defⁿ: For a symmetric bilinear form $H \in B(V)$ we associate a quadratic form $K : V \rightarrow \mathbb{F}$ where

$$K(x) := H(x, x) \quad \text{for all } x \in V$$

Example: (Conic sections)

Remember that any symmetric bilinear form H on \mathbb{R}^2 is represented by a 2×2 symmetric real matrix :

$$H\left(\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}\right) = (x_1 \ x_2) \begin{pmatrix} A & B \\ B & C \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$$

The quadratic form associated to H is simply

$$K\left(\begin{pmatrix} x \\ y \end{pmatrix}\right) = (x \ y) \begin{pmatrix} A & B \\ B & C \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = Ax^2 + 2Bxy + Cy^2$$

which is just a homogeneous quadratic polynomial in x and y .

For example, if $A=1, B=2, C=5$,

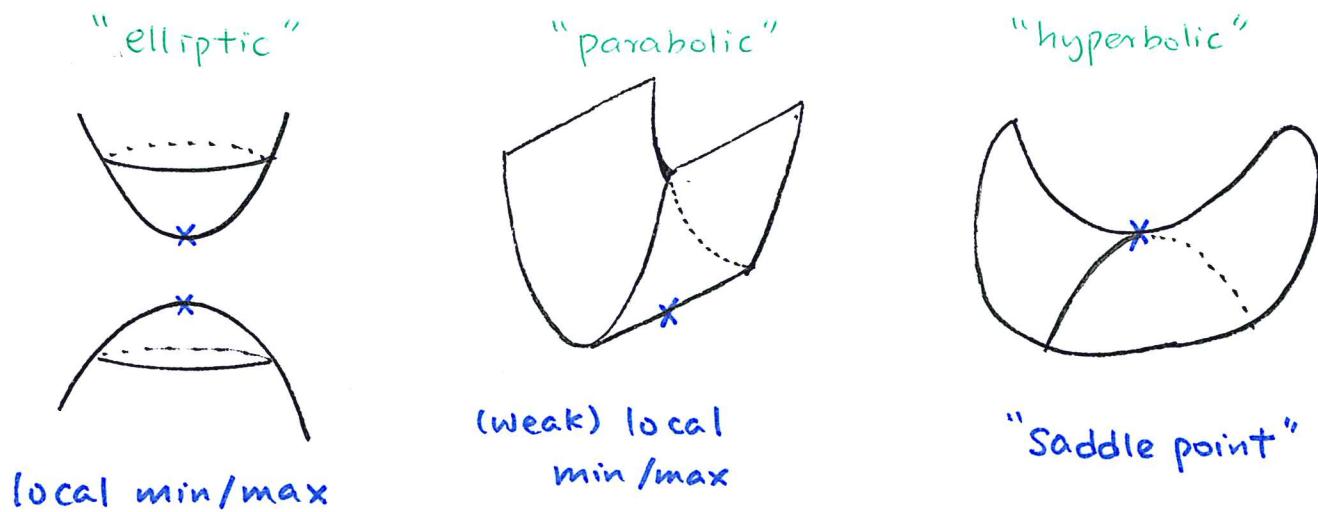
$$K(x, y) = x^2 + 2xy + 5y^2 = (x+2y)^2 + y^2 = u^2 + v^2$$

↑
if we change coordinates $\begin{cases} u = x + 2y \\ v = y \end{cases}$

By a linear change of coordinates, we can always express it in one of the following forms (Exercise: Can you prove this?)

$$K(u, v) = \begin{cases} u^2 + v^2 \text{ or } -u^2 - v^2 & \text{"elliptic"} \\ \pm u^2 & \text{"parabolic"} \\ -u^2 + v^2 & \text{"hyperbolic"} \end{cases}$$

Locally near $(0,0)$, their graphs look like



Example: Does the function $f(x, y) = 5x^2 + 4xy + 2y^2$ have a local min./max or saddle point at $(0,0)$?

Solution: By "completing the square",

$$\begin{aligned} f(x, y) &= 5x^2 + 4xy + 2y^2 = 3x^2 + 2(x+y)^2 \\ &= u^2 + v^2 \quad \text{if we let } u = \sqrt{3}x, v = \sqrt{2}(x+y) \end{aligned}$$

Hence, $(0,0)$ is a local minimum.

Alternatively, f is represented by the symmetric 2×2 matrix

$$\begin{pmatrix} x & y \\ 5 & 2 \\ y & 2 \end{pmatrix} \text{ which has eigenvalues } 1 \text{ and } 6 \xrightarrow[\text{positive}]{\text{both}} \text{local min}$$

$$\text{Char. eqn: } (5-\lambda)(2-\lambda) - 4 = 0 \Rightarrow \lambda^2 - 7\lambda + 6 = 0 \Rightarrow \lambda = 1 \text{ or } 6$$