# Linear Algebra II

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# §1 Bilinear forms

Let V be a finite dimensional vector space, dim V = n.

**Definition 1.0.1.** Let  $F = \mathbb{C}$ , we say a function  $f: V \times V \to V$  is a **semi bilinear form** if:

- $f(c_1\alpha + \beta, \gamma) = c_1 f(\alpha, \gamma) + f(\beta, \gamma);$
- $f(\alpha, c_1\beta + \gamma) = \overline{c}_1 f(\alpha, \beta) + f(\alpha, \gamma)$ .

Let Form(V) denote the (semi) bilinear forms on (complex) real vector space V.

For  $f \in \text{Form}(V)$ , fix a basis  $\mathcal{B} = \{\alpha_1, \dots, \alpha_n\}$  of V, let  $[f]_{\mathcal{B}} \in F^{n \times n}$  be the matrix

$$([f]_{\mathcal{B}})_{jk} = f(\alpha_k, \alpha_j).$$

which is called the matrix of f under  $\mathcal{B}$ .

For  $\alpha = \sum_{k=1}^{n} x_k \alpha_k$ ,  $\beta = \sum_{j=1}^{n} y_j \alpha_j \in V$ . It's clear that

$$f(\alpha,\beta) = \sum_{j,k=1}^{n} x_k \overline{y}_j f(\alpha_k, \alpha_j) = \sum_{j,k=1}^{n} x_k \overline{y}_j ([f]_{\mathcal{B}})_{jk} = [\beta]_{\mathcal{B}}^* [f]_{\mathcal{B}} [\alpha]_{\mathcal{B}}.$$

From this we know that the map  $\text{Form}(V) \to F^{n \times n}, f \mapsto [f]_{\mathcal{B}}$  is a linear isomorphism. Since if  $[f]_{\mathcal{B}} = 0$ , then  $f(\alpha, \beta) = 0$  for all  $\alpha, \beta \in V$ . Thus it's injective. Obviously it's surjective and linear, so

$$\dim \operatorname{Form}(V) = n^2$$

#### Example 1.0.2

Let  $A \in F^{n \times n}$ . Let  $f \in \text{Form}(F^{n \times 1})$  be

$$f(X,Y) = Y^*AX, \quad \forall X, Y \in F^{n \times 1}.$$

Let  $\mathcal{B}$  be the standard basis of F, it's clear that  $[f]_{\mathcal{B}} = A$ .

#### **Proposition 1.0.3**

Let  $\mathcal{B}' = \{\alpha'_1, \dots, \alpha'_n\}$  be another basis of  $V, P \in GL_n(F)$  satisfies

$$(\alpha'_1,\ldots,\alpha'_n)=(\alpha_1,\ldots,\alpha_n)P.$$

Then  $[f]_{\mathcal{B}'} = P^*[f]_{\mathcal{B}}P$ .

*Proof.* Since  $[\alpha]_{\mathcal{B}} = P[\alpha]_{\mathcal{B}'}$ , just plug this into the definition of  $[f]_{\mathcal{B}}$ , the rest is trivial.

**Definition 1.0.4.** Let  $f \in Form(V)$ .

- When  $F = \mathbb{R}$ , if  $\forall \alpha, \beta \in V$  we have  $f(\alpha, \beta) = f(\beta, \alpha)$ , then we say f is symmetrical (also called Hermite);
- When  $F = \mathbb{C}$ , if  $\forall \alpha, \beta \in V$  we have  $f(\alpha, \beta) = \overline{f(\beta, \alpha)}$ , we say f is Hermite.

### **Proposition 1.0.5**

When  $F = \mathbb{C}$ , f Hermite  $\iff f(\alpha, \alpha) \in \mathbb{R}$ ,  $\forall \alpha \in V$ .

*Proof.* For the " $\Leftarrow$ " direction, consider  $f(\alpha + \beta, \alpha + \beta) \in \mathbb{R}$ . Expanding we'll get  $f(\alpha, \beta) + f(\beta, \alpha) \in \mathbb{R}$ , i.e.

$$f(\alpha, \beta) + f(\beta, \alpha) = \overline{f(\alpha, \beta)} + \overline{f(\beta, \alpha)}.$$

Replace  $\alpha$  with  $i\alpha$ , we get

$$f(\alpha, \beta) - f(\beta, \alpha) = -\overline{f(\alpha, \beta)} + \overline{f(\beta, \alpha)}.$$

Combining two equations we get the conclusion.

**Definition 1.0.6.** Let  $f \in \text{Form}(V)$  be an Hermite form. If  $\forall \alpha \in V \setminus \{0\}$ ,  $f(\alpha, \alpha) > 0$ , we say f is **positive definite**.

Similarly we can define negative definite and semi positive definite.

Note that a positive definite Hermite form is nothing but an inner product.

# §1.1 Positive define matrices

In this section we'll dig deeper into properties of positive definite matrices.

It's clear that if a matrix A is positive definite, then A is inversible, and  $P^*AP$  is also positive definite. In particular,  $P^*P$  is positive definite.

# Theorem 1.1.1 (Cholesky decomposition)

Let  $A \in F^{n \times n}$  be a positive definite matrix, there exists a unique upper triangular matrix R with positive diagonal entries s.t.  $A = R^*R$ .

*Proof.* Consider the inner product  $f(X,Y) = Y^*AX$ . Let the standard inner product on V be  $f_0(X,Y) = Y^*X$ .

Since inner product spaces with same dimensions are isomorphic, so there exists a matrix  $R \in GL_n(F)$ , such that

$$R: (F^{n\times 1}, f) \to (F^{n\times 1}, f_0), \quad X \mapsto RX$$

is an isomorphism of inner product space, i.e.  $f_0(RX,RY) = f(X,Y)$ . This is equivalent to  $A = R^*R$ .

For any  $P \in GL_n(F)$ , P is also an isomorphism of  $(F^{n\times 1}, f) \to (F^{n\times 1}, f_0)$  iff  $RP^{-1}$  preserves the inner product  $f_0$ , iff  $RP^{-1} \in O(n)$  or U(n).

By QR decomposition,  $R = RP^{-1} \cdot P$ , so there must be a unique P s.t. P upper triangular with positive diagonal entries.

# Corollary 1.1.2

A positive definite  $\implies$  det A > 0.

**Definition 1.1.3.** Let  $A \in F^{n \times n}$ , for  $1 \le k \le n$ , define

$$\Delta_k(A) := \det(A_{1 \le i \le k}^{1 \le j \le k})$$

be the leading principal minor.

#### Theorem 1.1.4

Let  $A \in F^{n \times n}$  be an Hermite matrix. Then A positive definite  $\iff \Delta_k(A) > 0, k = 1, \dots, n$ .

# Lemma 1.1.5 (LU decomposition)

Let F be any field. For  $A \in GL_n(F)$ , the followings are equivalent:

- $\Delta_k(A) \neq 0, k = 1, \ldots, n;$
- A = LU, where L lower triangular, and U upper triangular with diagonal entries 1.

*Proof.* On one hand, Let  $L_k, U_k$  be the top-left  $k \times k$  submatrix of L, U, since L, U inversible,  $L_k, U_k$  inversible. By the triangular property,  $\Delta_k(A) = \det(L_k U_k) \neq 0$ .

On the other hand, it's sufficient to prove:

 $\exists N \text{ strictly upper triangular}, A(N+I_n) \text{ lower triangular}$ 

Let  $A_k$  be the k-th leading principal submatrix of A, and  $\alpha_{k+1}, \beta_{k+1} \in F^{n \times 1}$  the (k+1)-th column of A, N.

Now compute the first k rows of the (k+1)-th column of A(N+I), which is equal to  $A_k\beta'_{k+1}+\alpha'_{k+1}$ , where  $\alpha'_{k+1},\beta'_{k+1}$  is the first k entries of  $\alpha_{k+1},\beta_{k+1}$ .

Since  $A_k$  inversible,  $\exists \beta'_{k+1}$  s.t.  $A_k \beta'_{k+1} + \alpha'_{k+1} = 0$ .

Hence these  $\beta'_{k+1}$  forms a strictly upper triangular matrix N, as desired.

Proof of the theorem. Let A be an Hermite matrix, if A positive definite, then det  $A \ge 0$ . Let  $A_k$  be the upper left  $k \times k$  submatrix of A. For  $X \in F^{k \times 1} \setminus \{0\}$ , we have

$$X^*A_kX = \begin{pmatrix} X \\ 0 \end{pmatrix}^*A\begin{pmatrix} X \\ 0 \end{pmatrix} > 0.$$

Hence  $A_k$  positive definite,  $\det A_k = \Delta_k(A) \geq 0$ .

Conversely, by our lemma let A = LU, let  $D = (U^*)^{-1}L$ ,  $A = U^*DU$ .

Hence A Hermite  $\implies D$  Hermite. Moreover D is lower triangular, so D is diagonal.

Some computation yields that  $A_k = U_k^* D_k U_k$ . Therefore

$$\Delta_k(A) \ge 0 \implies \det(U_k^* D_k U_k) \ge 0 \implies \det D_k \ge 0.$$

From this we deduce that all the diagonal entries of D are positive, so D positive definite  $\implies A$  positive definite.

# §1.2 Bilinear forms on inner product spaces

Let V be an inner product space, given a basis of V, recall that there are two linear isomorphism:

$$Form(V) \to F^{n \times n}, f \mapsto [f]_{\mathcal{B}} \quad L(V) \to F^{n \times n}, T \mapsto [T]_{\mathcal{B}}$$

Hence we can define a map  $Form(V) \to L(V)$  by composing these two isomorphism. Denote this map by  $f \mapsto T_f$ . It seems like this map also depends on the choice of the basis, but in fact it's independent as long as  $\mathcal{B}$  is orthonormal!

Let  $\mathcal{B}'$  be another orthonormal basis, then  $[T_f]_{\mathcal{B}'} = P^{-1}[T_f]_{\mathcal{B}}P$ , while  $[f]_{\mathcal{B}'} = P^*[f]_{\mathcal{B}}P$ , but P is orthogonal (or unitary), so  $P^{-1} = P^*$ , i.e.  $T_f$  doesn't change under the new basis.

Since  $T_f$  do not depend on the basis, thus we wonder whether we can define this map intrinsically.

# **Proposition 1.2.1**

For all  $T \in L(V)$ , T induces a (semi) bilinear form  $f(\alpha, \beta) = \langle T\alpha, \beta \rangle$ . We claim that this map  $\mathcal{F}$  gives an isomorphism of L(V) and Form(V).

*Proof.* Clearly  $\mathcal{F}$  is injective:

$$\langle T\alpha, \beta \rangle = 0, \forall \beta \implies T\alpha = 0,$$

thus  $\ker \mathcal{F} = \{0\}.$ 

By dimenional reasons  $\mathcal{F}$  must be an isomorphism.

By considering  $\mathcal{F}^{-1}$ , we get an one-to-one map  $f \mapsto T_f$  s.t.

$$f(\alpha, \beta) = \langle T_f \alpha, \beta \rangle$$
.

We'll see that this definition coincide with the initial one. In fact it's sufficient to prove  $[T_f]_{\mathcal{B}} = [f]_{\mathcal{B}}$ , which is just a bunch of computation;)

**Remark 1.2.2** — We can construct  $T_f$  explicitly from f:

The inner product gives a conjugate linear isomorphism

$$\Phi: V \to V^*, \quad \Phi(\alpha)(\beta) = \langle \beta, \alpha \rangle = \overline{\langle \alpha, \beta \rangle}.$$

Similarly,  $f \in \text{Form}(V)$  gives a conjugate linear map

$$\Phi_f: V \to V^*, \quad \Phi_f(\alpha)(\beta) = \overline{f(\alpha, \beta)}.$$

Then  $T = \Phi^{-1} \circ \Phi_f$  is the desired linear map:

$$\langle T\alpha, \beta \rangle = \overline{\Phi(T\alpha)(\beta)} = \overline{\Phi_f(\alpha)(\beta)} = f(\alpha, \beta).$$

Hence all the properties of linear maps can be carried over to the forms, and vice versa (using the matrix representation).

# Corollary 1.2.3

Let  $F = \mathbb{C}$ ,  $T \in L(V)$ , T self-adjoint iff  $\langle T\alpha, \alpha \rangle \in \mathbb{R}, \forall \alpha \in V$ .

*Proof.* T self-adjoint iff f Hermite iff  $f(\alpha, \alpha) \in \mathbb{R}$ .

# Corollary 1.2.4

Let  $f \in \text{Form}(V)$ .

- If f Hermite, there exists an orthonormal basis of V s.t.  $[f]_{\mathcal{B}}$  is real diagonal.
- If  $F = \mathbb{C}$ , there exists an orthonormal basis such that  $[f]_{\mathcal{B}}$  upper triangular.

# §1.3 Spectral decomposition

# Theorem 1.3.1 (Spectral decomposition of normal maps)

Let  $T \in L(V)$  be a self-adjoint map (or normal map in complex field), let  $\sigma(T) = \{c_1, \ldots, c_k\}$ ,  $P_i \in L(V)$  are the projection onto  $V_{c_i}$ . Then for any  $f \in F[x]$ , we have

$$f(T) = \sum_{i=1}^{k} f(c_i) P_i.$$

In particular,  $T = \sum_{i=1}^{k} c_i P_i$ .

Proof. Consider the orthogonal direct sum

$$V = \bigoplus_{i=1}^{k} V_{c_i},$$

since previously we've proven that T is orthogonally diagonalizable (or unitarily diagonalizable). Using this decomposition, the conclusion is somewhat trivial.

# Corollary 1.3.2

Each  $P_i$  is a polynomial of T.

*Proof.* Take 
$$f_i \in F[x]$$
 s.t.  $f_i(c_i) = \delta_{ij}$ . Then  $f_i(T) = \sum_{j=1}^k f_i(c_j) P_j = P_i$ .

Using similar technique, we can consider functions other than polynomials of T, defined by  $\phi(T) = \sum_{i=1}^k \phi(c_i)T$ . By Lagrange interpolation, we can always find a polynomial p s.t.  $p(c_i) = \phi(c_i)$  for all  $c_i \in \sigma(T)$ .

### Example 1.3.3

If T semi positive definite normal matrix,  $\sigma(T) \subset [0, +\infty)$ , so we can define  $\sqrt{T} = \sum_{i=1}^{k} \sqrt{c_i} P_i$ .

# **Proposition 1.3.4**

T self-adjoint (normal)  $\implies \phi(T)$  self-adjoint (normal);  $\sigma(\phi(T)) = \phi(\sigma(T))$ .

*Proof.* Note that T and  $\phi(T)$  are diagonal matrices under orthonormal basis of  $V_{c_i}$ .

#### Theorem 1.3.5

Let  $T \in L(V)$  be semi positive definite.

- $\sqrt{T}$  semi positive definite, and  $\sqrt{T}^2 = T$ .
- T positive definite  $\iff \sqrt{T}$  positive definite.
- If  $S \in L(V)$  semi positive definite and  $S^2 = T$ , then  $S = \sqrt{T}$ .

Proof. Since  $[\sqrt{T}]_{\mathcal{B}} = \operatorname{diag}(\sqrt{c_1}I_{d_1}, \ldots, \sqrt{c_k}I_{d_k})$ , the first two statements are trivial. Let  $\sigma(S) = \{s_1, \ldots, s_r\}$ ,  $V_i = \ker(S - s_i \operatorname{id})$ . Since S self-adjoint,  $V = \bigoplus_{i=1}^r V_i$ .

For any  $\alpha \in V_i, T\alpha = S^2\alpha = s_i^2\alpha$ , thus  $V_i \subset \ker(T - s_i^2 \operatorname{id})$ . Since  $s_i \geq 0, \sqrt{T} = S$ .

Note that  $T^*T$  is always positive definite, so we can consider  $\sqrt{T^*T}$ . We call the eigen-values of  $\sqrt{T^*T}$  singular values of T.

In some sense,  $\sqrt{T^*T}$  is a semi positive approximation of T:

#### Lemma 1.3.6

For any  $\alpha \in V$ ,  $||T\alpha|| = ||\sqrt{T^*T}\alpha||$ . In particular,  $\ker T = \ker \sqrt{T^*T}$ .

Proof. Let  $N = \sqrt{T^*T}$ ,

$$||N\alpha||^2 = \langle N\alpha, N\alpha \rangle = \langle N^2\alpha, \alpha \rangle = \langle T^*T\alpha, \alpha \rangle = \langle T\alpha, T\alpha \rangle = ||T\alpha||^2.$$

### **Theorem 1.3.7** (Polar decomposition)

Let  $T \in L(V)$ ,

(1) There exists  $U \in L(V)$  orthogonal or unitary,  $N \in L(V)$  semi positive definite, T = UN

- (2) We must have  $N = \sqrt{T^*T}$ .
- (3) T invertible iff N positive definite, in this case U is unique.

**Remark 1.3.8** — This is a generalization of  $z = re^{i\theta}$  in  $\mathbb{C}$ . That's where the name comes from.

Proof. If (1) holds, then

$$T^* = NU^* \implies T^*T = NU^*UN = N^2 \implies N = \sqrt{T^*T}.$$

Since T, N are semi positive definite, T invertible iff T positive definite. Now we must have  $U = TN^{-1}$ , which is unique.

To prove (1), when T invertible, let N, U as above, by our lemma,

$$||U\alpha|| = ||TN^{-1}\alpha|| = ||\alpha||$$

Thus U is orthogonal or unitary.

When T is not invertible,  $\ker T = \ker N$ , thus  $\exists U_1 : \operatorname{Im}(N) \to \operatorname{Im}(T)$  s.t.  $T = U_1 N$ . (Just take  $N\alpha \mapsto T\alpha$ )

Moreover  $U_1$  is an isomorphism of inner product space:  $||U_1N\alpha|| = ||T\alpha|| = ||N\alpha||$ . So  $U_1$  preserves inner product and hence injective. By dimension reasons,  $U_1$  must be an isomorphism.

Now we can take an arbitary isomorphism  $U_2: \operatorname{Im}(N)^{\perp} \to \operatorname{Im}(T)^{\perp}, U:=U_1 \oplus U_2$  is the desired map.

Looking back at the singular values, consider the image of unit sphere  $S \subset V$  under T, N(S) is an ellipsoid:

$$N(S) = \left\{ \sum_{i=1}^{n} c_i x_i \alpha_i : \sum_{i=1}^{n} x_i^2 = 1 \right\}.$$

Since T = UN, U is a rotation, so T(S) is also an ellipsoid, whose axes lengths are  $2c_i$ , where  $c_i$  are singular values of T.

# Corollary 1.3.9 (Singular value decomposition, SVD)

Let  $A \in F^{n \times n}$ , then there exists decomposition  $A = U_1 D U_2$ , where D is a diagonal matrix with non-negative entries,  $U_1, U_2$  are orthogonal or unitary matrices.

*Proof.* Consider the polar decomposition A=UN, let  $N=PDP^{-1}$ , where P orthogonal or unitary, D non-negative diagonal. Thus we can take  $U_1=UP, U_2=P^{-1}$ .

Note that the diagonal entries of D is precisely the singular value of A.

### Corollary 1.3.10

Let  $T \in L(V)$ , then T map some orthogonal basis to another orthogonal basis.

*Proof.* Let T = UN be the polar decomposition. Let  $\alpha_1, \ldots, \alpha_n$  be an orthonormal basis s.t. N diagonal, then

$$T\alpha_i = UN\alpha_i = c_i U\alpha_i$$

obviously  $c_i U \alpha_i$  consititude an orthogonal basis.

# §1.4 Further on normal maps

For  $\theta \in \mathbb{R}$ , let  $Q_{\theta}$  be the rotation of angle  $\theta$ . The main goal of this section is to prove:

#### Theorem 1.4.1

Let V be a finite dimensional real inner product space,  $T \in L(V)$  normal. There exists an orthonormal basis  $\mathcal{B}$  s.t.

$$[T]_{\mathcal{B}} = \operatorname{diag}(a_1, \dots, a_l, r_1 Q_{\theta_1}, \dots, r_m Q_{\theta_m}),$$

where  $a_i \in \mathbb{R}, r_j > 0, \theta_j \in (0, \pi)$ .

Let's look at a corollary of this theorem first:

# Corollary 1.4.2

If T orthogonal, then

$$[T]_{\mathcal{B}} = \operatorname{diag}(I_{l_1}, -I_{l_2}, Q_{\theta_1}, \dots, Q_{\theta_m}).$$

*Proof.* Applying the theorem, since each block is orthogonal,  $a_i = \pm 1$ ,  $r_j = 1$ .

This gives us a comprehension of rotations in higher dimensional spaces.

Here we'll present multiple proofs to emphasize some intermediate result.

### **Proposition 1.4.3**

Let T be a normal map, if  $W \subset V$  is T-invariant, then  $T_W$  is also normal.

*Proof.* First note that  $W, W^{\perp}$  are  $T^*$ -invariant. For  $\alpha, \beta \in W$ , we have

$$\langle (T_W)^* \alpha, \beta \rangle = \langle \alpha, T_W \beta \rangle = \langle \alpha, T \beta \rangle = \langle T^* \alpha, \beta \rangle.$$

Thus  $(T_W)^* = T_W^*$ . The conclusion follows.

#### **Proposition 1.4.4**

Let T be a normal map, there exists an orthogonal decomposition  $V = \bigoplus_{i=1}^k V_i$ , such that each  $V_i$  is T-invariant, and  $T_{V_i}$  simple.

*Proof.* Note that if W is T-invariant, then  $W^{\perp}$  is also T-invariant. By induction and the previous proposition this is trivial.

Therefore to prove Theorem 1.4.1, we only need to prove the case when T is simple.

Proof of Theorem 1.4.1. WLOG dim V > 1.

Since T simple  $\implies f_T \in \mathbb{R}[x]$  prime, thus deg  $f_T = 2$ , dim V = 2 and  $f_T = (x - c)(x - \overline{c})$ .

Take any orthonormal basis  $\mathcal{B} = \{\alpha_1, \alpha_2\}$ , let r = |c|,  $A = r^{-1}[T]_{\mathcal{B}}$ . Clearly A normal and  $\sigma(A) = \{r^{-1}c, r^{-1}\overline{c}\}$ , so A is unitarily similar to diag $(r^{-1}c, r^{-1}\overline{c})$ , A is unitary.

Moreover A is a real matrix so A orthogonal, and det A = 1, thus  $A = Q_{\theta}, \theta \in [0, 2\pi]$ .

At last by T has no eigenvector, and we can change  $\alpha_2$  to  $-\alpha_2$ , so we can require  $\theta \in (0, \pi)$ .  $\square$ 

### **Proposition 1.4.5**

Let  $T \in L(V)$ , then  $\ker(T)^{\perp} = \operatorname{im}(T^*), \operatorname{im}(T)^{\perp} = \ker(T^*).$ 

*Proof.* Trivial, just some computation.

# **Proposition 1.4.6**

Let  $T \in L(V)$ ,  $\sigma(T^*) = \overline{\sigma(T)}$ ,

 $\forall c \in \sigma(T), \quad \dim \ker(T - cI) = \dim \ker(T^* - \bar{c}I).$ 

*Proof.* By the previous proposition,

 $\dim \ker(T - cI) = n - \dim \operatorname{im}(T^* - \overline{c}I) = \dim \ker(T^* - \overline{c}I)$ 

which also implies  $\sigma(T) = \overline{\sigma(T^*)}$ .

#### **Proposition 1.4.7**

If T normal, then  $\ker(T - cI) = \ker(T^* - \overline{c}I)$ .

*Proof.* Let  $W = \ker(T - cI)$ ,  $T_W^*$  is just  $(c \operatorname{id}_W)^* = \overline{c} \operatorname{id}_W$ . Thus  $W \subset \ker(T^*0\overline{c}I)$ , by dimensional reasons they must be equal.

#### **Proposition 1.4.8**

Let T be a normal map,  $f, g \in F[x]$  coprime  $\implies \ker(f(T)) \perp \ker(g(T))$ .

*Proof.* Since  $g(T)^* = \overline{g}(T^*)$ , g(T) is normal, thus  $\ker(g(T))^{\perp} = \operatorname{im}(g(T))$ .

Let  $W = \ker(f(T))$ , let  $a, b \in F[x]$  s.t. af + bg = 1, so  $a(T)f(T) + b(T)g(T) = \mathrm{id}_V$ . Restrict this equation to W, we get  $b(T)_W g(T)_W = \mathrm{id}_W$ , hence  $W \subset \mathrm{im}(g(T))$ .

# **Proposition 1.4.9**

Let T be a normal map,

- The primary decomposition of T are orthogonal decomposition;
- $\bullet$  The cyclic decomposition of T can be orthogonal.

*Proof.* The first one is trivial by previous proposition.

For cyclic decomposition, we proceed by induction on  $\dim V$ .

Let  $\alpha_1 \in V$  s.t.  $p_{\alpha_1} = p_r$ , then  $(R\alpha_1)^{\perp}$  are *T*-invariant, use induction hypo on it and we're done.

**Remark 1.4.10** — This means the primary cyclic decomposition of *T* can also be orthogonal.

This gives the second proof of Theorem 1.4.1:

*Proof.* WLOG T normal and primary cyclic, then  $p_T$  is primary, and T normal  $\implies T$  semisimple, so  $p_T$  has no multiple factors, thus  $p_T$  prime, which proves the result.

Next we present the third proof:

### Proposition 1.4.11

If  $A, B \in \mathbb{R}^{n \times n}$  are unitarily similar, then they are orthogonally similar.

# Proposition 1.4.12 (QS decomposition)

For any unitary matrix U, U = QS where Q real orthogonal, S unitary and symmetrical. Moreover  $\exists f \in \mathbb{C}[x]$  s.t.  $S = f(U^tU)$ .