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In these notes the concepts of circulants,  $\tau$  and Toeplitz matrices, Hessenberg algebras and displacement decompositions, spaces of class  $\mathbb{V}$  and best least squares fits on such spaces, are introduced and investigated. As a consequence of the results presented, the choice of matrices involved in displacement decompositions, the choice of preconditioners in solving linear systems and the choice of Hessian approximations in quasi-Newton minimization methods, become possible in wider classes of low complexity matrix algebras.

The Fourier matrix, circulants, and fast discrete transforms

Consider the following  $n \times n$  matrix

$$P_1 = \begin{bmatrix} 0 & 1 & & & \\ & 0 & 1 & & \\ & & \ddots & & \\ & & & & 1 \\ 1 & & & 0 \end{bmatrix}.$$

Let  $\omega \in \mathbb{C}$ . Note that

$$P_1 \begin{bmatrix} 1\\1\\1\\1 \end{bmatrix} = \begin{bmatrix} 1\\1\\1\\1 \end{bmatrix} = 1 \begin{bmatrix} 1\\1\\1\\1 \end{bmatrix}, P_1 \begin{bmatrix} 1\\\omega\\\omega^{n-1}\\1 \end{bmatrix} = \begin{bmatrix} \omega\\\omega^2\\\omega^{n-1}\\1 \end{bmatrix} = \omega \begin{bmatrix} 1\\\omega\\\omega^{n-1}\\1 \end{bmatrix},$$

where the latter identity holds if  $\omega^n = 1$ . More in general, if  $\omega^n = 1$ , we have the following vectorial identities

$$P_1 \begin{bmatrix} 1\\ \omega^j\\ \omega^{(n-1)j} \end{bmatrix} = \begin{bmatrix} \omega^j\\ \omega^{(n-1)j}\\ 1 \end{bmatrix} = \omega^j \begin{bmatrix} 1\\ \omega^j\\ \omega^{(n-1)j} \end{bmatrix}, \ j = 0, 1, \dots, n-1,$$

or, equivalently, the following matrix identity

$$P_1W = WD_{1\omega^{n-1}},$$

$$D_{1\omega^{n-1}} = \begin{bmatrix} 1 & & \\ & \omega & \\ & & & \\ & & & \omega^{n-1} \end{bmatrix}, \ W = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & \omega & \omega^j & \omega^{n-1} \\ & & & & \\ 1 & \omega^{n-1} & \omega^{(n-1)j} & \omega^{(n-1)(n-1)} \end{bmatrix}.$$

Proposition. If  $\omega^n = 1$  and if  $\omega^j \neq 1$  for 0 < j < n, then  $W^*W = nI$ . proof: since  $|\omega| = 1$ ,  $\overline{\omega} = \omega^{-1}$ , we have

$$\begin{split} [W^*W]_{ij} &= [\overline{W}W]_{ij} = \sum_{k=1}^n [\overline{W}]_{ik} [W]_{kj} = \sum_{k=1}^n \overline{\omega}^{(i-1)(k-1)} \omega^{(k-1)(j-1)} \\ &= \sum_{k=1}^n \omega^{(k-1)(j-i)} = \sum_{k=1}^n (\omega^{j-i})^{k-1}. \end{split}$$

Thus  $[W^*W]_{ij} = n$  if i = j, and  $[W^*W]_{ij} = \frac{1-(\omega^{j-i})^n}{1-\omega^{j-i}} = 0$  if  $i \neq j$  (note that the assumption  $\omega^j \neq 1$  for 0 < j < n is essential in order to make  $1 - \omega^{j-i} \neq 0$ ).

By the result of the above Proposition, we can say that the following (symmetric) Fourier matrix

$$F = \frac{1}{\sqrt{n}}W$$

is unitary, i.e.  $F^*F = I$ .

*Exercise.* Prove that  $F^2 = JP_1$  where J is the permutation matrix  $J\mathbf{e}_k =$  $\mathbf{e}_{n+1-k}, k = 1, \dots, n$  (*J* is usually called anti-identity).

The matrix identity satisfied by  $P_1$  and W can be of course rewritten in terms of F,  $P_1F = FD_{1\omega^{n-1}}$ , thus we obtain the equality

$$P_1 = F D_{1\omega^{n-1}} F^*$$

which states that the Fourier matrix diagonalizes the matrix  $P_1$ , or, more precisely, that the columns of the Fourier matrix form a system of n unitarily orthonormal eigenvectors for the matrix  $P_1$  with corresponding eigenvalues  $1, \omega, \ldots, \omega^{n-1}.$ 

But if F diagonalizes  $P_1$ , then it diagonalizes all polynomials in  $P_1$ :

$$\begin{split} P_1^{k-1} &= FD_{1\omega^{n-1}}^{k-1}F^*, \\ \sum_{k=1}^n a_k P_1^{k-1} &= F\sum_{k=1}^n a_k D_{1\omega^{n-1}}^{k-1}F^* \\ &= F\left[\begin{array}{ccc} \sum_{k=1}^n a_k & & \\ & \sum_{k=1}^n a_k \omega^{k-1} & \\ & & \sum_{k=1}^n a_k \omega^{(n-1)(k-1)} \end{array}\right]F^* \\ &= Fd(W\mathbf{a})F^* = \sqrt{n}Fd(F\mathbf{a})F^* \end{split}$$

where by  $d(\mathbf{z})$  we mean the diagonal matrix whose diagonal entries are  $z_1, z_2, \ldots, z_n$ . Let us investigate the matrices  $P_1^{k-1}, k = 1, \ldots, n$ , and the matrix  $\sum_{k=1}^n a_k P_1^{k-1}$ in the case n = 4:

$$P_{1}^{0} = I, P_{1}^{1} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \end{bmatrix}, P_{1}^{2} = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}, P_{1}^{3} = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix},$$
$$P_{1}^{4} = P_{1}^{3}P_{1} = P_{1}^{T}P_{1} = I = P_{1}^{0},$$
$$\sum_{k=1}^{4} a_{k}P_{1}^{k-1} = \begin{bmatrix} a_{1} & a_{2} & a_{3} & a_{4} \\ a_{4} & a_{1} & a_{2} & a_{3} \\ a_{2} & a_{4} & a_{1} & a_{2} \end{bmatrix} = \sqrt{4}Fd(F\mathbf{a})F^{*}, F = \frac{1}{\sqrt{4}} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & \omega & \omega^{2} & \omega^{3} \\ 1 & \omega^{2} & \omega^{4} & \omega^{6} \end{bmatrix}$$

$$\sum_{k=1}^{4} a_k P_1^{k-1} = \begin{bmatrix} a_4 & a_1 & a_2 & a_3\\ a_3 & a_4 & a_1 & a_2\\ a_2 & a_3 & a_4 & a_1 \end{bmatrix} = \sqrt{4} F d(F\mathbf{a}) F^*, \ F = \frac{1}{\sqrt{4}} \begin{bmatrix} 1 & \omega & \omega^2 & \omega^3\\ 1 & \omega^2 & \omega^4 & \omega^6\\ 1 & \omega^3 & \omega^6 & \omega^9 \end{bmatrix}$$
$$\omega^4 = 1, \ \omega^j \neq 1, \ 0 < j < 4 \ (\omega = e^{\pm \mathbf{i} 2\pi/4}).$$

Note that, for *n* generic, we have the identities  $\mathbf{e}_1^T P_1^{k-1} = \mathbf{e}_k^T$ , k = 1, ..., n, and  $P_1^n = I$  (prove them!). So, the set  $C = \{p(P_1)\}$  of all polynomials in  $P_1$ 

is spanned by the matrices  $J_k = P_1^{k-1}$ ; the particular polynomial  $\sum_{k=1}^n a_k J_k$  is simply denoted by  $C(\mathbf{a})$ . Note that  $C(\mathbf{a})$  is the matrix of C with first row  $\mathbf{a}^T$ :

$$C(\mathbf{a}) = \sum_{k=1}^{n} a_k J_k = \begin{bmatrix} a_1 & a_2 & a_{n-1} & a_n \\ a_n & a_1 & & a_{n-1} \\ & & & & \\ a_3 & & & & a_2 \\ a_2 & a_3 & & a_n & & a_1 \end{bmatrix} = Fd(F^T \mathbf{a})d(F^T \mathbf{e}_1)^{-1}F^{-1}.$$

C is known as the space of circulant matrices.

*Exercise.* (i) Repeat all, starting from the  $n \times n$  matrix

$$P_{-1} = \begin{bmatrix} 0 & 1 & & \\ & 0 & 1 & & \\ & & \ddots & & \\ & & & \ddots & \\ & & & & 1 \\ -1 & & & 0 \end{bmatrix}$$

and arriving to the (-1)-circulant matrix whose first row is  $\mathbf{a}^T$ ,  $\mathbf{a} \in \mathbb{C}^n$ :

$$C_{-1}(\mathbf{a}) = \begin{bmatrix} a_1 & a_2 & a_{n-1} & a_n \\ -a_n & a_1 & & a_{n-1} \\ & & & & \\ -a_3 & & & a_2 \\ -a_2 & -a_3 & -a_n & a_1 \end{bmatrix}.$$

(ii) Let T be a Toeplitz  $n \times n$  matrix, i.e.  $T = (t_{i-j})_{i,j=1}^n$ , for some  $t_k \in \mathbb{C}$ . Show that T can be written as the sum of a circulant and of a (-1)-circulant, that is,  $T = C(\mathbf{a}) + C_{-1}(\mathbf{b})$ ,  $\mathbf{a}, \mathbf{b} \in \mathbb{C}^n$ .

Why circulant matrices can be interesting in the applications of linear algebra? The main reason is in the fact that the matrix-vector product  $C(\mathbf{a})\mathbf{z}$  can be computed in at most  $O(n \log_2 n)$  arithmetic operations (whereas, usually, a matrix-vector product requires  $n^2$  multiplications).

Proposition FFT. Given  $\mathbf{z} \in \mathbb{C}^n$ , the complexity of the matrix-vector product  $F\mathbf{z}$  is at most  $O(n \log_2 n)$ . Such operation is called discrete Fourier transform (DFT) of  $\mathbf{z}$ . As a consequence, the matrix-vector product  $C(\mathbf{a})\mathbf{z}$  is computable by two DFTs (after the preprocessing DFT  $F\mathbf{a}$ ).

proof: since  $\omega^{(i-1)(k-1)}$  is the (i,k) entry of W and  $z_k$  is the k entry of  $\mathbf{z} \in \mathbb{C}^n$ , we have

$$(W\mathbf{z})_{i} = \sum_{k=1}^{n} \omega^{(i-1)(k-1)} z_{k} = \sum_{j=1}^{n/2} \omega^{(i-1)(2j-2)} z_{2j-1} + \sum_{j=1}^{n/2} \omega^{(i-1)(2j-1)} z_{2j} = \sum_{j=1}^{n/2} (\omega^{2})^{(i-1)(j-1)} z_{2j-1} + \sum_{j=1}^{n/2} \omega^{(i-1)(2(j-1)+1)} z_{2j} = \sum_{j=1}^{n/2} (\omega^{2})^{(i-1)(j-1)} z_{2j-1} + \omega^{i-1} \sum_{j=1}^{n/2} (\omega^{2})^{(i-1)(j-1)} z_{2j}.$$

Note that  $\omega$  is in fact a function of n, i.e. the right notation for  $\omega$  should be  $\omega_n$ . Then  $\omega^2 = \omega_n^2$  is such that  $(\omega_n^2)^{n/2} = 1$  and  $(\omega_n^2)^i \neq 1$  0 < i < n/2; in other words  $\omega_n^2 = \omega_{n/2}$ . So, we have the identities

$$(W_n \mathbf{z})_i = \sum_{j=1}^{n/2} \omega_{n/2}^{(i-1)(j-1)} z_{2j-1} + \omega_n^{i-1} \sum_{j=1}^{n/2} \omega_{n/2}^{(i-1)(j-1)} z_{2j}, \ i = 1, 2, \dots, n. \quad (?)$$

It follows that, for  $i = 1, \ldots, \frac{n}{2}$ ,

$$(W_{n}\mathbf{z})_{i} = (W_{n/2} \begin{bmatrix} z_{1} \\ z_{3} \\ z_{n-1} \end{bmatrix})_{i} + \omega_{n}^{i-1} (W_{n/2} \begin{bmatrix} z_{2} \\ z_{4} \\ z_{n} \end{bmatrix})_{i}.$$

Moreover, by setting  $i = \frac{n}{2} + k$ ,  $k = 1, \ldots, \frac{n}{2}$ , in (?), we obtain

$$\begin{aligned} (W_n \mathbf{z})_{\frac{n}{2}+k} &= \sum_{j=1}^{n/2} \omega_{n/2}^{\frac{n}{2}(j-1)} \omega_{n/2}^{(k-1)(j-1)} z_{2j-1} + \omega_n^{\frac{n}{2}} \omega_n^{k-1} \sum_{j=1}^{n/2} \omega_{n/2}^{\frac{n}{2}(j-1)} \omega_{n/2}^{(k-1)(j-1)} z_{2j} \\ &= \sum_{j=1}^{n/2} \omega_{n/2}^{(k-1)(j-1)} z_{2j-1} - \omega_n^{k-1} \sum_{j=1}^{n/2} \omega_{n/2}^{(k-1)(j-1)} z_{2j} \\ &= (W_{n/2} \begin{bmatrix} z_1 \\ z_3 \\ z_{n-1} \end{bmatrix})_k - \omega_n^{k-1} (W_{n/2} \begin{bmatrix} z_2 \\ z_4 \\ z_n \end{bmatrix})_k, \ k = 1, \dots, \frac{n}{2}. \end{aligned}$$

 $(\omega_n^{\frac{n}{2}} = -1; \text{ think } \omega = e^{\pm i 2\pi/n}).$  Thus

$$W_{n}\mathbf{z} = \begin{bmatrix} I & D_{1\omega_{n}^{\frac{n}{2}-1}} \\ I & -D_{1\omega_{n}^{\frac{n}{2}-1}} \end{bmatrix} \begin{bmatrix} W_{n/2} & 0 \\ 0 & W_{n/2} \end{bmatrix} Q\mathbf{z},$$

$$D_{1\omega_{n}^{\frac{n}{2}-1}} = \begin{bmatrix} 1 & & & & \\ & \omega_{n} & & \\ & & & \omega_{n}^{\frac{n}{2}-1} \end{bmatrix}, Q = \begin{bmatrix} 1 & & & & \\ 0 & 0 & 1 & & \\ & & & & 1 & 0 \\ 0 & 1 & & & \\ 0 & 0 & 0 & 1 & & \\ & & & & & 0 & 1 \end{bmatrix}.$$

$$(??)$$

If  $c_n$  denotes the complexity of the matrix-vector product  $F_n \mathbf{z}$ , then, by the previous formula,

$$c_n \leq 2c_{n/2} + rn, \ r \text{ constant.}$$

But this implies  $c_n = O(n \log_2 n)$ . The proof of the last assertion is left to the reader.

Of course, any time a  $n \times n$  matrix U, well defined for all n, satisfies for n even an identity of the type

$$U_n = \begin{bmatrix} \text{sparse matrix} \end{bmatrix} \begin{bmatrix} U_{n/2} & 0\\ 0 & U_{n/2} \end{bmatrix} \begin{bmatrix} \text{permutation matrix} \end{bmatrix},$$

the matrix-vector product  $U_n \mathbf{z}$  can be computed in at most  $O(n \log_2 n)$  arithmetic operations. The above identity is verified for at least 10 matrices U, the Fourier transform and its (-1) version, and the eight Hartley-type transforms. Note, however, that there are also other 16 discrete transforms of complexity  $O(n \log_2 n)$ , sine-type and the cosine-type transforms. See [],[].

*Exercise* G. Prove that the  $n \times n$  matrix  $G = G_n$  defined by

$$G_{ij} = \frac{1}{\sqrt{n}} \left( \cos \frac{(2i+1)(2j+1)\pi}{2n} + \sin \frac{(2i+1)(2j+1)\pi}{2n} \right), \ i, j = 0, \dots, n-1,$$

is symmetric, persymmetric, real, unitary, and satisfies the identity:

$$G_{n} = \frac{1}{\sqrt{2}} \begin{bmatrix} R_{+} & R_{-} \\ -R_{-}J & R_{+}J \end{bmatrix} \begin{bmatrix} G_{n/2} & 0 \\ 0 & G_{n/2} \end{bmatrix} Q, \ R_{\pm} = D_{c} \pm D_{s}J,$$

 $(J \frac{n}{2} \times \frac{n}{2} \text{ anti-identity})$  for some suitable  $\frac{n}{2} \times \frac{n}{2}$  diagonal matrices  $D_c$ ,  $D_s$ . Prove, moreover, that each row of  $G_n$  has at least a zero entry when n = 2 + 4s (this is like to say, we will see, that the space  $\{Gd(\mathbf{z})G : \mathbf{z} \in \mathbb{C}^n\}$  is not a *h*-space for such values of *n*); and that, instead, for all other n,  $[G_n]_{1k} \neq 0 \forall k$  (i.e., for all other *n*, the space  $\{Gd(\mathbf{z})G : \mathbf{z} \in \mathbb{C}^n\}$  is a 1-space).

*Exercise.* Prove that the space  $C_{-1}^S + JC_{-1}^{SK}$ ,  $C_{-1}^S$  symmetric  $n \times n$  (-1)-circulants,  $C_{-1}^{SK}$  skewsymmetric (-1)-circulants, is a commutative matrix algebra (a matrix A is skewsymmetric if  $A^T = -A$ ).

The sine matrix and the (commutative) algebra of  $\tau$  matrices

Consider the  $n \times n$  matrix

$$P_0 + P_0^T = \begin{bmatrix} 0 & 1 & & \\ 1 & 0 & 1 & & \\ 0 & 1 & 0 & & \\ & & & 1 & \\ & & & 1 & 0 \end{bmatrix},$$

and set  $J_1 = I$ , and  $J_2 = P_0 + P_0^T$ . Note that  $\mathbf{e}_1^T J_1 = \mathbf{e}_1^T$ ,  $\mathbf{e}_1^T J_2 = \mathbf{e}_2^T$ . Moreover, since

we have  $\mathbf{e}_1^T((P_0 + P_0^T)^2 - I) = [0 \ 0 \ 1 \ 0 \ \cdots \ 0] = \mathbf{e}_3^T$ . Set  $J_3 = (P_0 + P_0^T)^2 - I = J_2(P_0 + P_0^T) - J_1$ ; then  $\mathbf{e}_1^T J_3 = \mathbf{e}_3^T$ .

More in general, set  $J_{i+1} = J_i(P_0 + P_0^T) - J_{i-1}$ , i = 2, 3, ..., n-1. The matrix  $J_{i+1}$  is a polynomial in  $P_0 + P_0^T$  of degree *i* with the property  $\mathbf{e}_1^T J_{i+1} = \mathbf{e}_{i+1}^T$ . proof: assume  $\mathbf{e}_1^T J_j = \mathbf{e}_i^T$ , j = 1, ..., i; then

$$\mathbf{e}_{1}^{T}J_{i+1} = \mathbf{e}_{1}^{T}(J_{i}(P_{0}+P_{0}^{T})-J_{i-1}) = (\mathbf{e}_{i}^{T}(P_{0}+P_{0}^{T}))-\mathbf{e}_{i-1}^{T} = (\mathbf{e}_{i-1}^{T}+\mathbf{e}_{i+1}^{T})-\mathbf{e}_{i-1}^{T} = \mathbf{e}_{i+1}^{T}$$

Since  $J_1, J_2, \ldots, J_n$  are linearly independent, we can say that they span the set  $\{p(P_0 + P_0^T)\}$  of all polynomials in the matrix  $P_0 + P_0^T$  (use the Cailey-Hamilton theorem). We call such set  $\tau$ . Note that the matrices of  $\tau$  are determined once their first row is known; with the symbol  $\tau(\mathbf{a})$  we denote the matrix of  $\tau$  whose first row is  $\mathbf{a}^T$ , i.e. the matrix  $\sum_k a_k J_k$ .

Let us find a useful representation of  $\tau$  and, in particular, of  $\tau(\mathbf{a})$ . First

observe that the following vectorial equalities hold:

$$\begin{bmatrix} 1 & & \\ 1 & 0 & 1 & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} \sin \frac{j\pi}{n+1} \\ \sin \frac{2j\pi}{n+1} \\ \sin \frac{nj\pi}{n+1} \end{bmatrix} = 2\cos \frac{j\pi}{n+1} \begin{bmatrix} \sin \frac{j\pi}{n+1} \\ \sin \frac{2j\pi}{n+1} \\ \sin \frac{nj\pi}{n+1} \end{bmatrix}, \ j = 1, \dots, n.$$

Such n equalities can be rewritten as a simple matrix identity  $(P_0 + P_0^T)S = SD$ where S is the matrix

$$S_{ij} = \sqrt{\frac{2}{n+1}} \sin \frac{ij\pi}{n+1}, \ i, j = 1, \dots, n,$$

and D is the diagonal matrix with diagonal entries  $D_{jj} = 2 \cos \frac{j\pi}{n+1}$ . Note that the matrix S, called sine matrix, is real, symmetric and unitary (prove it!).

*Remark.* Let  $F_{2(n+1)}$  be the Fourier matrix of order 2(n+1). Then the sine matrix S satisfies the following relation:

$$\mathbf{i}(I - F_{2(n+1)}^2)F_{2(n+1)} = \begin{bmatrix} 0 & \mathbf{0}^T & 0 & \mathbf{0}^T \\ \mathbf{0} & S & \mathbf{0} & -SJ \\ 0 & \mathbf{0}^T & 0 & \mathbf{0}^T \\ \mathbf{0} & -JS & \mathbf{0} & JSJ \end{bmatrix}$$

(note that  $F_{2(n+1)}^2$  is a permutation matrix). As a consequence, a sine transform can be computed by performing a discrete Fourier transform.

So, the columns of the sine matrix S form a system of unitarily orthonormal eigenvectors for the matrix  $P_0 + P_0^T$ . In other words, the unitary matrix S diagonalizes  $P_0 + P_0^T$  and, of course, diagonalizes any polynomial in  $P_0 + P_0^T$ , i.e. any  $\tau$  matrix:

$$P_0 + P_0^T = SDS, \quad (P_0 + P_0^T)^k = SD^kS, \tau = \{p(P_0 + P_0^T)\} = \{\sum_{k=1}^n a_k(P_0 + P_0^T)^{k-1} : a_k \in \mathbb{C}\} = \{Sd(\mathbf{z})S : \mathbf{z} \in \mathbb{C}^n\}.$$

In particular, it is clear that the matrix of  $\tau$  with first row  $\mathbf{a}^T$  is

$$\tau(\mathbf{a}) = \sum_{k=1}^{n} a_k J_k = Sd(S^T \mathbf{a})d(S^T \mathbf{e}_1)^{-1}S^{-1}.$$

The latter formula states that matrix-vector products involving  $\tau$  matrices have complexity at most  $O(n \log_2 n)$ .

Proposition PC. Given  $X \in \mathbb{C}^{n \times n}$  generic, we have

$$\{p(X)\} \subset \{A \in \mathbb{C}^{n \times n} : AX = XA\},\\ \dim(\{p(X)\}) \le n \le \dim\{A \in \mathbb{C}^{n \times n} : AX = XA\}$$

and if one equality holds then the other equality holds too. So, X is non derogatory if and only if  $\dim(\{p(X)\}) = n = \dim\{A \in \mathbb{C}^{n \times n} : AX = XA\}.$ 

The above Proposition suggests the following further representation of the space  $\tau {:}$ 

$$\tau = \{ A \in \mathbb{C}^{n \times n} : A(P_0 + P_0^T) = (P_0 + P_0^T)A \}.$$

The fact that any matrix of  $\tau$  must commute with  $P_0 + P_0^T$  is equivalent to require that the following  $n^2$  cross-sum conditions hold:

$$a_{i,j-1} + a_{i,j+1} = a_{i-1,j} + a_{i+1,j}, \ i, j = 1, \dots, n$$

where we have set  $a_{0,j} = a_{n+1,j} = a_{i,0} = a_{i,n+1} = 0$ , i, j = 1, ..., n. We can use such conditions in order to write down the generic  $\tau$  matrix whose first row is  $[a_1 a_2 \cdots a_n]$ . For example, for n = 4, we can say that

$$\tau(\mathbf{a}) = \begin{bmatrix} a_1 & a_2 & a_3 & a_4 \\ a_2 & a_1 + a_3 & a_2 + a_4 & a_3 \\ a_3 & a_2 + a_4 & a_1 + a_3 & a_2 \\ a_4 & a_3 & a_2 & a_1 \end{bmatrix}, \ J_1 = I, \ J_2 = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix},$$
$$J_3 = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}, \ J_4 = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix}, \ J_2 - J_4 = \begin{bmatrix} 0 & 1 & 0 & -1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ -1 & 0 & 1 & 0 \end{bmatrix},$$

and so on.

*Exercise.* Prove that for n even the matrix  $J_2$  is invertible, and, if possible, compute the inverse.

solution: We know that if  $J_2$  is invertible, then  $J_2^{-1} \in \tau$  (from the fact that  $J_2$  commutes with  $P_0 + P_0^T$  it follows that also  $J_2^{-1}$  commutes with  $P_0 + P_0^T$ !). Thus  $J_2^{-1} = \tau(\mathbf{z})$  for some  $\mathbf{z} \in \mathbb{C}^n$ . Note that the matrix identity  $\tau(\mathbf{z})J_2 = I$  is equivalent to the vectorial identity  $\mathbf{z}^T J_2 = \mathbf{e}_1^T$ . So, for example, for n = 4 we have the condition

$$\begin{bmatrix} z_1 \ z_2 \ z_3 \ z_4 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 1 \ 0 \ 0 \ 0 \end{bmatrix},$$

which yields  $z_1 = 0$ ,  $z_2 = 1$ ,  $z_3 = 0$ ,  $z_4 = -1$ , and thus

$$J_2^{-1} = \begin{bmatrix} 0 & 1 & 0 & -1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ -1 & 0 & 1 & 0 \end{bmatrix} = J_3 - J_4$$

The proof for  $n = 6, 8, \ldots$  is left to the reader.

*Exercise* Ttau. Let T be a symmetric Toeplitz  $n \times n$  matrix, i.e.  $T = (t_{|i-j|})_{i,j=1}^n$ , for some  $t_k \in \mathbb{C}$ . Show that T = A + B where A is a  $\tau$  matrix of order n and

$$B = \begin{bmatrix} 0 & 0 & 0 \\ 0 & R & 0 \\ 0 & 0 & 0 \end{bmatrix}, \ R \in \tau, \ R \in \mathbb{C}^{(n-2) \times (n-2)}.$$

*Exercise.* Write down rank one  $\tau$  matrices (try first n = 2, n = 3, n = 4, n = 5, n = 6, ...)

Hessenberg algebras

Let X be a lower Hessenberg  $3 \times 3$  matrix

$$X = \begin{bmatrix} a_{11} & b_1 & 0\\ a_{21} & a_{22} & b_2\\ a_{31} & a_{32} & a_{33} \end{bmatrix}.$$

We now show that the space of all polynomials in X is spanned by three matrices  $J_1, J_2, J_3$  such that  $\mathbf{e}_1^T J_k = \mathbf{e}_k^T$ , k = 1, 2, 3, provided that  $X_{ii+1} \neq 0$ , i = 1, 2. As  $J_1$  we take the identity,  $J_1 = X^0 = I$ . Let us define  $J_2$ :

$$X - a_{11}I = \begin{bmatrix} 0 & b_1 & 0 \\ a_{21} & a_{22} - a_{11} & b_2 \\ a_{31} & a_{32} & a_{33} - a_{11} \end{bmatrix},$$

 $b_1 \neq 0 \Rightarrow$ 

$$J_2 = \frac{1}{b_1}(X - a_{11}I) = \begin{bmatrix} 0 & 1 & 0 \\ a_{21}/b_1 & (a_{22} - a_{11})/b_1 & b_2/b_1 \\ a_{31}/b_1 & a_{32}/b_1 & (a_{33} - a_{11})/b_1 \end{bmatrix}.$$

Note that  $\mathbf{e}_1^T J_2 = \mathbf{e}_2^T$ . Then, let us define  $J_3$ :

$$J_2^2 = \begin{bmatrix} a_{22}/b_1 & (a_{22} - a_{11})/b_1 & b_2/b_1 \\ & & & \\ & &$$

 $b_2 \neq 0 \Rightarrow$ 

$$J_3 = \frac{b_1}{b_2} (J_2^2 - \frac{a_{22}}{b_1}I - \frac{a_{22} - a_{11}}{b_1}J_2).$$

Note that  $\mathbf{e}_1^T J_3 = \mathbf{e}_3^T$ . Finally, Cailey-Hamilton theorem yields the thesis,  $\{p(X)\} =$ Span $\{J_1, J_2, J_3\}.$ 

The following Proposition generalizes to a generic n the above remarks. For a detailed proof see [].

*Proposition.* Let X be a lower Hessenberg  $n \times n$  matrix. Then the space  $H_X$  of all polynomials in X

$$H_X = \{\sum_{k=1}^n \alpha_k X^{k-1} : \alpha_k \in \mathbb{C}\}$$

is spanned by *n* matrices  $J_1, \ldots, J_n$  such that  $\mathbf{e}_1^T J_k = \mathbf{e}_k^T$ ,  $k = 1, \ldots, n$ , provided that  $X_{ii+1} \neq 0$ ,  $i = 1, \ldots, n-1$ ; in such case  $H_X$  is called Hessenberg algebra, and for any  $\mathbf{a} \in \mathbb{C}^n$  there is a unique matrix in  $H_X$  with first row  $\mathbf{a}^T$  which is denoted by  $H_X(\mathbf{a})$ , i.e.  $H_X(\mathbf{a}) = \sum_k a_k J_k$ .

Of course, by Proposition PC, any Hessenberg algebra  $H_X$  admits also the following representation

$$H_X = \{ A \in \mathbb{C}^{n \times n} : AX = XA \}.$$

Until now we have seen two examples of Hessenberg algebras,  $\xi$ -circulants  $\xi \neq 0$  (X = P<sub>\xi</sub>) and tau matrices (X = P<sub>0</sub> + P<sub>0</sub><sup>T</sup>). Both can be simultaneously

diagonalized by a suitable matrix. An example of Hessenberg algebra whose matrices cannot be simultaneously diagonalized is  $H_{P_0}$ , the space of all upper triangular Toeplitz matrices. The matrix  $H_{P_0}(\mathbf{a})$  is displayed here below

$$H_{P_0}(\mathbf{a}) = \begin{bmatrix} a_1 & a_2 & & a_n \\ & a_1 & a_2 & & \\ & & \ddots & \ddots & \\ & & & \ddots & a_2 \\ & & & & & a_1 \end{bmatrix}.$$

Even the matrix  $P_0$ , i.e. the matrix generating the space, is not diagonalizable. (We shall see, however, that  $H_{P_0}$  can be embedded in the space of  $2n \times 2n$  circulants, which are diagonalizable).

Note that when  $X = P_0$  or, more in general, when  $X = P_{\xi}$ , the matrices  $J_k$  are simply the powers of X, i.e.  $J_k = X^{k-1}$ . For example, for n = 3

$$J_1 = I, \ J_2 = X = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ \xi & 0 & 0 \end{bmatrix}, \ J_3 = X^2 = \begin{bmatrix} 0 & 0 & 1 \\ \xi & 0 & 0 \\ 0 & \xi & 0 \end{bmatrix}.$$

Hessenberg algebras make up a subclass of commutative matrix algebras of the class of 1-spaces defined here below

Definition.  $\mathcal{L} \subset \mathbb{C}^{n \times n}$  is said to be a 1-space if  $\mathcal{L} = \text{Span} \{J_1, \ldots, n\}$  with  $J_k$  such that  $\mathbf{e}_1^T J_k = \mathbf{e}_k^T$ ,  $k = 1, \ldots, n$ .

An example of 1-space  $\mathcal{L}$  which is not a Hessenberg algebra is the following

$$\mathcal{L} = \{ \begin{bmatrix} A & JB \\ JB & A \end{bmatrix} : A, B \ \frac{n}{2} \times \frac{n}{2} \text{ circulants} \}.$$

One can easily prove that  $\mathcal{L}$  is a non commutative matrix algebra. An example of 1-space which is not a matrix algebra is the set of all  $n \times n$  symmetric Toeplitz matrices (see the next section).

### Toeplitz linear systems and displacement decompositions

A  $n \times n$  Toeplitz matrix is a matrix of the form  $T = (t_{i-j})_{i,j=1}^n$ . In applications often one has to solve Toeplitz linear systems  $T\mathbf{x} = \mathbf{b}$ .

For example, here below is a  $3 \times 3$  Toeplitz matrix:

$$T = \left[ \begin{array}{ccc} t_0 & t_{-1} & t_{-2} \\ t_1 & t_0 & t_{-1} \\ t_2 & t_1 & t_0 \end{array} \right].$$

An example of Toeplitz matrix T is the coefficient matrix of the linear system arising when solving, by finite differences or by finite elements, the boundary value differential problem  $-u^{"} = f$ ,  $u(a) = \alpha$ ,  $u(b) = \beta$ . In such case T is symmetric and its first row is  $[2 - 1 \ 0 \ \cdots \ 0]$ . Another example, important is applied probability, is  $T = (t^{|i-j|})_{i,j=1}^{n}$  with |t|  $(t \in \mathbb{C})$  less than 1.

*Exercise.* The vector space of all  $n \times n$  symmetric Toeplitz matrices is a 1-space, being  $(t_{|i-j|})_{i,j=1}^n$  equal to the sum  $\sum_{k=1}^n t_{k-1}J_k$  where the  $J_k$  are the symmetric

To eplitz matrices with first row  $\mathbf{e}_k^T$ . Prove that such 1-space is not a matrix algebra.

In the framework of displacement theory it is possible to obtain some decompositions of  $T^{-1}$  involving matrices from Hessenberg algebras (or from more general commutative *h*-spaces) of the type

$$T^{-1} = \sum_{k=1}^{\alpha} M_k N_k, \ 2 \le \alpha \le 4.$$

Usually, the matrices  $M_k$  and  $N_k$ , appearing in such formulas, can be multiplied by a vector in  $O(n \log_2 n)$  arithmetic operations; thus fast direct solvers of Toeplitz linear system naturally arise. Here below there is one example of such formulas:

$$T^{-1} = L_1 U_1 + L_2 U_2. (GS)$$

The  $L_j$  and  $U_j$  are suitable lower and upper triangular Toeplitz matrices, i.e. elements or transposed of elements from the Hessenberg algebra  $H_{P_0}$ .

*Remark.* By the Gohberg-Semencul formula (GS), if the  $L_j$  and  $U_j$  are known (a way to obtain them is indicated in []), then the matrix-vector product  $T^{-1}\mathbf{b}$  can be computed in at most  $O(n\log_2 n)$  arithmetic operations. That is, assuming preprocessing on T, the complexity of the problem of solving any Toeplitz system  $T\mathbf{x} = \mathbf{b}$  is at most  $O(n\log_2 n)$ .

proof: it is enough to prove that any Toeplitz matrix (in particular the triangular ones) can be multiplied by a vector by means of a finite number of discrete Fourier transforms. The latter result is immediate if we observe that any  $n \times n$ Toeplitx matrix can be embedded into a  $(2n + k) \times (2n + k)$ ,  $k \ge 0$ , circulant matrix; for example, if n = 3 we have

$$T = \begin{bmatrix} t_0 & t_{-1} & t_{-2} \\ t_1 & t_0 & t_{-1} \\ t_2 & t_1 & t_0 \end{bmatrix}, C = \begin{bmatrix} t_0 & t_{-1} & t_{-2} & t_2 & t_1 \\ t_1 & t_0 & t_{-1} & t_{-2} & t_2 \\ t_2 & t_1 & t_0 & t_{-1} & t_{-2} \\ t_2 & t_1 & t_0 & t_{-1} & t_{-2} \\ t_{-2} & t_2 & t_1 & t_0 & t_{-1} \\ t_{-1} & t_{-2} & t_2 & t_1 & t_0 \end{bmatrix}$$

(k = 0). It is clear that the vector  $T \cdot \mathbf{z}$  is the first part of the vector  $C \begin{bmatrix} \mathbf{z} \\ \mathbf{0} \end{bmatrix}$ .

The above representation (GS) for the inverse of a Toeplitz matrix, which can be very useful in order to solve Toeplitz linear systems efficiently [], follows from the *displacement decomposition* formula stated in the following theorem

Theorem DD. Let X be a lower Hessenberg  $n \times n$  matrix. Assume that  $X_{ij} = X_{i+1,j+1}$ ,  $i, j = 1, \ldots, n-1$  (X has Toeplitz structure), and that  $b = X_{ii+1} \neq 0$ , and consider the (commutative) Hessenberg algebra  $H_X$  generated by X (note that  $H_X$  is a 1-space).

Assume that  $A \in \mathbb{C}^{n \times n}$  is such that  $AX - XA = \sum_{m=1}^{\alpha} \mathbf{x}_m \mathbf{y}_m^T$ . Then

$$bA = -\sum_{m=1}^{\alpha} H_{P_0}(\tilde{\mathbf{x}}_m)^T H_X(\mathbf{y}_m) + bH_X(A^T \mathbf{e}_1)$$
(DD)

where, for  $\mathbf{z} \in \mathbb{C}^n$ ,  $H_{P_0}(\mathbf{z})$  is the upper triangular Toeplitz matrix with first row  $\mathbf{z}^T$ ,  $H_X(\mathbf{z})$  is the matrix of  $H_X$  with first row  $\mathbf{z}^T$ , and  $\tilde{\mathbf{z}}$  is the vector  $[0 \, z_1 \cdots z_{n-1}]^T$ .

Note: Besides DD several other displacement decompositions hold, which can be general like DD, i.e. representing generic matrices A, or specialized for centrosymmetric A (see []). Such decompositions yield formulas for the inverses of Toeplitz, Toeplitz plus Hankel, and Toeplitz plus Hankel-like matrices useful in order to solve Toeplitz plus Hankel-like linear systems. Recall that a Hankel matrix is nothing else a matrix of the form JT where T is Toeplitz (the well known Hilbert matrix is an example of Hankel matrix).

In order to prove Theorem DD, the following Lemma is fundamental.

Lemma []. Let  $\mathcal{L}$  be a commutative 1-space of  $n \times n$  matrices, i.e.  $\mathcal{L} = \{\sum_k \alpha_k J_k : \alpha_k \in \mathbb{C}\}$ , with  $J_k \in \mathbb{C}^{n \times n}$  such that  $\mathbf{e}_1^T J_k = \mathbf{e}_k^T$ ,  $J_k J_s = J_s J_k$ . Let X be an element of  $\mathcal{L}$ , and assume that  $A \in \mathbb{C}^{n \times n}$  is such that  $AX - XA = \mathbf{x}\mathbf{y}^T$ . Then  $\mathbf{x}^T \mathcal{L}(\mathbf{y})^T = \mathbf{0}^T$ .

proof: note that the equality  $J_k J_s = J_s J_k$  implies  $\mathbf{e}_1^T J_k J_s = \mathbf{e}_1^T J_s J_k$ ,  $\mathbf{e}_k^T J_s = \mathbf{e}_s^T J_k \forall s, k$ , thus

$$\begin{aligned} \mathbf{x}^{T} \mathcal{L}(\mathbf{y})^{T} \mathbf{e}_{r} &= \mathbf{x}^{T} (\sum_{k} y_{k} J_{k})^{T} \mathbf{e}_{r} = \mathbf{x}^{T} \sum_{k} y_{k} J_{k}^{T} \mathbf{e}_{r} \\ &= \mathbf{x}^{T} \sum_{k} y_{k} J_{r}^{T} \mathbf{e}_{k} = \mathbf{x}^{T} J_{r}^{T} \sum_{k} y_{k} \mathbf{e}_{k} \\ &= \mathbf{x}^{T} J_{r}^{T} \mathbf{y} = \sum_{i,j} x_{i} y_{j} [J_{r}^{T}]_{ij} \\ &= \sum_{i,j} x_{i} y_{j} [J_{r}]_{ji} = \sum_{i,j} [AX - XA]_{ij} [J_{r}]_{ji} \\ &= \sum_{i} [(AX - XA)J_{r}]_{ii} = \sum_{i} [(AJ_{r})X - X(AJ_{r})]_{ii} \\ &= \operatorname{tr} \left( (AJ_{r})X \right) - \operatorname{tr} \left( X(AJ_{r}) \right) = 0 \end{aligned}$$

(recall that the two matrices MN and NM,  $M, N \in \mathbb{C}^{n \times n}$ , have the same characteristic polynomial, even if (in case  $\det(M) = \det(N) = 0$ ) MN and NM might be not similar each other).

We now report a draft of the proof of Theorem DD (for a more detailed proof see []). In order to obtain the equality (DD), which is of the type

$$bA = E + bH_X(A^T \mathbf{e}_1),$$

it is enough to prove that

$$EX - XE = (bA)X - X(bA), \qquad (***)$$

and to observe that the first row of E is null. In fact, the above equality implies (bA - E)X - X(bA - E) = 0, and thus  $bA - E \in H_X$ . The Lemma, applied for  $\mathcal{L} = H_X$ , is fundamental in proving (\*\*\*).

A matrix algebra which is not a 1-space: the class of spaces in  $\mathbb{V}$ 

Remember that a matrix A is said symmetric if  $A^T = A$   $(a_{ji} = a_{ij})$ , skewsymmetric if  $A^T = -A$   $(a_{ji} = -a_{ij})$  and persymmetric if  $A^T = JAJ$   $(a_{ji} = a_{n+1-i,n+1-j})$ .

Consider a  $6 \times 6$  symmetric (-1)-circulant matrix  $A \in C_{-1}^S$ ,

$$A = \begin{bmatrix} a_0 & a_1 & a_2 & 0 & -a_2 & -a_1 \\ a_1 & a_0 & a_1 & a_2 & 0 & -a_2 \\ a_2 & a_1 & a_0 & a_1 & a_2 & 0 \\ 0 & a_2 & a_1 & a_0 & a_1 & a_2 \\ -a_2 & 0 & a_2 & a_1 & a_0 & a_1 \\ -a_1 & -a_2 & 0 & a_2 & a_1 & a_0 \end{bmatrix},$$

a 6 × 6 skewsymmetric (-1)-circulant matrix  $B \in C_{-1}^{SK}$ , and the matrix JB,

$$B = \begin{bmatrix} 0 & b_1 & b_2 & b_3 & b_2 & b_1 \\ -b_1 & 0 & b_1 & b_2 & b_3 & b_2 \\ -b_2 & -b_1 & 0 & b_1 & b_2 & b_3 \\ -b_3 & -b_2 & -b_1 & 0 & b_1 & b_2 \\ -b_2 & -b_3 & -b_2 & -b_1 & 0 & b_1 \\ -b_1 & -b_2 & -b_3 & -b_2 & -b_1 & 0 \end{bmatrix}, \ JB = \begin{bmatrix} -b_1 & -b_2 & -b_3 & -b_2 & -b_1 & 0 \\ -b_2 & -b_3 & -b_2 & -b_1 & 0 & b_1 \\ -b_3 & -b_2 & -b_1 & 0 & b_1 & b_2 \\ -b_2 & -b_1 & 0 & b_1 & b_2 & b_3 \\ -b_1 & 0 & b_1 & b_2 & b_3 & b_2 \\ 0 & b_1 & b_2 & b_3 & b_2 & b_1 \end{bmatrix}.$$

The vector space  $\gamma$  of all matrices of the type A + JB has dimension equal to 6 (recall that  $\dim(A + JB) = \dim(A) + \dim(JB) - \dim(A \cap JB)$ ).

We now show that there is not a basis  $\{J_k\}$  for  $\gamma$  such that  $\mathbf{e}_1^T J_k = \mathbf{e}_k^T$ , k = 1, 2, 3, 4, 5, 6, i.e.  $\gamma$  is not a 1-space. Note that

$$\mathbf{e}_1^T(A+JB) = [(a_0 - b_1)(a_1 - b_2)(a_2 - b_3)(-b_2)(-a_2 - b_1)(-a_1)],$$

so, the equality  $\mathbf{e}_1^T(A + JB) = \mathbf{e}_2^T$  is satisfied if and only if

$$a_0 - b_1 = 0, a_1 - b_2 = 1, a_2 - b_3 = 0, -b_2 = 0, -a_2 - b_1 = 0, -a_1 = 0.$$

Since we have both the conditions  $b_2 = 0$  and  $b_2 = -1$ , a matrix  $J_2 \in \gamma$  such that  $\mathbf{e}_1^T J_2 = \mathbf{e}_2^T$  cannot exist.

However, there exists a basis  $\{J_k\}$  of  $\gamma$  satisfying the equalities  $(\mathbf{e}_1 + \mathbf{e}_6)^T J_k = \mathbf{e}_k^T$ , k = 1, 2, 3, 4, 5, 6. For example, a matrix  $J_2 \in \gamma$  with the property  $(\mathbf{e}_1 + \mathbf{e}_6)^T J_2 = \mathbf{e}_2^T$  is obtained as follows. Note that

$$\mathbf{e}_6^T(A+JB) = [(-a_1)(-a_2+b_1)(b_2)(a_2+b_3)(a_1+b_2)(a_0+b_1)],$$

so, for the sum of the first and of the sixth rows of A + JB, we obtain the formula

$$(\mathbf{e}_1 + \mathbf{e}_6)^T (A + JB) = [(a_0 - b_1 - a_1)(a_1 - b_2 - a_2 + b_1)(a_2 - b_3 + b_2)(-b_2 + a_2 + b_3)(-a_2 - b_1 + a_1 + b_2)(-a_1 + a_0 + b_1)]$$

Thus the condition  $(\mathbf{e}_1 + \mathbf{e}_6)^T (A + JB) = \mathbf{e}_2^T$  is satisfied if and only if the following system of equations has solution

$$a_0 - b_1 - a_1 = 0, \quad a_1 - b_2 - a_2 + b_1 = 1, a_2 - b_3 + b_2 = 0, \quad -b_2 + a_2 + b_3 = 0, -a_2 - b_1 + a_1 + b_2 = 0, \quad -a_1 + a_0 + b_1 = 0.$$

and such system has the unique solution  $a_0 = a_1 = \frac{1}{2}$ ,  $a_2 = 0$ ,  $b_2 = b_3 = -\frac{1}{2}$ ,

 $b_1 = 0$ . The matrix  $J_2 \in \gamma$  such that  $(\mathbf{e}_1 + \mathbf{e}_6)^T J_2 = \mathbf{e}_2^T$  is displayed here below:

$$J_2 = \begin{bmatrix} \frac{1}{2} & 1 & \frac{1}{2} & \frac{1}{2} & 0 & -\frac{1}{2} \\ 1 & 1 & 1 & 0 & 0 & 0 \\ \frac{1}{2} & 1 & \frac{1}{2} & \frac{1}{2} & 0 & -\frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 0 & -\frac{1}{2} \\ 0 & 0 & 0 & 0 & 0 & 0 \\ -\frac{1}{2} & 0 & -\frac{1}{2} & -\frac{1}{2} & 0 & \frac{1}{2} \end{bmatrix}$$

Analogously, one obtains the other  $J_k$  such that  $(\mathbf{e}_1 + \mathbf{e}_6)^T J_k = \mathbf{e}_k^T$ :

$$\begin{split} J_1 &= \begin{bmatrix} 1 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & 0 \\ \frac{1}{2} & 1 & \frac{1}{2} & \frac{1}{2} & 0 & -\frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 1 & 0 & -\frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 & -\frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & 0 & -\frac{1}{2} & -\frac{1}{2} & 0 & -\frac{1}{2} \\ 0 & -\frac{1}{2} & -\frac{1}{2} & -\frac{1}{2} & 0 & -\frac{1}{2} \\ 0 & -\frac{1}{2} & -\frac{1}{2} & -\frac{1}{2} & -\frac{1}{2} & 0 \end{bmatrix}, \\ J_3 &= \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & 1 & 0 & -\frac{1}{2} & -\frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & 0 \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 0 \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 0 & \frac{1}{2} \\ -\frac{1}{2} & -\frac{1}{2} & 0 & 0 & \frac{1}{2} \\ -\frac{1}{2} & 0 & 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & 0 & -\frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & 0 & -\frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & 0 & -\frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & 0 & -\frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & 0 & -\frac{1}{2} \\ 0 & 0 & 0 & 0 & 0 & 0 \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 0 & -\frac{1}{2} \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & 0 \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 0 & \frac{1}{2} \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & 0 \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 0 & \frac{1}{2} \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 1 & 1 \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 1 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 1 & 1 \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 1 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 1 & 1 \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 1 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 1 & 1 \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 1 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 1 & 1 \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 1 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 1 & 1 \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 1 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 1 & 1 \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 1 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 1 & 1 \\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 1 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & \frac{1}{2}$$

,

$$J_6 = \begin{bmatrix} 0 & -\frac{1}{2} & -\frac{1}{2} & -\frac{1}{2} & -\frac{1}{2} & 0\\ -\frac{1}{2} & 0 & -\frac{1}{2} & -\frac{1}{2} & 0 & \frac{1}{2}\\ -\frac{1}{2} & -\frac{1}{2} & 0 & 0 & \frac{1}{2} & \frac{1}{2}\\ -\frac{1}{2} & -\frac{1}{2} & 0 & 1 & \frac{1}{2} & \frac{1}{2}\\ -\frac{1}{2} & 0 & \frac{1}{2} & \frac{1}{2} & 1 & \frac{1}{2}\\ 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & 1 \end{bmatrix}$$

 $\begin{array}{l} (a_1 = a_2 = 0, \, a_0 = \frac{1}{2}, \, b_1 = b_2 = b_3 = \frac{1}{2}).\\ \text{Of course } \gamma = \text{Span} \left\{ J_1, J_2, J_3, J_4, J_5, J_6 \right\} \text{ (the } J_k \text{ are linearly independent!).}\\ \text{The matrix } \sum_k a_k J_k \text{ is denoted by } \gamma(\mathbf{a}). \text{ Note that } (\mathbf{e}_1 + \mathbf{e}_6)^T \gamma(\mathbf{a}) = \mathbf{a}^T, \text{ so } \gamma(\mathbf{a}) \text{ is called the matrix of } \gamma \text{ whose } (\mathbf{e}_1 + \mathbf{e}_6)\text{-row is } \mathbf{a}^T. \end{array}$ 

More in general, one can easily prove that the set  $\gamma = C_{-1}^S + J C_{-1}^{SK}, C_{-1}^S n \times n$  symmetric (-1)-circulants,  $C_{-1}^{SK} n \times n$  skewsymmetric (-1)-circulants, is a vector space of dimension n, and is a commutative matrix algebra. Moreover, it is a 1-space if and only if n is one of the integers  $\{3, 4, 5, 7, 8, 9, 11, 12, 13, 15, \ldots\}$ ; for the remaining values of n, i.e. for  $n = 2 + 4s \ s \in \mathbb{Z}$ , no row of a matrix A + JB of  $\gamma$  determines A + JB, that is, there is no index h for which there exists a basis  $\{J_k\}$  of  $\gamma$  with the property  $\mathbf{e}_h^T J_k = \mathbf{e}_k^T$ . Instead, for all *n* the sum of the first and of the *n*th row of A + JB determines A + JB, i.e. there exists a basis  $\{J_k\}$  of  $\gamma$  with the property  $(\mathbf{e}_1 + \mathbf{e}_n)^T J_k = \mathbf{e}_k^T$ .

The matrices of  $\gamma$  can be simultaneously diagonalized by a fast discrete transform. More precisely, for any value of n the following equality holds:

$$\gamma = \{Gd(\mathbf{z})G^{-1} : \mathbf{z} \in \mathbb{C}^n\},\$$
  
$$G_{ij} = \frac{1}{\sqrt{n}} \left(\cos\frac{(2i-1)(2j-1)\pi}{2n} + \sin\frac{(2i-1)(2j-1)\pi}{2n}\right), \ i, j = 1, \dots, n$$

(see []). Note that the matrix G is real, symmetric, persymmetric and unitary. The fact that the matrix-vector product  $G\mathbf{z}$  can be computed in at most  $O(n \log_2 n)$  arithmetic operations follows from the representation of  $G_n$ ,  $G_n := G$ , stated in Exercise G.

Note that for n = 6 the matrix G has ten zeros among its entries, and that these zeros are positioned as follows:

Thus each of the vectors  $G^T \mathbf{e}_k$ , k = 1, ..., 6, has at least one zero entry, i.e. the matrix  $d(G^T \mathbf{e}_k)^{-1}$  is never well defined. The latter assertion is yet true whenever  $n = 2 + 4s \ s \in \mathbb{Z}$ . In other words,  $\gamma$  can be represented as

$$\gamma = \{ Gd(G^T \mathbf{z})d(G^T \mathbf{e}_k)^{-1}G^{-1} : \mathbf{z} \in \mathbb{C}^n \}.$$

if and only  $n \neq 2 + 4s \ s \in \mathbb{Z}$  (for such n one can choose k = 1).

However, as one may guess from the above discussion (detailed, for n = 6), it can be easily shown that  $[G^T(\mathbf{e}_1 + \mathbf{e}_n)]_k \neq 0 \ \forall k$  and  $\forall n$ . So, we have the following representation for  $\gamma$ 

$$\gamma = \{Gd(G^T\mathbf{z})d(G^T(\mathbf{e}_1 + \mathbf{e}_n))^{-1}G^{-1} : \mathbf{z} \in \mathbb{C}^n\}$$

valid for all n (also for n = 2 + 4s where  $d(G^T \mathbf{e}_k)$  are not invertible). The latter formula confirms the fact that the matrices of  $\gamma$  are uniquely defined by the sum of their 1st and nth rows; in particular, since the sum of the first and of the nth row of the matrix  $Gd(G^T \mathbf{a})d(G^T(\mathbf{e}_1 + \mathbf{e}_n))^{-1}G^{-1}$  is equal to  $\mathbf{a}^T$ , we can say that such matrix is exactly the matrix  $\gamma(\mathbf{a})$  (already defined above for n = 6), i.e. the matrix of  $\gamma$  whose  $(\mathbf{e}_1 + \mathbf{e}_n)$ -row is  $\mathbf{a}^T$ .

It is now natural to introduce a class of spaces which include (besides the 1-spaces, like the Hessenberg algebras) also spaces like the algebra  $\gamma$ .

Definition. A subset  $\mathcal{L}$  of  $\mathbb{C}^{n \times n}$  is said to be a space in  $\mathbb{V}$ , if  $\mathcal{L} = \text{Span} \{J_1, \ldots, J_n\}$ with  $\mathbf{v}^T J_k = \mathbf{e}_k^T$  for some  $\mathbf{v} \in \mathbb{C}^n$ . Given  $\mathbf{z} \in \mathbb{C}^n$ , the matrix  $\sum_k z_k J_k \in \mathcal{L}$ is denoted by  $\mathcal{L}(\mathbf{z})$ . Since  $\mathbf{v}^T \mathcal{L}(\mathbf{z}) = \mathbf{z}^T$ ,  $\mathcal{L}(\mathbf{z})$  is called the matrix of  $\mathcal{L}$  whose  $\mathbf{v}$ -row is  $\mathbf{z}^T$ .

*Example.*  $\mathcal{L} = sd M = \{Md(\mathbf{z})M^{-1} : \mathbf{z} \in \mathbb{C}^n\}$  is in  $\mathbb{V}$  since

$$\mathcal{L} = \operatorname{Span} \{J_1, \dots, J_n\}, \ J_k = M d(M^T \mathbf{e}_k) d(M^T \mathbf{v})^{-1} M^{-1},$$

for any vector **v** such that  $[M^T \mathbf{v}]_i \neq 0 \ \forall i$ , and

$$\mathbf{v}^T J_k = \mathbf{v}^T M d(M^T \mathbf{e}_k) d(M^T \mathbf{v})^{-1} M^{-1} = \mathbf{e}_k^T M d(M^T \mathbf{v}) d(M^T \mathbf{v})^{-1} M^{-1} = \mathbf{e}_k^T.$$

Note that  $\mathcal{L}(\mathbf{z}) = M d(M^T \mathbf{z}) d(M^T \mathbf{v})^{-1} M^{-1}$ .

A matrix  $X \in \mathbb{C}^{n \times n}$  is said to be non derogatory if the condition p(X) = 0, p polynomial, implies  $\partial p \ge n$ . Note that by the Cailey-Hamilton theorem the characteristic polynomial of X is null in X. So, X is non derogatory if and only if the set  $\{p(X)\}$  of all polynomials in X has dimension n. In [] it is stated the following result, which proves that  $\mathbb{V}$  is a wide class of spaces of matrices.

Theorem ND. Let X be a  $n \times n$  matrix with complex entries. Then X is non derogatory if and only if  $\{p(X)\} := \{p(X) : p \text{ polynomials}\}$  is in  $\mathbb{V}$ .

The Proposition here below collects several properties of the spaces in  $\mathbb{V}$ . They will be used (in the next section) in order to prove important properties of the best least squares fit in  $\mathcal{L}$  of a matrix A, holding for all spaces  $\mathcal{L}$  of a particular subclass of  $\mathbb{V}$ .

Proposition  $\mathbb{V}$  (properties of spaces in  $\mathbb{V}$ ). Let  $\mathcal{L}$  be a space in  $\mathbb{V}$ , i.e.  $\mathcal{L} =$ Span  $\{J_1, \ldots, J_n\}$  with  $\mathbf{v}^T J_k = \mathbf{e}_k^T$  for some  $\mathbf{v} \in \mathbb{C}^n$ .

(1) If  $X \in \mathcal{L}$  and  $\mathbf{v}^T X = \mathbf{0}^T$ , then X = 0, thus  $\mathbf{v}^T X = \mathbf{v}^T Y$ ,  $X, Y \in \mathcal{L}$ , implies X = Y

proof: 
$$\mathbf{0}^T = \mathbf{v}^T X = \mathbf{v}^T \sum_k \alpha_k J_k = \sum_k \alpha_k \mathbf{e}_k^T = [\alpha_1 \cdots \alpha_n] \Rightarrow \alpha_k = 0 \ \forall k.$$
  
(2) If  $J_i X \in \mathcal{L}, X \in \mathbb{C}^{n \times n}$ , then  $J_i X = \sum_k [X]_{ik} J_k$ 

proof: there exist  $\alpha_k$  such that  $J_i X = \sum_k \alpha_k J_k$ ; multiplying the latter identity by  $\mathbf{v}^T$  we have

$$\mathbf{e}_i^T X = \mathbf{v}^T J_i X = \sum_k \alpha_k \mathbf{e}_k^T = [\alpha_1 \cdots \alpha_n]$$

which implies  $\alpha_k = [X]_{ik}$ .

(3) Let  $P_k \in \mathbb{C}^n$  be defined by  $\mathbf{e}_s^T P_k = \mathbf{e}_k^T J_s$  (note that  $\mathbf{e}_k^T = \mathbf{v}^T J_k = \sum_i v_i \mathbf{e}_i^T J_k = \sum_i v_i \mathbf{e}_k^T P_i = \mathbf{e}_k^T \sum_i v_i P_i$ , and thus  $\sum v_k P_k = I$ ). Then the following assertions are equivalent:

- (i)  $\mathcal{L}$  is closed under matrix multiplication
- (ii)  $J_i J_j = \sum_k [J_j]_{ik} J_k \ \forall i, j$

(iii) 
$$P_r J_j = J_j P_r \ \forall r, j$$

(iv) 
$$P_k P_r = \sum_i [P_r]_{ki} P$$

proof: The implication (i)  $\Rightarrow$  (ii) follows from (2) for  $X = J_j$ . The opposite implication is obvious. The fact that conditions (ii) and (iii) are equivalent follows by taking the (r, s) entry of the equality in (ii):

$$\mathbf{e}_{i}^{T} P_{r} J_{j} \mathbf{e}_{s} = \mathbf{e}_{r}^{T} J_{i} J_{j} \mathbf{e}_{s} = \sum_{k} [J_{j}]_{ik} [J_{k}]_{rs} = \sum_{k} [J_{j}]_{ik} [P_{r}]_{ks} = [J_{j} P_{r}]_{is}.$$

The fact that conditions (iii) and (iv) are equivalent follows from the identities:

$$[P_k P_r]_{ms} = [J_m P_r]_{ks} = [P_r J_m]_{ks} = [J_k J_m]_{rs} = \sum_i [J_k]_{ri} [J_m]_{is} = \sum_i [P_r]_{ki} [P_i]_{ms}$$

(3.5) If  $I \in \mathcal{L}$ , then  $\sum_i v_i J_i = I$  and  $\mathbf{v}^T P_k = \mathbf{e}_k^T$ , i.e. also the space Span  $\{P_1, \ldots, P_n\}$  is in  $\mathbb{V}$  (with the same  $\mathbf{v}$ )

proof: both I and  $\sum_i v_i J_i$  have  $\mathbf{v}^T$  as **v**-row, and both, by assumption, are in  $\mathcal{L}$ , so they must be equal; moreover, we have

$$\mathbf{v}^T P_k = \sum_i v_i \mathbf{e}_i^T P_k = \sum_i v_i \mathbf{e}_k^T J_i = \mathbf{e}_k^T \sum_i v_i J_i = \mathbf{e}_k^T.$$

(4) If  $\mathcal{L}$  is closed under matrix multiplication, then

$$\mathcal{L}(\mathcal{L}(\mathbf{z})^T\mathbf{z}') = \mathcal{L}(\mathbf{z}')\mathcal{L}(\mathbf{z}), \; \forall \, \mathbf{z}, \mathbf{z}' \in \mathbb{C}^n$$

proof: since  $\mathcal{L}$  is closed, the matrix  $\mathcal{L}(\mathbf{z}')\mathcal{L}(\mathbf{z})$  is in  $\mathcal{L}$ ; moreover, its **v**-row is  $\mathbf{z'}^T \mathcal{L}(\mathbf{z})$ ; the thesis follows from the fact that also  $\mathcal{L}(\mathcal{L}(\mathbf{z})^T \mathbf{z'})$  is the matrix of  $\mathcal{L}$  whose **v**-row is  $\mathbf{z'}^T \mathcal{L}(\mathbf{z})$ .

(5) Assume  $I \in \mathcal{L}$  and  $\mathcal{L}$  closed under matrix multiplication. Then  $X \in \mathcal{L}$  is non singular if and only if  $\exists \mathbf{z} \in \mathbb{C}^n$  such that  $\mathbf{z}^T X = \mathbf{v}^T$ ; in this case  $X^{-1} = \mathcal{L}(\mathbf{z})$ 

proof: by inspecting the kth row of the matrix  $\mathcal{L}(\mathbf{z})X$ , and applying properties (3) and (3.5), we obtain the identities

$$\mathbf{e}_k^T \mathcal{L}(\mathbf{z}) X = \mathbf{e}_k^T \sum_s z_s J_s X = \sum_s z_s \mathbf{e}_s^T P_k X = \mathbf{z}^T P_k X = \mathbf{z}^T X P_k = \mathbf{v}^T P_k = \mathbf{e}_k^T,$$

or, equivalently, the equality  $\mathcal{L}(\mathbf{z})X = I$ , which implies that X is non singular and  $X^{-1} = \mathcal{L}(\mathbf{z})$ .

The best least squares fit to A in  $\mathcal{L} \subset \mathbb{C}^{n \times n}$ 

Given a subspace  $\mathcal{L} \subset \mathbb{C}^{n \times n}$  and a  $n \times n$  matrix A, it is well defined  $\mathcal{L}_A$ , the projection on  $\mathcal{L}$  of A. In the following theorem we state some assumptions on  $\mathcal{L}$  (in particular we consider *n*-dimensional subspaces of  $\mathbb{C}^{n \times n}$ ) which assure that  $\mathcal{L}_A$  is hermitian whenever A is, and that the eigenvalues of  $\mathcal{L}_A$  are bounded by those of A. Such assumptions imply that  $\mathcal{L} \in \mathbb{V}$ .

Theorem  $\mathcal{L}_A$ . Assumptions:

 $\mathcal{L} \subset \mathbb{C}^{n \times n}, I \in \mathcal{L}, \mathcal{L} = \text{Span} \{J_1, \dots, J_n\} \text{ with } J_k \text{ such that}$ 

$$J_i^H J_j = \sum_{k=1}^n \overline{[J_k]_{ij}} J_k, \ i, j = 1, \dots, n.$$
(\*)

 $A \in \mathbb{C}^{n \times n}$ .

 $\mathcal{L}_A \in \mathcal{L}, \|A - \mathcal{L}_A\|_F \leq \|A - X\|_F, \forall X \in \mathcal{L} \text{ (such matrix } \mathcal{L}_A \text{ is well defined since } \mathbb{C}^{n \times n} \text{ is a Hilbert space with respect the norm } \|\cdot\|_F \text{ induced by the inner product } (A, B)_F = \sum_{ij} \overline{a_{ij}} b_{ij} \text{ and } \mathcal{L} \text{ is a subspace of } \mathbb{C}^{n \times n}$ ).

Thesis: If  $A = A^H$ , then  $\mathcal{L}_A = \mathcal{L}_A^H$  and  $\min \lambda(A) \le \lambda(\mathcal{L}_A) \le \max \lambda(A)$ .

Note: if A is real symmetric, then  $\mathcal{L}_A$  is in general hermitian; it is real symmetric under the further condition that  $\mathcal{L}$  is spanned by real matrices (prove it!).

Note: we shall see that the hypotheses of Theorem  $\mathcal{L}_A$  are satisfied by spaces of the type  $\{Md(\mathbf{z})M^{-1}: \mathbf{z} \in \mathbb{C}^n\}$  if  $M^H M$  is diagonal and its diagonal entries are positive; however, the same hypotheses can be satisfied also by non commutative spaces (we shall see an example, for others see []), so also in the latter cases we can say that the conclusions of Theorem  $\mathcal{L}_A$  hold.

Applications of Theorem  $\mathcal{L}_A$ . If the conditions of Theorem  $\mathcal{L}_A$  are satisfied, then  $\mathcal{L}_A$  is positive definite (i.e.  $\mathcal{L}_A = \mathcal{L}_A^H$  and  $\mathbf{z} \in \mathbb{C}^n \ \mathbf{z} \neq \mathbf{0} \Rightarrow \mathbf{z}^H \mathcal{L}_A \mathbf{z}$  positive) whenever A is positive definite. So, in order to solve the linear system

$$A\mathbf{x} = \mathbf{b}, A$$
 positive definite

we can solve the equivalent system

$$\mathcal{L}_A^{-1}A\mathbf{x} = \mathcal{L}_A^{-1}\mathbf{b}$$

whose coefficient matrix has real and positive eigenvalues, often better distributed than those of A (this results, for example when solving  $A\mathbf{x} = \mathbf{b}$  iteratively, in less iterations).

Moreover, if the conditions of Theorem  $\mathcal{L}_A$  are satisfied and  $\mathcal{L}$  is spanned by real matrices, then then  $\mathcal{L}_A$  is real positive definite (i.e.  $\mathcal{L}_A = \mathcal{L}_A^T$ ,  $\mathcal{L}_A \in \mathbb{R}^{n \times n}$ , and  $\mathbf{z} \in \mathbb{C}^n \ \mathbf{z} \neq \mathbf{0} \Rightarrow \mathbf{z}^H \mathcal{L}_A \mathbf{z}$  positive) whenever A is real positive definite. That is, the following implications hold

 $B_k \text{ real positive definite } \Rightarrow \\ \mathcal{L}_{B_k} \text{ real positive definite } \Rightarrow \\ \varphi(\mathcal{L}_{B_k}, \mathbf{s}_k, \mathbf{y}_k) \text{ real positive definite } \Rightarrow \\ \mathcal{L}_{\varphi(\mathcal{L}_{B_k}, \mathbf{s}_k, \mathbf{y}_k)} \text{ real positive definite } \end{cases}$ 

(provided  $\mathbf{s}_k^T \mathbf{y}_k$  is positive), and thus both the S and NS LQN search directions,

$$\mathbf{d}_{k+1} = -\varphi(\mathcal{L}_{B_k}, \mathbf{s}_k, \mathbf{y}_k)^{-1} \nabla f(\mathbf{x}_{k+1}) \text{ and } \mathbf{d}_{k+1} = -\mathcal{L}_{\varphi(\mathcal{L}_{B_k}, \mathbf{s}_k, \mathbf{y}_k)}^{-1} \nabla f(\mathbf{x}_{k+1}),$$

are well defined descent directions in  $\mathbf{x}_{k+1}$  for the function  $f : \mathbb{R}^n \to \mathbb{R}$  (see [] for the definitions of  $B_k$ ,  $\mathbf{s}_k$ ,  $\mathbf{y}_k$ ,  $\varphi$ , S and NS LQN).

proof (of Theorem  $\mathcal{L}_A$ ): The matrix  $\mathcal{L}_A = \sum_s \alpha_s J_s$  is uniquely defined by the following condition

$$(X, A - \mathcal{L}_A)_F = 0, \ X \in \mathcal{L}$$

or, equivalently, by the *n* conditions  $(J_k, A - \sum_s \alpha_s J_s)_F = 0, k = 1, ..., n$ , which can be rewritten as follows

$$\sum_{s=1}^{n} (J_k, J_s)_F \alpha_s = (J_k, A)_F, \ k = 1, \dots, n.$$

In other words we have the formula

$$\mathcal{L}_A = \sum_{s} [B^{-1}\mathbf{c}]_s J_s, \ B_{ks} = (J_k, J_s)_F, \ c_k = (J_k, A)_F, \ k, s = 1, \dots, n.$$

*Remark.* B is positive definite, i.e.  $B = B^*$  and  $\mathbf{z}^H B \mathbf{z} > 0 \ \forall \mathbf{z} \in \mathbb{C}^n \ \mathbf{z} \neq \mathbf{0}$ .

proof:  $\overline{B_{ks}} = \overline{(J_k, J_s)_F} = (J_s, J_k)_F = B_{sk}$ , that is, B is a hermitian matrix. Moreover, since  $0 < (\sum_s z_s J_s, \sum_s z_s J_s)_F = \sum_{k,s} \overline{z_k} z_s (J_k, J_s)_F = \mathbf{z}^H B \mathbf{z}$  whenever  $\mathbf{z} \neq \mathbf{0}$ , the matrix B is also positive definite.

*Remark.* Let  $v_k \in \mathbb{C}$  be such that  $I = \sum_k v_k J_k$  (such  $v_k$  exist because  $I \in \mathcal{L}$ ). Then the vector  $\mathbf{v}$  whose entries are the  $v_k$  satisfies the equalities  $\mathbf{v}^T J_k = \mathbf{e}_k^T$ , thus  $\mathcal{L} \in \mathbb{V}$  and all results stated for spaces in  $\mathbb{V}$  hold for our space  $\mathcal{L}$ . proof: Multiply (\*) by  $\overline{v_i}$  and sum on *i*:

$$\overline{v_i}J_i^H J_j = \sum_k \overline{v_i}\overline{[J_k]_{ij}}J_k, \ J_j = \sum_k (\sum_i \overline{v_i}\overline{[J_k]_{ij}})J_k = \sum_k (\mathbf{v}^H \overline{J_k}\mathbf{e}_j)J_k.$$

This implies  $\mathbf{v}^H \overline{J_k} \mathbf{e}_j = 0$  if  $k \neq j$  and  $\mathbf{v}^H \overline{J_k} \mathbf{e}_j = 1$  if k = j, i.e.  $\mathbf{v}^T J_k = \mathbf{e}_k^T$ .

As an immediate consequence of the above two Remarks, we have that  $\mathcal{L}_A$  is the matrix of  $\mathcal{L}$  whose **v**-row is  $(B^{-1}\mathbf{c})^T$ ,

$$\mathcal{L}_A = \sum_s [B^{-1}\mathbf{c}]_s J_s = \mathcal{L}(B^{-1}\mathbf{c}),$$

and, moreover,

$$\mathcal{L}_A = \mathcal{L}(B^{-1}\mathbf{c}) = \mathcal{L}((B^H)^{-1}\mathbf{c}) = \mathcal{L}((\overline{B}^{-1})^T\mathbf{c}).$$

*Remark.*  $\mathcal{L}$  is closed under conjugate transposition. proof: multiply (\*) by  $v_i$  and sum on j:

$$J_i^H v_j J_j = \sum_k v_j \overline{[J_k]_{ij}} J_k, \ J_i^H = \sum_k (\sum_j v_j \overline{[J_k]_{ij}}) J_k \ \Rightarrow \ J_i^H \in \mathcal{L}.$$

The latter Remark yields part of the thesis of Theorem  $\mathcal{L}_A$ , because it implies that  $\mathcal{L}_A^H \in \mathcal{L}$ , and this fact together with the equalities

$$||A - \mathcal{L}_A||_F = ||A^H - \mathcal{L}_A^H||_F = ||A - \mathcal{L}_A^H||_F$$

(remember that our A is hermitian!) and the unicity of the best approximation of A, yield the identity  $\mathcal{L}_A = \mathcal{L}_A^H$ . In other words, under our conditions on  $\mathcal{L}$ the projection on  $\mathcal{L}$  of a hermitian matrix is hermitian too.

*Remark.*  $\mathcal{L}$  is closed under matrix multiplication ( $\mathcal{L}$  is a matrix algebra).

proof: the set  $\{J_i^H\}$  forms an alternative basis for  $\mathcal{L}$  (prove it!), thus there exist  $z_i^{(s)} \in \mathbb{C}$  such that  $J_s = \sum_i z_i^{(s)} J_i^H$ . Multiply (\*) by  $z_i^{(s)}$  and sum on i,

$$z_i^{(s)} J_i^H J_j = \sum_k z_i^{(s)} \overline{[J_k]_{ij}} J_k, \ J_s J_j = \sum_k (\sum_i z_i^{(s)} \overline{[J_k]_{ij}}) J_k,$$

to observe that  $J_s J_j \in \mathcal{L}$ .

*Remark.*  $\overline{B} = \sum_k \overline{\operatorname{tr}(J_k)} J_k$ , thus  $\overline{B} \in \mathcal{L}$ , and, since  $\overline{B}$  is non singular (it is positive definite!), by the result  $\mathbb{V}$  (5) also the matrix  $\overline{B}^{-1}$  is in  $\mathcal{L}$ . proof: by equality (\*) we have:

$$B_{ij} = (J_i, J_j)_F = \sum_{r,t} \overline{[J_i]_{rt}} [J_j]_{rt} = \sum_{r,t} [J_i^H]_{tr} [J_j]_{rt}$$
  
=  $\sum_t [J_i^H J_j]_{tt} = \sum_t \sum_k \overline{[J_k]_{ij}} [J_k]_{tt} = \sum_k \operatorname{tr} (J_k) \overline{[J_k]_{ij}}.$ 

The latter two Remarks, together with  $\mathbb{V}$  (4), let us rewrite again  $\mathcal{L}_A$  as follows

$$\mathcal{L}_A = \ldots = \mathcal{L}((\overline{B}^{-1})^T \mathbf{c}) = \mathcal{L}(\mathbf{c})\overline{B}^{-1}.$$

Now note that there exists a hermitian matrix M such that  $M^2 = \overline{B}^{-1}$ , and that the matrices  $\mathcal{L}_A$  and  $M\mathcal{L}(\mathbf{c})M$  have the same eigenvalues (by the last

representation of  $\mathcal{L}_A$  they are similar!). So, if  $\lambda(\mathcal{L}_A)$  is the generic eigenvalue of  $\mathcal{L}_A$ , then there exists  $\mathbf{x} \in \mathbb{C}^n \|\mathbf{x}\|_2 = 1$  such that

$$\lambda(\mathcal{L}_A) = \mathbf{x}^H M \mathcal{L}(\mathbf{c}) M \mathbf{x} = (M \mathbf{x})^H \mathcal{L}(\mathbf{c}) (M \mathbf{x}).$$

*Remark.* If  $\mathbf{z} \in \mathbb{C}^n$ , then  $\mathbf{z}^H \mathcal{L}(\mathbf{c})\mathbf{z} = \sum_k (P_k^H \mathbf{z})^H A(P_k^H \mathbf{z})$ . proof: here again equality (\*) is fundamental:

$$\begin{aligned} \mathbf{z}^{H} \mathcal{L}(\mathbf{c}) \mathbf{z} &= \mathbf{z}^{H} (\sum_{k} (J_{k}, A)_{F} J_{k}) \mathbf{z} \\ &= \sum_{i,j=1}^{n} \overline{z_{i}} z_{j} \sum_{r,t=1}^{n} a_{rt} \sum_{k} [J_{k}]_{ij} \overline{[J_{k}]_{rt}} \\ &= \sum_{i,j=1}^{n} \overline{z_{i}} z_{j} \sum_{r,t=1}^{n} a_{rt} \overline{[J_{i}^{H}J_{j}]_{rt}} \\ &= \sum_{r,t=1}^{n} a_{rt} \sum_{i,j=1}^{n} \overline{z_{i}} z_{j} \sum_{k} \overline{[J_{i}^{H}]_{rk}} \overline{[J_{j}]_{kt}} \\ &= \sum_{k} \sum_{r,t} a_{rt} (\sum_{i} z_{i} [P_{k}^{H}]_{ri}) (\sum_{j} z_{j} [P_{k}^{H}]_{tj}) \\ &= \sum_{k} \sum_{r,t} a_{rt} \overline{(P_{k}^{H} \mathbf{z})_{r}} (P_{k}^{H} \mathbf{z})_{t}. \end{aligned}$$

By the above Remark we have:

$$\begin{array}{lll} \lambda(\mathcal{L}_A) &=& \sum_k (P_k^H M \mathbf{x})^H A(P_k^H M \mathbf{x}) \leq \max \lambda(A) \sum_k (P_k^H M \mathbf{x})^H (P_k^H M \mathbf{x}) \\ &=& \max \lambda(A) \mathbf{x}^H M(\sum_k P_k P_k^H) M \mathbf{x}. \end{array}$$

But the matrix in the brackets is nothing else the matrix  $\overline{B}$ :

Remark.  $\overline{B} = \sum_k P_k P_k^H$  proof:

$$B_{ij} = (J_i, J_j)_F = \sum_{r,s} \overline{[J_i]_{rs}} [J_j]_{rs} = \sum_{r,s} \overline{[P_r]_{is}} [P_r]_{js} = \sum_{r,s} \overline{[P_r]_{is}} [P_r^T]_{sj} = \sum_r [\overline{P_r} P_r^T]_{ij}.$$

We can now conclude one of the inequalities (stated in Theorem  $\mathcal{L}_A$ ) satisfied by the eigenvalues of A and  $\mathcal{L}_A$ :

$$\lambda(\mathcal{L}_A) \leq \max \lambda(A) \mathbf{x}^H M M^{-2} M \mathbf{x} = \max \lambda(A) \mathbf{x}^H \mathbf{x} = \max \lambda(A).$$

Analogously, one can prove that  $\lambda(\mathcal{L}_A) \geq \min \lambda(A)$ .  $\Box$ 

*Exercise* DG. Prove that the *dihedral group* space

$$\mathcal{L} = \left\{ \begin{bmatrix} X & JY \\ JY & X \end{bmatrix} : X, Y \; \frac{n}{2} \times \frac{n}{2} \text{ circulants} \right\}$$

satisfies the hypothesis of Theorem  $\mathcal{L}_A$ , i.e.  $I \in \mathcal{L}$ ,  $\mathcal{L} = \text{Span} \{J_1, \ldots, J_n\}$  with  $J_k$  linearly independent such that

$$J_i^H J_j = \sum_{k=1}^n \overline{[J_k]_{ij}} J_k, \ i, j = 1, \dots, n.$$

Thus, the projection  $\mathcal{L}_A$  on  $\mathcal{L}$  is hermitian and such that  $\min \lambda(A) \leq \lambda(\mathcal{L}_A) \leq \max \lambda(A)$  whenever A is hermitian. Note that  $\mathcal{L}$  is not commutative.

Proposition  $\mathbb{CV}(\text{properties of commutative spaces in } \mathbb{V})$  [mitia]. Let  $\mathcal{L}$  be a space in  $\mathbb{V}$ , i.e.  $\mathcal{L} = \text{Span}\{J_1, \ldots, J_n\}$  with  $\mathbf{v}^T J_k = \mathbf{e}_k^T$  for some  $\mathbf{v} \in \mathbb{C}^n$ . Assume that  $\mathcal{L}$  is commutative. Then

(1)  $\mathbf{e}_i^T J_j = \mathbf{e}_i^T J_i, \forall i, j, \text{ and thus } J_k = P_k$ 

proof:  $J_i J_j = J_j J_i \Rightarrow \mathbf{v}^T J_i J_j = \mathbf{v}^T J_j J_i$ , and the definition  $\mathbf{v}^T J_k = \mathbf{e}_k^T$  yields the thesis.

(2) 
$$\mathbf{z}^T \mathcal{L}(\mathbf{z}') = \mathbf{z}'^T \mathcal{L}(\mathbf{z})$$

proof: by (1) we have

$$\mathbf{z}^{T}\mathcal{L}(\mathbf{z}') = \sum_{i} z_{i} \mathbf{e}_{i}^{T} \sum_{k} z_{k}' J_{k} = \sum_{i} z_{i} \sum_{k} z_{k}' \mathbf{e}_{k}^{T} J_{i} = \sum_{k} z_{k}' \mathbf{e}_{k}^{T} \sum_{i} z_{i} J_{i}.$$

(3)  $I = \mathcal{L}(\mathbf{v}) \in \mathcal{L}$ 

proof: note that  $\mathcal{L}(\mathbf{v}) = \sum v_i J_i \in \mathcal{L}$  and  $\mathbf{e}_k^T \mathcal{L}(\mathbf{v}) = \sum_i v_i \mathbf{e}_i^T J_k = \mathbf{v}^T J_k = \mathbf{e}_k^T$ , so  $I = \mathcal{L}(\mathbf{v}) \in \mathcal{L}$ .

(4)  $\mathcal{L}$  is closed under matrix multiplication

proof: from (1) we have that  $J_k = P_k$ , thus  $P_k J_s = J_k J_s = J_s J_k = J_s P_k$ , which is one of the necessary and sufficient conditions for the multiplicative closure.

Example of commutative  $\mathcal{L} \in \mathbb{V}$ . Let M be a non singular  $n \times n$  matrix with complex entries, and set  $\mathcal{L} = sdM = \{Md(\mathbf{z})M^{-1} : \mathbf{z} \in \mathbb{C}^n\}$ . Note that  $\mathcal{L} \in \mathbb{V}$ , in fact if  $\mathbf{v}$  is any vector such that  $[M^T\mathbf{v}]_k \neq 0, \forall k$ , then the matrices  $J_k = Md(M^T\mathbf{e}_k)d(M^T\mathbf{v})^{-1}M^{-1}$  satisfy the identities  $\mathbf{v}^T J_k = \mathbf{e}_k^T$  and span  $\mathcal{L}$ . We have, for  $\mathcal{L}$ , the following alternative representation:

$$\mathcal{L} = \{ Md(M^T \mathbf{z}) d(M^T \mathbf{v})^{-1} M^{-1} : \mathbf{z} \in \mathbb{C}^n \}.$$

It is clear that the matrix of  $\mathcal{L}$  whose **v**-row is  $\mathbf{z}^T$  is

$$\mathcal{L}(\mathbf{z}) = M d(M^T \mathbf{z}) d(M^T \mathbf{v})^{-1} M^{-1}.$$

Obviously,  $\mathcal{L}$  is commutative.

Proposition ch $\mathbb{V}$  (properties of commutative, closed under conjugate transposition spaces in  $\mathbb{V}$ ) [stefano]. Let  $\mathcal{L}$  be a space in  $\mathbb{V}$ , i.e.  $\mathcal{L} = \text{Span} \{J_1, \ldots, J_n\}$ with  $\mathbf{v}^T J_k = \mathbf{e}_k^T$  for some  $\mathbf{v} \in \mathbb{C}^n$ . Assume that  $\mathcal{L}$  is commutative and closed under conjugate transposition. Then, besides the above  $\mathbb{C}\mathbb{V}$  (1),(2),(3),(4), we have

$$J_i^H J_j = \sum_k \overline{[J_k]_{ij}} J_k, \ i, j = 1, \dots, n$$

proof: since  $J_i^H \in \mathcal{L}$  and  $\mathcal{L}$  is commutative, one has  $J_i^H J_j = J_j J_i^H$ ; since  $\mathcal{L}$  is closed under matrix multiplication and  $J_i^H \in \mathcal{L}$ , one has that  $J_j J_i^H \in \mathcal{L}$ ; by  $\mathbb{V}$  (2), it follows that

$$J_i^H J_j = J_j J_i^H = \sum_k [J_i^H]_{jk} J_k = \sum_k \overline{[J_i]_{kj}} J_k = \sum_k \overline{[J_k]_{ij}} J_k,$$

where in the latter identity we have used property cV(1).

Example of commutative, closed under conjugate transposition  $\mathcal{L} \in \mathbb{V}$ . Let M be a non singular  $n \times n$  matrix with complex entries, and set  $\mathcal{L} = sdM = \{Md(\mathbf{z})M^{-1} : \mathbf{z} \in \mathbb{C}^n\}$ . We already know that  $\mathcal{L}$  is a space in  $\mathbb{V}$  which is commutative. We want to prove that  $\mathcal{L}$  is closed under conjugate transposition if and only if  $M^H M$  is diagonal with positive diagonal entries. Moreover, in

such case there is a diagonal matrix  $d(\mathbf{w}) \ \mathbf{w} \in \mathbb{C}^n$  such that  $\tilde{M} = Md(\mathbf{w})$  is unitary and  $\mathcal{L} = \{\tilde{M}d(\mathbf{z})\tilde{M}^{-1}: \mathbf{z} \in \mathbb{C}^n\}$ .

One implication is easy: assume  $M^{\hat{H}}M = D$ ,  $D_{ii}$  positive  $\forall i, D_{ij} = 0 \ i \neq j$ ; then

$$(Md(\mathbf{z})M^{-1})^{H} = (M^{H})^{-1}d(\overline{\mathbf{z}})M^{H} = MD^{-1}d(\overline{\mathbf{z}})DM^{-1} = Md(\overline{\mathbf{z}})M^{-1} \in \mathcal{L}.$$

Now assume that  $\mathcal{L}$  is closed under conjugate transposition. Thus, for any  $\mathbf{z} \in \mathbb{C}^n$  there exists  $\mathbf{w} \in \mathbb{C}^n$  such that  $(Md(\mathbf{z})M^{-1})^H = Md(\mathbf{w})M^{-1}$ . But this implies

$$d(\overline{\mathbf{z}})C = Cd(\mathbf{w}), \ C = M^H M,$$

or, equivalently,  $c_{ij}\overline{z}_i = c_{ij}w_j$ . Assume the  $z_i$  distinct. Then the identities  $c_{i1}\overline{z}_i = c_{i1}w_1$ ,  $i = 1, \ldots, n$ , imply  $c_{i1} = 0$  for all i except one of them, say  $i_1$  (otherwise,  $c_{i1} \neq 0$  and  $c_{k1} \neq 0$ ,  $i \neq k$ , would imply  $w_1 = \overline{z}_i = \overline{z}_k$ !). Analogously, the identities  $c_{i2}\overline{z}_i = c_{i2}w_2$ ,  $i = 1, \ldots, n$ , imply  $c_{i2} = 0$  for all i except one of them, say  $i_2$ , and such  $i_2$  must be different from  $i_1$  otherwise C would be singular. Proceeding in this way one concludes that

C = DR, D diagonal, R permutation.

The fact that C is hermitian implies that the permutation matrix R must be symmetric. The fact that the diagonal entries of C cannot be zero implies that R is the identity. Finally, the fact that C is positive definite implies that  $D_{ii}$ are real and positive. Let us prove the last assertion. For  $\tilde{M} = Md(\mathbf{w})$  we have

$$\tilde{M}^{H}\tilde{M} = d(\overline{\mathbf{w}})M^{H}Md(\mathbf{w}) = d(\overline{\mathbf{w}})Dd(\mathbf{w}).$$

Choose  $|w_i| = 1/\sqrt{D_{ii}}$ ; then  $\tilde{M}^H \tilde{M} = I$  and, for any  $\mathbf{z} \in \mathbb{C}^n$ ,  $Md(\mathbf{z})M^{-1} = Md(\mathbf{w})d(\mathbf{z})d(\mathbf{w})^{-1}M^{-1} = \tilde{M}d(\mathbf{z})\tilde{M}^{-1} = \tilde{M}d(\mathbf{z})\tilde{M}^H$ .

Question. Find an example of  $\mathcal{L}$  satisfying the assumptions of Proposition ch $\mathbb{V}$  which is not of the type  $\{Md(\mathbf{z})M^{-1}: \mathbf{z} \in \mathbb{C}^n\}, M^H M$  diagonal with positive diagonal entries.

*Remark.* If  $\mathcal{L} = \{Ud(\mathbf{z})U^* : \mathbf{z} \in \mathbb{C}^n\}$  where U is a unitary matrix, then the thesis of Theorem  $\mathcal{L}_A$  can be proved very simply as follows. Since  $\forall \mathbf{z} \in \mathbb{C}^n$ 

$$||A - Ud(\mathbf{z})U^*||_F = ||U^*AU - d(\mathbf{z})||_F$$

it is clear that  $||A - Ud(\mathbf{z})U^*||_F$  is minimum for  $z_i = (U^*AU)_{ii}$ . So, the following formula for  $\mathcal{L}_A$  holds

$$\mathcal{L}_A = U \operatorname{diag}\left((U^* A U)_{ii}\right) U^*$$

from which it immediately follows the assertion:  $A = A^* \Rightarrow \mathcal{L}_A$  hermitian and min  $\lambda(A) \leq \lambda(\mathcal{L}_A) \leq \max \lambda(A)$ . As an application compute  $\mathcal{L}_{\mathbf{x}\mathbf{y}^T}$ ,  $\mathbf{x}, \mathbf{y} \in \mathbb{C}^n$ .

## Other remarks/exercises on spaces in $\mathbb{V}$

*Exercise.* Prove that there are (besides I) infinite matrices in the algebra  $\gamma$  whose first row is  $\mathbf{e}_1^T$ .

*Exercise*. Consider the matrix in Exercise G. Prove that the sum of its first and last rows is a vector with all entries nonzero (i.e.  $\gamma$  is a space in  $\mathbb{V}$  with  $\mathbf{v} = \mathbf{e}_1 + \mathbf{e}_n$ )

proof: for  $j = 1, \ldots, n$  we have

$$[G]_{1j} + [G]_{nj} = \left(\cos\frac{(2j-1)\pi}{2n} + \sin\frac{(2j-1)\pi}{2n}\right) \\ + \left(\cos\frac{(2n-1)(2j-1)\pi}{2n} + \sin\frac{(2n-1)(2j-1)\pi}{2n}\right) \\ = \left(\cos\frac{(2j-1)\pi}{2n} + \sin\frac{(2j-1)\pi}{2n}\right) \\ + \left(\cos((2j-1)\pi - \frac{(2j-1)\pi}{2n}\right) + \sin((2j-1)\pi - \frac{(2j-1)\pi}{2n})) \\ = \cos\frac{(2j-1)\pi}{2n} + \sin\frac{(2j-1)\pi}{2n} \\ - \cos\frac{(2j-1)\pi}{2n} + \sin\frac{(2j-1)\pi}{2n} = 2\sin\frac{(2j-1)\pi}{2n} \neq 0.$$

*Exercise.* Prove that any *n*-dimensional space  $\mathcal{L} \subset \mathbb{C}^{n \times n}$  with the property  $A\mathbf{e}_j = \mathbf{0} \ \forall A \in \mathcal{L}$  cannot be in  $\mathbb{V}$ . (Suggestion: show that  $J_j \in \mathcal{L}$  such that  $\mathbf{v}^T J_j = \mathbf{e}_j^T$  does not exist).

*Exercise.* Let  $\mathcal{L}$  be in  $\mathbb{V}$  and closed under matrix multiplication. Prove that

(i) If the matrices in *L* are symmetric, then *L* is commutative
(ii) If the matrices in *L* are persymmetric, then *L* is commutative

proof: by the assumptions on  $\mathcal{L}$ , we have, in the symmetric case,

$$J_s J_k = J_s^T J_k^T = (J_k J_s)^T = J_k J_s,$$

and, in the persymmetric case,

$$J_{s}J_{k} = J_{s}JJJ_{k} = JJ_{s}^{T}J_{k}^{T}J = J((J_{k}J_{s})^{T})J = J(JJ_{k}J_{s}J)J = J_{k}J_{s}.$$

Question. And if we start from the more general definition  $\mathcal{L} \subset \mathbb{C}^{n \times n}$ ,  $\mathcal{L} =$ Span  $\{J_1, \ldots, J_n\}$  with  $\mathbf{v}^T J_k = \mathbf{u}_k$  for some  $\mathbf{v}, \mathbf{u}_k \in \mathbb{C}^n$  such that  $\mathbf{u}_k^H \mathbf{u}_s = 0$   $k \neq s$ ?

EXERCISE.

$$P_{\xi} = \begin{bmatrix} 0 & 1 & & \\ & 0 & 1 & & \\ & & \ddots & \\ & & & 1 \\ \xi & & & 0 \end{bmatrix}, \ \xi \neq 0$$

(1) If  $\rho^n = \xi$  and  $\omega^n = 1$ , then

$$P_{\xi} \begin{bmatrix} 1\\ \rho \omega^{j}\\ \rho^{n-1} \omega^{(n-1)j} \end{bmatrix} = \rho \omega^{j} \begin{bmatrix} 1\\ \rho \omega^{j}\\ \rho^{n-1} \omega^{(n-1)j} \end{bmatrix}, \quad j = 0, 1, \dots, n-1.$$

Thus, if  $W = (\omega^{ij})_{i,j=0}^{n-1}$  and  $D_{1\rho^{n-1}} = \text{diag}(\rho^i, i = 0, ..., n-1)$ , then

$$P_{\xi}(D_{1\rho^{n-1}}W) = (D_{1\rho^{n-1}}W)\rho D_{1\omega^{n-1}}, \ D_{1\omega^{n-1}} = \operatorname{diag}(\omega^j, j = 0, \dots, n-1).$$

(2) If, moreover,  $|\xi|=1$  and  $\omega^i \neq 1 \; 0 < i < n,$  then  $U=\frac{1}{\sqrt{n}}D_{1\rho^{n-1}}W$  is unitary and

$$C_{\xi} := H_{P_{\xi}} = \{\sum_{k=1}^{n} z_k P_{\xi}^{k-1} : z_k \in \mathbb{C}\} \\ = \{Ud(\mathbf{z})U^* : \mathbf{z} \in \mathbb{C}^n\} = \{Ud(U^T\mathbf{z})d(U^T\mathbf{e}_1)^{-1}U^{-1} : \mathbf{z} \in \mathbb{C}^n\}.$$

The matrix  $C_{\xi}(\mathbf{a}) := \sum_{k=1}^{n} a_k P_{\xi}^{k-1} = U d(U^T \mathbf{a}) d(U^T \mathbf{e}_1)^{-1} U^{-1}$  is the  $\xi$ -circulant matrix whose first row is  $\mathbf{a}^T$ :

$$C_{\xi}(\mathbf{a}) = \begin{bmatrix} a_1 & a_2 & a_{n-1} & a_n \\ \xi a_n & a_1 & & a_{n-1} \\ & & & \\ \xi a_3 & & & a_2 \\ \xi a_2 & \xi a_3 & \xi a_n & a_1 \end{bmatrix}$$

#### EXERCISE.

Set  $X = P_1 + P_1^T$ .

1) Find a convenient representation for the set  $C^S$  of all polynomials in X, and deduce the dimension of  $C^S$ .

2) Prove that any matrix of the form  $C_1 + JC_2$ ,  $C_1, C_2$  circulants, J antiidentity, belongs to the space  $\{A \in \mathbb{C}^{n \times n} : AX = XA\}$ . Deduce a lower bound for the dimension of  $\{A \in \mathbb{C}^{n \times n} : AX = XA\}$ .

Repeat the exercise for  $X = P_{-1} + P_{-1}^T$ .

### EXERCISE

1) Write precisely an algorithm that computes the matrix-vector product  $T \cdot \mathbf{z}, T = (t_{i-j})_{i,j=1}^n$ , in  $O(n \log_2 n)$  arithmetic operations.

2) Write an algorithm that computes the matrix-vector product  $(T^{-1}) \cdot \mathbf{z}$  in  $O(n \log_2 n)$  arithmetic operations (after preprocessing on T).

#### EXERCISE

Do exercise DG

EXERCISE Prove Theorem DD

EXERCISE Let T be a  $n \times n$  Toeplitz matrix

1) Prove that  $TP_0 - P_0T$  has rank at most 2, and write the Gohberg-Semencul formula for  $T^{-1}$ 

2) Let H be a Hankel matrix, i.e.  $H = (h_{i+j-2})_{i,j=1}^n$ , and show that (T + i) $H)(P_0+P_0^T)-(P_0+P_0^T)(T+H)$  has rank at most 4

3) Write the matrix of the algebra  $\tau$  whose first row is  $\begin{bmatrix} 2 & -1 & 0 & 0 & \cdots & 0 \end{bmatrix}$ , and observe that it is a Toeplitz matrix. Call A such matrix and compute  $A^{-1}$ explicitly by using the fact that  $A^{-1} \in \tau$  (so, it is sufficient to compute its first row). Is  $A^{-1}$  a Toeplitz matrix ?

4) Do exercise Ttau

EXERCISE Under the assumptions of Theorem  $\mathcal{L}_A$ 

1) prove the last of the following identities

$$\mathcal{L}_A = \mathcal{L}(B^{-1}\mathbf{c}) = \mathcal{L}(\mathbf{c})\overline{B}^{-1} = \overline{B}^{-1}\mathcal{L}(\mathbf{c})$$

(hint: find an expression of  $\overline{B}$  in terms of the matrices  $P_k$ ,  $\mathbf{e}_s^T P_k = \mathbf{e}_k^T J_s \,\forall s, k$ ) 2) prove that B is positive definite (besides hermitian)

EXERCISE (0) Let  $\mathcal{L} \in \mathbb{C}^{n \times n}$ ,  $\mathcal{L} = \text{Span} \{J_1, \dots, J_n\}$  with  $\mathbf{v}^T J_k = \mathbf{e}_k^T$  for some vector  $\mathbf{v}$  (that is,  $\mathcal{L} \in \mathbb{V}$ ). Assume that  $\mathcal{L}$  is closed under matrix multiplication, that  $I \in \mathcal{L}$ , and that  $J_i^H = \alpha_i J_{t_i}, |\alpha_i| = 1 \ \forall i$ , for some  $t_i \in \{1, 2, \dots, n\}$ . (1) Prove that  $J_i^H J_j = \sum_k \overline{[J_k]_{ij}} J_k, \ \forall i, j$ .

(2) Assume that  $\{1, 2, ..., n\}$  is a group with identity element 1. Prove that the space

$$\mathcal{L} = \{ A \in \mathbb{C}^{n \times n} : a_{ij} = a_{si,sj}, \forall i, j, s \in \{1, \dots, n\} \}$$
  
= 
$$\{ A \in \mathbb{C}^{n \times n} : a_{ij} = a_{1,i^{-1}j}, \forall i, j \in \{1, \dots, n\} \}$$

satisfies the assumptions (0). (Any space  $\mathcal{L}$  of this type is usually called *group* matrix algebra; for example, circulants and the  $\mathcal{L}$  in Exercise DG are group matrix algebras).

(3) Prove that the space  $\mathcal{L}$  spanned by the matrices  $J_1 = I$ ,

$$J_2 = \begin{bmatrix} 1 & & \\ 1 & & \\ & -\mathbf{i} \end{bmatrix}, \ J_3 = \begin{bmatrix} 1 & & \\ & -\mathbf{i} \\ & & \end{bmatrix}, \ J_4 = \begin{bmatrix} & & 1 \\ & & \mathbf{i} \\ & -\mathbf{i} \end{bmatrix}, \ J_4 = \begin{bmatrix} & & 1 \\ & & \mathbf{i} \\ & -\mathbf{i} \end{bmatrix},$$

satisfies the assumptions (0). Prove that  $\mathcal{L}$  is not commutative.

EXERCISE Let T be a symmetric Toeplitz matrix.

(1) Compute the first row of  $C_T$  where C is the space of circulant matrices

(2) Compute the first row of  $\tau_T$  where  $\tau$  is the space of tau matrices

(3) Compute the first row of  $\mathcal{L}_T$  where  $\mathcal{L}$  is the space in the point (3) of the previous exercise (so, n = 4)

(4) Compute the vector  $(\mathbf{e}_1 + \mathbf{e}_6)^T \gamma_T$  where  $\gamma$  is the space of all  $6 \times 6$  gamma matrices (i.e. assume n = 6).

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# Solving EXERCISES:

 $C_{-1}^S + JC_{-1}^{SK}$  is a commutative matrix algebra. Assume  $A_i \in C_{-1}^S$ ,  $B_i \in C_{-1}^{SK}$ . Note that  $A_i$  and  $B_i$  are also persymmetric. Then

$$(A_0 + JB_0)(A_1 + JB_1) = A_0A_1 + A_0JB_1 + JB_0A_1 + JB_0JB_1 = A_0A_1 + JA_0B_1 + JB_0A_1 + B_0^TB_1.$$

Since  $A_0A_1$  is (-1)-circulant and  $(A_0A_1)^T = A_1^T A_0^T = A_1A_0 = A_0A_1$ ,  $A_0B_1$  is (-1)-circulant and  $(A_0B_1)^T = B_1^T A_0^T = -B_1A_0 = -A_0B_1$  ( $C_{-1}$  is closed under matrix multiplication and is commutative), we have  $A_0A_1 \in C_{-1}^S$  and  $A_0B_1 \in C_{-1}^{SK}$ .

Since  $B_0^T B_1$  is (-1)-circulant and  $(B_0^T B_1)^T = B_1^T B_0 = -B_1 B_0 = -B_1 (-B_0^T) = B_1 B_0^T = B_0^T B_1$ ,  $B_0 A_1$  is (-1)-circulant and  $(B_0 A_1)^T = A_1^T B_0^T = -A_1 B_0 = -A_1 B_0$ 

 $-B_0A_1$  ( $C_{-1}$  is closed under transposition and matrix multiplication, and commutative), we have  $B_0^TB_1 \in C_{-1}^S$  and  $B_0A_1 \in C_{-1}^{SK}$ .

Proof of Theorem DD.

$$\begin{split} & \left[ \sum H_{P_0}(\mathbf{x}_m)^T H_X(\mathbf{y}_m) \right] X - X \left[ \sum H_{P_0}(\mathbf{x}_m)^T H_X(\mathbf{y}_m) \right] \\ &= \sum \left[ H_{P_0}(\mathbf{x}_m)^T X - X H_{P_0}(\mathbf{x}_m)^T \right] H_X(\mathbf{y}_m) \\ &= b \sum \left[ H_{P_0}()^T P_0 - P_0 H_{P_0}()^T \right] H_X(\mathbf{y}_m) \\ &= b \sum \left[ \begin{bmatrix} 0 & & & \\ & (\mathbf{x}_m)_1 & & \\ & 0 & (\mathbf{x}_m)_{n-1} & (\mathbf{x}_m)_1 \end{bmatrix} - \begin{bmatrix} & (\mathbf{x}_m)_1 & & \\ & (\mathbf{x}_m)_{n-1} & (\mathbf{x}_m)_1 & \\ & 0 & & 0 \end{bmatrix} \right] H_X(\mathbf{y}_m) \\ &= b \sum \begin{bmatrix} -(\mathbf{x}_m)_1 & & \\ & -(\mathbf{x}_m)_{n-1} & (\mathbf{x}_m)_1 \end{bmatrix} H_X(\mathbf{y}_m) \\ &= b \sum (-\mathbf{x}_m \mathbf{e}_1^T + \mathbf{e}_n \mathbf{x}_m^T J) H_X(\mathbf{y}_m) \\ &= b \sum (-\mathbf{x}_m \mathbf{y}_m^T + \mathbf{e}_n \mathbf{x}_m^T H_X(\mathbf{y}_m)^T J) = \dots \end{split}$$

Compute the first row of  $C_T$ . Let  $J_k \in \mathbb{C}^{n \times n}$  be the circulant matrices with first row  $\mathbf{e}_k^T$ ,  $k = 1, \ldots, n$ . Let us compute  $(B^{-1}\mathbf{c})^T$  with respect to such basis (i.e. the first row of  $C_A$ ) when A is a symmetric Toeplitz matrix. So, we have

$$A = T = \begin{bmatrix} t_0 & t_1 & t_{n-1} \\ t_1 & t_0 & & \\ & & t_1 \\ t_{n-1} & t_1 & t_0 \end{bmatrix},$$

$$(J_1, A)_F = nt_0, (J_2, A)_F = (n-1)t_1 + t_{n-1}, (J_3, A)_F = (n-2)t_2 + 2t_{n-2}, (J_4, A)_F = (n-3)t_3 + 3t_{n-3}, \dots (J_n, A)_F = t_{n-1} + (n-1)t_1,$$

and thus

$$(J_k, A)_F = (n - k + 1)t_{k-1} + (k - 1)t_{n-k+1}, \ k = 1, \dots, n.$$

Since B = nI, we have  $B^{-1} = \frac{1}{n}I$ , and

$$[B^{-1}\mathbf{c}]_k = \frac{1}{n}((n-k+1)t_{k-1} + (k-1)t_{n-k+1}), \ k = 1, \dots, n.$$

Compute the first row of  $\tau_T$ . Let  $J_k \in \mathbb{C}^{n \times n}$  be the  $\tau$  matrices with first row  $\mathbf{e}_k^T$ ,  $k = 1, \ldots, n$ . Let us compute  $(B^{-1}\mathbf{c})^T$  with respect to such basis (i.e. the first row of  $\tau_A$ ) when A is a symmetric Toeplitz matrix. So, we have

$$(J_1, A)_F = nt_0, (J_2, A)_F = 2(n-1)t_1, (J_3, A)_F = (n-2)t_0 + 2(n-2)t_2, \\ (J_4, A)_F = 2(n-3)t_1 + 2(n-3)t_3, (J_5, A)_F = (n-4)t_0 + 2(n-4)t_2 + 2(n-4)t_4$$

One can guess that

$$(J_k, A)_F = (n - k + 1)[\delta_{k,o}t_0 + 2\sum_{j=1}^{[k/2]} t_{k-2j+1}], \ k = 1, \dots, n,$$

where  $\delta_{k,o} = 1$  if k is odd and  $\delta_{k,o} = 0$  if k is even. Since  $B^{-1} = \frac{1}{2n+2}(3J_1 - J_3)$  (prove such formula!), we have

$$\begin{array}{l} (B^{-1}\mathbf{c})_1 = \frac{1}{2n+2}(3(J_1,A)_F - (J_3,A)_F), \ (B^{-1}\mathbf{c})_2 = \frac{1}{2n+2}(2(J_2,A)_F - (J_4,A)_F), \\ (B^{-1}\mathbf{c})_k = \frac{1}{2n+2}(-(J_{k-2},A)_F + 2(J_k,A)_F - (J_{k+2},A)_F), \\ (B^{-1}\mathbf{c})_{n-1} = \frac{1}{2n+2}(-(J_{n-3},A)_F + 2(J_{n-1},A)_F), \ (B^{-1}\mathbf{c})_n = \frac{1}{2n+2}(-(J_{n-2},A)_F + 3(J_n,A)_F) \end{array}$$

It is not difficult to conclude that

$$(B^{-1}\mathbf{c})_1 = \frac{1}{n+1}((n+1)t_0 - (n-2)t_2), (B^{-1}\mathbf{c})_k = \frac{1}{n+1}((n-k+3)t_{k-1} - (n-k-1)t_{k+1}), \ k = 2, \dots, n-1, (B^{-1}\mathbf{c})_n = \frac{1}{n+1}(3t_{n-1}).$$

Compute the sum of the first row and of the last row of  $\gamma_T$ . We know that

$$\gamma_T = \sum_{k=1}^6 [B^{-1}\mathbf{c}]_k J_k = Gd(G^T B^{-1}\mathbf{c})d(G^T(\mathbf{e}_1 + \mathbf{e}_6))^{-1}G^{-1}, (\mathbf{e}_1 + \mathbf{e}_6)^T \gamma_T = (B^{-1}\mathbf{c})^T, \ B_{ij} = (J_i, J_j)_F, \ c_i = (J_i, T)_F,$$

where the  $J_k$  are the matrices in  $\gamma$  with the property  $(\mathbf{e}_1 + \mathbf{e}_6)^T J_k = \mathbf{e}_k^T$ ,  $k = 1, \ldots, 6$  (they have been written explicitly above). So, we have to compute the vector  $B^{-1}\mathbf{c}$ . Recall that we expect (from theory) that  $B, B^{-1}$  are matrices in  $\gamma$ :

$$B = 3 \begin{bmatrix} 1 & 2 & 1 & 0 & -1 & -2 \\ 2 & 1 & 2 & 1 & 0 & -1 \\ 1 & 2 & 1 & 2 & 1 & 0 \\ 0 & 1 & 2 & 1 & 2 & 1 \\ -1 & 0 & 1 & 2 & 1 & 2 \\ -2 & -1 & 0 & 1 & 2 & 1 \end{bmatrix}, B^{-1} = \frac{1}{12} \begin{bmatrix} -2 & 1 & 1 & 0 & -1 & -1 \\ 1 & -2 & 1 & 1 & 0 & -1 \\ 1 & 1 & -2 & 1 & 1 & 0 \\ 0 & 1 & 1 & -2 & 1 & 1 \\ -1 & 0 & 1 & 1 & -2 & 1 \\ -1 & -1 & 0 & 1 & 1 & -2 \end{bmatrix}$$
$$\mathbf{c} = \begin{bmatrix} 3t_0 \\ 3t_0 + 5t_1 - t_5 \\ 3t_0 + 5t_1 + 4t_2 - 2t_4 - t_5 \\ 3t_0 + 5t_1 - t_5 \\ 3t_0 + 5t_1 - t_5 \\ 3t_0 \end{bmatrix}.$$

Note that  $B, B^{-1}$  are symmetric (-1)-circulant matrices, which is a stronger condition than the expected  $B, B^{-1} \in \gamma$ . Moreover, note that the vector **c** is centrosymmetric; thus, since  $B^{-1}$  is centrosymmetric, i.e.  $JB^{-1} = B^{-1}J$  (it is both symmetric and persymmetric!), the vector  $B^{-1}\mathbf{c}$  is expected to be centrosymmetric. In fact, we have

$$\gamma_T(\mathbf{e}_1 + \mathbf{e}_6) = B^{-1}\mathbf{c} = \frac{1}{12} \begin{bmatrix} -6t_0 + 5t_1 + 4t_2 - 2t_4 - t_5 \\ 8t_2 - 4t_4 \\ 6t_0 + 5t_1 - 4t_2 + 2t_4 - t_5 \\ 6t_0 + 5t_1 - 4t_2 + 2t_4 - t_5 \\ 8t_2 - 4t_4 \\ -6t_0 + 5t_1 + 4t_2 - 2t_4 - t_5 \end{bmatrix}.$$

Note that a  $\gamma$  matrix A + JB with  $(\mathbf{e}_1 + \mathbf{e}_6)^T (A + JB)$  centrosymmetric must be in  $C_{-1}^S$ ; more precisely, the following implication holds:

$$(\mathbf{e}_1 + \mathbf{e}_6)^T (A + JB) = [z_1 \, z_2 \, z_3 \, z_3 \, z_2 \, z_1]$$
  

$$\Rightarrow \quad a_2 = z_3, \ a_1 = z_2 + z_3, \ a_0 = z_1 + z_2 + z_3, \ b_i = 0 \ \forall i.$$

In our case:

$$a_2 = \frac{1}{12}(6t_0 + 5t_1 - 4t_2 + 2t_4 - t_5), \ a_1 = \frac{1}{12}(6t_0 + 5t_1 + 4t_2 - 2t_4 - t_5), \ a_0 = \frac{1}{12}(10t_1 + 8t_2 - 4t_4 - 2t_5)$$

But  $\gamma_T$  should coincide with the T.Chan  $(C_1)_T$ ! ... is there something wrong ?

An alternative preconditioner for Toeplitz systems?

Consider the  $n - 2 \times n - 2 \tau$  matrix with first row  $[a_1 \ a_2 \ \cdots \ a_{n-2}]$  and call it  $\tau_{a_1 a_{n-2}}$ . For example

$$n = 5: \tau_{a_1 a_3} = \begin{bmatrix} a_1 & a_2 & a_3 \\ a_2 & a_1 + a_3 & a_2 \\ a_3 & a_2 & a_1 \end{bmatrix}, n = 6: \tau_{a_1 a_4} = \begin{bmatrix} a_1 & a_2 & a_3 & a_4 \\ a_2 & a_1 + a_3 & a_2 + a_4 & a_3 \\ a_3 & a_2 + a_4 & a_1 + a_3 & a_2 \\ a_4 & a_3 & a_2 & a_1 \end{bmatrix}, \dots$$

Consider the  $n \times n$  matrix

$$A_{a_1 a_{n-2}} = \begin{bmatrix} 0 & \mathbf{0}^T & 0 \\ \mathbf{0} & \tau_{a_1 a_{n-2}} & \mathbf{0} \\ 0 & \mathbf{0}^T & 0 \end{bmatrix}.$$

For example,

$$n = 5: A_{a_1 a_3} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & a_1 & a_2 & a_3 & 0 \\ 0 & a_2 & a_1 + a_3 & a_2 & 0 \\ 0 & a_3 & a_2 & a_1 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}, n = 6: A_{a_1 a_4} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & a_1 & a_2 & a_3 & a_4 & 0 \\ 0 & a_2 & a_1 + a_3 & a_2 + a_4 & a_3 & 0 \\ 0 & a_3 & a_2 + a_4 & a_1 + a_3 & a_2 & 0 \\ 0 & a_4 & a_3 & a_2 & a_1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}, \dots$$

We want to find  $\tau_{A_{a_1a_{n-2}}}$ , i.e. the  $n \times n \tau$  matrix defined by the following minimization property

$$\|\tau_{A_{a_1a_{n-2}}} - A_{a_1a_{n-2}}\|_F = \min\{\|X - A_{a_1a_{n-2}}\|_F : X \in \tau, X \ n \times n\}.$$

Let  $J_k$  be the  $n \times n \tau$  matrices defined by the conditions  $\mathbf{e}_1^T J_k = \mathbf{e}_k^T$ , k = 1, ..., n, and set  $\tau_{A_{a_1a_{n-2}}} = \tau(\mathbf{z}) := \sum_k z_k J_k$ . We want to find  $\mathbf{z}$ . We know that if  $B_{ij} = (J_i, J_j)_F$ ,  $c_i = (J_i, A_{a_1a_{n-2}})_F$ , then  $\mathbf{z} = B^{-1}\mathbf{c}$ . For example for n = 5:

$$B = \begin{bmatrix} 5 & 0 & 3 & 0 & 1 \\ 0 & 8 & 0 & 4 & 0 \\ 3 & 0 & 9 & 0 & 3 \\ 0 & 4 & 0 & 8 & 0 \\ 1 & 0 & 3 & 0 & 5 \end{bmatrix}, c = \begin{bmatrix} 3a_1 + a_3 \\ 4a_2 \\ 3a_1 + 3a_3 \\ 4a_2 \\ 3a_3 + a_1 \end{bmatrix}, B^{-1} = \frac{1}{12} \begin{bmatrix} 3 & 0 & -1 & 0 & 0 \\ 0 & 2 & 0 & -1 & 0 \\ -1 & 0 & 2 & 0 & -1 \\ 0 & -1 & 0 & 2 & 0 \\ 0 & 0 & -1 & 0 & 3 \end{bmatrix}.$$

Then

$$\mathbf{z} = B^{-1}\mathbf{c} = \frac{1}{6} \begin{bmatrix} 3a_1 \\ 2a_2 \\ a_1 + a_3 \\ 2a_2 \\ 3a_3 \end{bmatrix}$$

n = 6:

Since  $B_{ij} = (J_i, J_j)$  and  $B \in \tau$ , one can easily obtain the first row of B and write down B:

B =	$\begin{bmatrix} 6 \\ 0 \\ 4 \\ 0 \\ 2 \end{bmatrix}$	$\begin{array}{c} 0 \\ 10 \\ 0 \\ 6 \\ 0 \end{array}$	$egin{array}{c} 4 \\ 0 \\ 12 \\ 0 \\ 6 \end{array}$	$egin{array}{c} 0 \\ 6 \\ 0 \\ 12 \\ 0 \end{array}$	$2 \\ 0 \\ 6 \\ 0 \\ 10$	$\begin{array}{c} 0 \\ 2 \\ 0 \\ 4 \\ 0 \end{array}$	, c =	$ \begin{array}{c} 4a_2 + 2a_3 \\ 6a_2 + 2a_4 \\ 6a_3 + 4a_1 \\ 6a_2 + 4a_4 \\ 6a_3 + 2a_1 \end{array} $	$, B^{-1} = \frac{1}{2 \cdot 6 + 2}$	$     \begin{bmatrix}       3 \\       0 \\       -1 \\       0 \\       0     \end{bmatrix} $	$\begin{array}{c} 0 \\ 2 \\ 0 \\ -1 \\ 0 \end{array}$	$-1 \\ 0 \\ 2 \\ 0 \\ -1$	$\begin{array}{c} 0 \\ -1 \\ 0 \\ 2 \\ 0 \end{array}$	$\begin{array}{c} 0 \\ 0 \\ -1 \\ 0 \\ 2 \end{array}$	$\begin{array}{c} 0 \\ 0 \\ -1 \\ 0 \end{array}$	
	2	0	-	-						0	0	-		2	-	
	0	2	0	4	0	6_		$4a_4 + 2a_2$		0	0	0	-1	0	3	

Then

$$\mathbf{z} = B^{-1}\mathbf{c} = \frac{1}{7} \begin{bmatrix} 4a_1 \\ 3a_2 \\ a_1 + 2a_3 \\ 2a_2 + a_4 \\ 3a_3 \\ 4a_4 \end{bmatrix}$$

Exercise. Prove that

$$B^{-1} = \frac{1}{2n+2}(3J_1 - J_3)$$

(first find B and w such that  $\mathbf{w}^T B = \mathbf{e}_1^T$ , then  $B^{-1} = \tau(\mathbf{w})$ ).

By observing the **z** obtained for n = 5, n = 6 (see above), and in the cases n = 7 and n = 8,

$$\mathbf{z}^{T} = \frac{1}{8} [5a_{1} \ 4a_{2} \ a_{1} + 3a_{3} \ 2a_{2} + 2a_{4} \ 3a_{3} + a_{5} \ 4a_{4} \ 5a_{5}],$$
  
$$\mathbf{z}^{T} = \frac{1}{9} [6a_{1} \ 5a_{2} \ a_{1} + 4a_{3} \ 2a_{2} + 3a_{4} \ 3a_{3} + 2a_{5} \ 4a_{4} + a_{6} \ 5a_{5} \ 6a_{6}],$$

one can conjecture a formula for  $\mathbf{z}$  in case n is generic: Exercise. Prove that  $\tau_{A_{a_1a_{n-2}}} = \tau(\mathbf{z})$  with

$$\mathbf{z}^{T} = \frac{1}{n+1} \left[ \begin{bmatrix} 0 & 0 & a_{1} & 2a_{2} & \cdots & (n-4)a_{n-4} & (n-3)a_{n-3} & (n-2)a_{n-2} \end{bmatrix} + \begin{bmatrix} (n-2)a_{1} & (n-3)a_{2} & (n-4)a_{3} & \cdots & 2a_{n-3} & a_{n-2} & 0 & 0 \end{bmatrix} \right].$$
(\*\*)

APPLICATION. Let  $T=(t_{\mid i-j\mid})_{i,j=1}^n$  be a symmetric Toeplitz matrix. It is easy to realize that

$$T = \tau_{t_0 t_{n-1}} - A_{t_2 t_{n-1}}, \ A_{t_2 t_{n-1}} = \begin{bmatrix} 0 & \mathbf{0}^T & 0 \\ \mathbf{0} & \tau_{t_2 t_{n-1}} & \mathbf{0} \\ 0 & \mathbf{0}^T & 0 \end{bmatrix}.$$

From the equality

$$T = \tau_{t_0 t_{n-1}} - \tau_{A_{t_2 t_{n-1}}} + \tau_{A_{t_2 t_{n-1}}} - A_{t_2 t_{n-1}}$$

follows the inequality:

$$||T - \tau_T||_F \le ||\tau_{A_{t_2 t_{n-1}}} - A_{t_2 t_{n-1}}||_F = ||T - (\tau_{t_0 t_{n-1}} - \tau_{A_{t_2 t_{n-1}}})||_F.$$

However, the  $\tau$  matrix  $\tau_{t_0t_{n-1}} - \tau_{A_{t_2t_{n-1}}}$  could be for some reasons (...) a preconditioner for T better than  $\tau_T$ . Note that  $\tau_{A_{t_2t_{n-1}}} = \tau(\mathbf{z})$  where  $\mathbf{z}^T$  is like in (\*\*) but with  $a_i$  replaced by  $t_{i+1}$ ,  $i = 1, \ldots, n-2$ .

# Displacement decompositions.

Let  $\mathcal{L}$  be in  $\mathbb{V}$ , and X a matrix in  $\mathcal{L}$ . Assume that  $\mathcal{L}$  is commutative.

Denote by  $\mathcal{L}(\mathbf{z})$  the matrix of  $\mathcal{L}$  whose **v**-row is  $\mathbf{z}^T$ , and assume that  $\mathcal{L}(\mathbf{z})^T Q =$  $\tilde{Q}\mathcal{L}(\mathbf{z}), \, \forall \, \mathbf{z} \in \mathbb{C}^n$ . (For example, the latter condition is satisfied with  $Q = \tilde{Q} = I$ if  $\mathcal{L}$  is symmetric, or with  $Q = \tilde{Q} = J$  if  $\mathcal{L}$  is persymmetric). Assume  $AX - XA = \sum_m \mathbf{x}_m \mathbf{y}_m^T$ . Then

$$\sum_{m} \mathbf{x}_{m}^{T} \mathcal{L}(\mathbf{y}_{m})^{T} = \mathbf{0}^{T}$$

 $(\sum_{m} \mathbf{x}_{m}^{T} \mathcal{L}(\mathbf{y}_{m})^{T} \mathbf{e}_{k} = \ldots = 0), \text{ and thus}$ 

$$\begin{aligned} AX - XA &= \sum_{m} \mathbf{x}_{m} \mathbf{y}_{m}^{T} \\ &= \sum_{m} \mathbf{x}_{m} \mathbf{v}^{T} \mathcal{L}(\mathbf{y}_{m}) \\ &= \sum_{m} \mathbf{x}_{m} \mathbf{w}^{T} \mathcal{L}(\mathbf{y}_{m})^{-1} \mathcal{L}(\mathbf{y}_{m}) \\ &= \sum_{m} \mathbf{x}_{m} \mathbf{w}^{T} \mathcal{L}(\mathbf{y}_{m}) \mathcal{L}(\mathbf{w})^{-1} \\ &= \sum_{m} \mathbf{x}_{m} \mathbf{w}^{T} \mathcal{L}(\mathbf{y}_{m}) \mathcal{L}(\mathbf{w})^{-1} - \mathbf{b} \sum_{m} \mathbf{x}_{m}^{T} \mathcal{L}(\mathbf{y}_{m})^{T} Q \mathcal{L}(\mathbf{w})^{-1} \\ &= \sum_{m} \mathbf{x}_{m} \mathbf{w}^{T} \mathcal{L}(\mathbf{y}_{m}) \mathcal{L}(\mathbf{w})^{-1} - \mathbf{b} \sum_{m} \mathbf{x}_{m}^{T} \tilde{Q} \mathcal{L}(\mathbf{y}_{m}) \mathcal{L}(\mathbf{w})^{-1} \\ &= \sum_{m} (\mathbf{x}_{m} \mathbf{w}^{T} - \mathbf{b} \mathbf{x}_{m}^{T} \tilde{Q}) \mathcal{L}(\mathbf{y}_{m}) \mathcal{L}(\mathbf{w})^{-1} \\ &= \sum_{m} (\mathbf{x}_{m} \mathbf{w}^{T} - \mathbf{b} \mathbf{x}_{m}^{T} \tilde{Q}) \mathcal{L}(\mathcal{L}(\mathbf{w})^{T-1} \mathbf{y}_{m}) \end{aligned}$$

(w must be chosen such that  $\mathcal{L}(\mathbf{w})$  is non singular; it can be simply chosen equal to  $\mathbf{v}$  so that  $\mathcal{L}(\mathbf{w}) = I$ ).

We want to find  $Z_m$  ( $Z_m = R_m + E_m$ ) such that

$$Z_m X - X Z_m = \mathbf{x}_m \mathbf{w}^T - \mathbf{b} \mathbf{x}_m^T \tilde{Q} \tag{(*)}$$

so that  $A - \sum_{m} Z_m \mathcal{L}(\mathcal{L}(\mathbf{w})^{T^{-1}} \mathbf{y}_m)$  must be a matrix commuting with X. ...

Note that the left hand side in (\*) has zero trace. Thus, if  $\mathcal{L}$  is symmetric, since  $\tilde{Q} = I$ , then **b** must be chosen equal to **w** (otherwise the right hand side may have nonzero trace). If  $\mathcal{L}$  is persymmetric, since  $\tilde{Q} = J$ , then **b** must be chosen equal to  $J\mathbf{w}$  (for the same reason).

Symmetric case:

$$Z_m X - X Z_m = \mathbf{x}_m \mathbf{w}^T - \mathbf{w} \mathbf{x}_m^T.$$

Persymmetric case:

$$Z_m X - X Z_m = \mathbf{x}_m \mathbf{w}^T - J \mathbf{w} \mathbf{x}_m^T J.$$

NOTE: Assume moreover  $\mathcal{L}$  both symmetric and persymmetric. Then we

have also

$$JAJX - XJAJ = JAXJ - JXAJ = J(AX - XA)J$$
  

$$= \sum_{m} J\mathbf{x}_{m}\mathbf{y}_{m}^{T}J$$
  

$$= \sum_{m} J\mathbf{x}_{m}\mathbf{v}^{T}\mathcal{L}(\mathbf{y}_{m})J$$
  

$$= \sum_{m} J\mathbf{x}_{m}\mathbf{w}^{T}\mathcal{L}(\mathbf{y}_{m})\mathcal{L}(\mathbf{w})^{-1}$$
  

$$= \sum_{m} J\mathbf{x}_{m}\mathbf{w}^{T}J\mathcal{L}(\mathbf{y}_{m})\mathcal{L}(\mathbf{w})^{-1} - \mathbf{c}\sum_{m} \mathbf{x}_{m}^{T}\mathcal{L}(\mathbf{y}_{m})S\mathcal{L}(\mathbf{w})^{-1}$$
  

$$= \sum_{m} J\mathbf{x}_{m}\mathbf{w}^{T}J\mathcal{L}(\mathbf{y}_{m})\mathcal{L}(\mathbf{w})^{-1} - \mathbf{c}\sum_{m} \mathbf{x}_{m}^{T}\tilde{S}\mathcal{L}(\mathbf{y}_{m})\mathcal{L}(\mathbf{w})^{-1}$$
  

$$= \sum_{m} (J\mathbf{x}_{m}\mathbf{w}^{T}J - \mathbf{c}\mathbf{x}_{m}^{T}\tilde{S})\mathcal{L}(\mathbf{y}_{m})\mathcal{L}(\mathbf{w})^{-1}$$
  

$$= \sum_{m} (J\mathbf{x}_{m}\mathbf{w}^{T}J - \mathbf{c}\mathbf{x}_{m}^{T}\tilde{S})\mathcal{L}(\mathcal{L}(\mathbf{w})^{-1}\mathbf{y}_{m})$$

By summing this result with the previous one, we obtain

$$(A+JAJ)X - X(A+JAJ) = \sum_{m} (\mathbf{x}_{m}\mathbf{w}^{T} + J\mathbf{x}_{m}\mathbf{w}^{T}J - \mathbf{b}\mathbf{x}_{m}^{T}\tilde{Q} - \mathbf{c}\mathbf{x}_{m}^{T}\tilde{S})\mathcal{L}(\mathcal{L}(\mathbf{w})^{-1}\mathbf{y}_{m})$$

(w must be chosen such that  $\mathcal{L}(\mathbf{w})$  is non singular; it can be simply chosen equal to  $\mathbf{v}$  so that  $\mathcal{L}(\mathbf{w}) = I$ ).

We want to find  $Z_m$  ( $Z_m = R_m + E_m$ ) such that

$$Z_m X - X Z_m = \mathbf{x}_m \mathbf{w}^T + J \mathbf{x}_m \mathbf{w}^T J - \mathbf{b} \mathbf{x}_m^T \tilde{Q} - \mathbf{c} \mathbf{x}_m^T \tilde{S} \qquad (*')$$

(see page 209 of DiF,Zell, LAA 268 (1998) for a matrix  $Z_m$  satisfying (\*')) so that  $A + JAJ - \sum_m Z_m \mathcal{L}(\mathcal{L}(\mathbf{w})^{-1}\mathbf{y}_m)$  must be a matrix commuting with X. ... Note that if A is centrosymmetric, i.e. A=JAJ, (as in the case of A =symmetric Toeplitz matrix), then one can conclude that  $2A - \sum_m Z_m \mathcal{L}(\mathcal{L}(\mathbf{w})^{-1}\mathbf{y}_m)$  must be a matrix commuting with X, from which we have a representation for A...

Note that the left hand side in (\*) has zero trace. Thus, if  $S = \tilde{S} = I$  then  $\mathbf{c} = \mathbf{w}$  and  $Q = \tilde{Q}$  can be chosen either equal to I (in such case  $\mathbf{b} = \mathbf{w}$ ) or equal to J (in such case  $\mathbf{b} = J\mathbf{w}$ ). If  $S = \tilde{S} = J$  then  $\mathbf{c} = J\mathbf{w}$  and  $Q = \tilde{Q}$  can be chosen either equal to I (in such case  $\mathbf{b} = \mathbf{w}$ ) or equal to J (in such case  $\mathbf{b} = J\mathbf{w}$ ). Compare with page 209 of DiF,Zell, LAA 268 (1998).

Displacement decompositions involving spaces in  $\mathbb{V}$ 

1) Look for  $\tau = \tau_1(\mathbf{z})$  matrices of rank one.

2) Verify if there exist  $\mathbf{v}$  such that  $\tau_{\mathbf{v}}(A^T\mathbf{v}) = \tau_1(A^T\mathbf{v})\tau_1(\mathbf{v})^{-1} = \tau_1(\mathbf{z})$  with  $\mathbf{z}$  as in 1), and such that  $\tau_1(\mathbf{v})$  is non singular (or equivalently that  $\tau$  is a in  $\mathbb{V}$  for such  $\mathbf{v}$ ).

3) Find a displacement decomposition of the type

$$A = \Box + \tau_v(A^T \mathbf{v}), \ \mathbf{v}^T \Box = \mathbf{0}^T$$

where  $\mathbf{v}$  is as in 2) (so that  $\tau_v(A^T\mathbf{v})$  has rank 1 !). Attempt:

$$\mathbf{v}^T \sum_{m=1}^{\alpha} Z_m \mathcal{L}_m = \sum_{m=1}^{\alpha} (\mathbf{v}^T Z_m) \mathcal{L}_m = \sum_{m=1}^{\alpha} (\mathbf{v}^T \begin{bmatrix} \frac{1}{v_1} [\dots 1m] \\ \frac{1}{v_2} [\dots 2m] \\ \frac{1}{v_n} [\dots 2m] \end{bmatrix}) \mathcal{L}_m$$
$$[\dots 1m] + [\dots 2m] + \dots + [\dots nm] = \mathbf{x}_m^T$$

#### $\tau$ matrices of rank 1

First observe that the entry (1,1) of a non null  $\tau$  matrix A of rank 1 must be nonzero. In fact, if  $A_{11} = 0$ , then also  $A_{12} = 0$  because  $A_{12} \neq 0$  would imply  $A_{21} \neq 0$  (A is symmetric!) and so A would have rank at least 2. Thus  $A_{11} = 0$ implies  $A_{12} = 0$  and, by symmetry,  $A_{21} = 0$ . But then also  $A_{13}$  must be zero because  $A_{13} \neq 0$  would imply  $A_{31} \neq 0$  and so A would have rank at least 2. Going on one proves that the first row and column of A must be null, so A itself must be the null matrix.

A faster proof is the following: if  $A \in \tau$ ,  $A = \mathbf{u}\mathbf{v}^T$  and  $A_{11} = 0$ , then  $u_1v_1 = 0$ , and thus either  $u_1$  or  $v_1$  must be zero; but  $u_1 = 0$  ( $v_1=0$ ) implies  $\mathbf{e}_1^T A = \mathbf{0}^T$  ( $A\mathbf{e}_1 = \mathbf{0}$ ), that is, A = 0 since  $A \in \tau$ .

n odd:

 $\tau$  matrices A of rank one for n = 3: since A is in  $\tau$ , is symmetric, is persymmetric, and its second column is a multiple of its first column, it must be (unless a multiplier) of the type

$$A = \begin{bmatrix} 1 & \alpha \\ \alpha & 1 + \beta \\ \beta & \alpha \end{bmatrix}, \ \alpha^2 = 1 + \beta, \ \alpha\beta = \alpha.$$

We have two possible alternatives,  $\alpha = 0$ ,  $\beta = -1$ , and  $\beta = 1$ ,  $\alpha = \pm \sqrt{2}$  $(\alpha^2 - 2 = 0)$ :

$$A = \begin{bmatrix} 1 & 0 & -1 \\ 0 & 0 & 0 \\ -1 & 0 & 1 \end{bmatrix}, A = \begin{bmatrix} 1 & \pm\sqrt{2} & 1 \\ \pm\sqrt{2} & 2 & \pm\sqrt{2} \\ 1 & \pm\sqrt{2} & 1 \end{bmatrix}.$$

So, there are three  $3 \times 3$  rank one  $\tau$  matrices. Note that they are orthogonal with respect the inner product  $(\cdot, \cdot)_F$  or, equivalently, the three vectors that define them are orthogonal.

 $\tau$  matrices A of rank one for n = 5: since A is in  $\tau$ , is symmetric, is persymmetric, and its second column is a multiple of its first column, it must be (unless a multiplier) of the type

$$A = \begin{bmatrix} 1 & \alpha & & \\ \alpha & 1 + \beta & & \\ \beta & \alpha + \gamma & & \\ \gamma & \beta + \delta & & \\ \delta & \gamma & & \end{bmatrix}, \ \alpha^2 = 1 + \beta, \ \alpha\beta = \alpha + \gamma, \ \alpha\gamma = \beta + \delta, \ \alpha\delta = \gamma.$$

First observe that  $\delta \neq 0$  since  $\delta = 0$  would imply  $\gamma = \beta = 0$ ,  $\alpha = 0$ ,  $\beta = -1$ . So, the conditions become

$$\delta \alpha^2 = \delta + \beta \delta, \ \alpha \beta = \alpha + \alpha \delta, \ \alpha^2 \delta = \beta + \delta, \ \alpha \delta = \gamma$$

The first and the third imply  $\beta = \beta \delta$ . We have two possible alternatives. For  $\beta = 0$  we have the equation  $\alpha^2 - 1 = 0$  and thus  $\alpha = \pm 1$ ,  $\beta = 0$ ,  $\gamma = \mp 1$ ,  $\delta = -1$ . For  $\beta \neq 0$  we have the equation  $\alpha^3 - 3\alpha = 0$  and thus either  $\alpha = 0$ ,

$$\begin{split} \delta &= 1, \, \beta = -1, \, \gamma = 0 \text{ or } \alpha = \pm \sqrt{3}, \, \delta = 1, \, \beta = 2, \, \gamma = \pm \sqrt{3}. \\ A &= \begin{bmatrix} 1 & \pm 1 & 0 \\ \pm 1 & 1 & 0 \\ 0 & 0 & 0 \\ \mp 1 & -1 & 0 \\ -1 & \mp 1 & 0 \end{bmatrix}, \, \begin{bmatrix} 1 & 0 & -1 \\ 0 & 0 & 0 \\ -1 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & -1 \end{bmatrix}, \, \begin{bmatrix} 1 & \pm \sqrt{3} & 2 \\ \pm \sqrt{3} & 3 & \pm 2\sqrt{3} \\ 2 & \pm 2\sqrt{3} & 4 \\ \pm \sqrt{3} & 3 & \pm 2\sqrt{3} \\ 1 & \pm \sqrt{3} & 2 \end{bmatrix} \end{split}$$

So, there are five  $5 \times 5$  rank one  $\tau$  matrices. Note that they are orthogonal with respect the inner product  $(\cdot, \cdot)_F$  or, equivalently, the five vectors that define them are orthogonal.

 $\tau$  matrices A of rank one for n = 7: since A is in  $\tau$ , is symmetric, is persymmetric, and its second column is a multiple of its first column, it must be (unless a multiplier) of the type

$$A = \begin{bmatrix} 1 & \alpha & & \\ \alpha & 1 + \beta & & \\ \beta & \alpha + \gamma & & \\ \gamma & \beta + \delta & & \\ \delta & \gamma + \sigma & & \\ \sigma & \delta + \rho & \\ \rho & \sigma & & \end{bmatrix}, \ \alpha^2 = 1 + \beta, \ \alpha\beta = \alpha + \gamma, \ \alpha\gamma = \beta + \delta, \ \alpha\delta = \gamma + \sigma, \ \alpha\sigma = \delta + \rho, \ \alpha\rho = \sigma.$$

.

First observe that  $\rho \neq 0$  since  $\rho = 0$  would imply  $\sigma = \delta = \gamma = \beta = 0$ ,  $\alpha = 0$ ,  $\beta = -1$ . So, the conditions become

$$\rho\alpha^{2} = \rho + \beta\rho, \ \alpha\beta = \alpha + \gamma, \ \alpha\gamma = \beta + \delta, \ \alpha\delta = \gamma + \alpha\rho, \ \alpha^{2}\rho = \delta + \rho, \ \alpha\rho = \sigma$$

The second implies  $\gamma = \alpha(\beta - 1)$ . The first and the fifth imply  $\delta = \beta \rho$ . So that the third and the fourth become  $\alpha \gamma = \beta(1 + \rho)$ ,  $\alpha \rho(\beta - 1) = \gamma$ . It follows that  $\alpha \rho(\beta - 1) = \alpha(\beta - 1)$ .

We have three possible alternatives. For  $\alpha = 0$  we have  $\alpha = 0$ ,  $\beta = -1$ ,  $\gamma = 0$ ,  $\delta = 1$ ,  $\sigma = 0$ ,  $\rho = -1$ . For  $\beta = 1$  we have  $\alpha = \pm\sqrt{2}$ ,  $\beta = 1$ ,  $\gamma = 0$ ,  $\delta = -1$ ,  $\sigma = \pm\sqrt{2}$ ,  $\rho = -1$ . For  $\rho = 1$  we have  $\alpha^2 = 2 \pm \sqrt{2}$ ,  $\beta = 1 \pm \sqrt{2}$ ,  $\gamma = \alpha(\pm\sqrt{2})$ ,  $\delta = \beta$ ,  $\sigma = \alpha$ ,  $\rho = 1$ .

$$A = \begin{bmatrix} 1 & 0 & -1 & 0 & & \\ 0 & 0 & 0 & 0 & & \\ -1 & 0 & 1 & 0 & & \\ 0 & 0 & 0 & 0 & & \\ 1 & 0 & -1 & 0 & & \\ 0 & 0 & 0 & 0 & & \\ -1 & 0 & 1 & 0 & & \end{bmatrix}, \begin{bmatrix} 1 & \pm\sqrt{2} & 1 & 0 & & \\ \pm\sqrt{2} & 2 & \pm\sqrt{2} & 0 & \\ 1 & \pm\sqrt{2} & 1 & 0 & & \\ 0 & 0 & 0 & 0 & & \\ -1 & \mp\sqrt{2} & -1 & 0 & & \\ \mp\sqrt{2} & -2 & \mp\sqrt{2} & 0 & \\ -1 & \mp\sqrt{2} & -1 & 0 & & \end{bmatrix},$$

$$\begin{bmatrix} 1 & \alpha & 1 \pm \sqrt{2} & \alpha(\pm\sqrt{2}) \\ \alpha & \alpha^2 & \alpha(1 \pm \sqrt{2}) & \alpha^2(\pm\sqrt{2}) \\ 1 \pm \sqrt{2} & \alpha(1 \pm \sqrt{2}) & 3 \pm 2\sqrt{2} & \alpha(\pm\sqrt{2} + 2) \\ \alpha(\pm\sqrt{2}) & \alpha^2(\pm\sqrt{2}) & \alpha(\pm\sqrt{2} + 2) & 2\alpha^2 \\ 1 \pm \sqrt{2} & \alpha(1 \pm \sqrt{2}) & 3 \pm 2\sqrt{2} & \alpha(\pm\sqrt{2} + 2) \\ \alpha & \alpha^2 & \alpha(1 \pm \sqrt{2}) & \alpha^2(\pm\sqrt{2}) \\ 1 & \alpha & 1 \pm \sqrt{2} & \alpha(\pm\sqrt{2}) \end{bmatrix}, \ \alpha^2 = 2 \pm \sqrt{2}.$$

So, there are seven  $7 \times 7$  rank one  $\tau$  matrices. Note that they are orthogonal with respect the inner product  $(\cdot, \cdot)_F$  or, equivalently, the seven vectors that define them are orthogonal.

Another important remark. If  $A \in \tau$  is such that  $\mathbf{e}_2^T A = \alpha \mathbf{e}_1^T A$ , then  $\forall i$  there exist  $\xi_i$  such that  $\mathbf{e}_i^T A = \xi_i \mathbf{e}_1^T A$ . In other words, in order to make a  $\tau$  matrix of rank one it is sufficient to impose that its second row (column) is a multiple of its first row (column).

proof: Let  $\mathbf{z}^T$  be the first row of A. Then  $A = \tau(\mathbf{z})$ . The proof is by induction on i:

$$\mathbf{e}_{i}^{T}A = \mathbf{e}_{i}^{T}\tau(\mathbf{z}) = \mathbf{z}^{T}\tau(\mathbf{e}_{i}) = \mathbf{z}^{T}(\tau(\mathbf{e}_{i-1})\tau(\mathbf{e}_{2}) - \tau(\mathbf{e}_{i-2}))$$
  
$$= \mathbf{e}_{i-1}^{T}\tau(\mathbf{z})\tau(\mathbf{e}_{2}) - \mathbf{e}_{i-2}^{T}\tau(\mathbf{z}) = \xi_{i-1}\mathbf{z}^{T}\tau(\mathbf{e}_{2}) - \xi_{i-2}\mathbf{z}^{T}$$
  
$$= \xi_{i-1}\xi_{2}\mathbf{z}^{T} - \xi_{i-2}\mathbf{z}^{T} = (\xi_{i-1}\xi_{2} - \xi_{i-2})\mathbf{z}^{T}.$$

So,  $\xi_1 = 1$ ,  $\xi_2 = \alpha$ ,  $\xi_i = \xi_{i-1}\xi_2 - \xi_{i-2}$ ,  $i = 3, \dots, n$ .

n even:

n=2:

Assume that  $A \in \mathbb{C}^{2 \times 2}$  is in  $\tau$ . Then, since it is symmetric, is persymmetric, and its second column is a multiple of its first column, it must be (unless a multiplier) of the type

$$A = \begin{bmatrix} 1 & \alpha \\ \alpha & 1 \end{bmatrix}, \ \alpha^2 = 1.$$

$$(1): \qquad A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}, \\ \alpha - 1 = 0$$

$$(-1): \qquad A = \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}$$

$$\alpha + 1 = 0$$

So, there are two  $2 \times 2$  rank one  $\tau$  matrices. Note that they are orthogonal with respect the inner product  $(\cdot, \cdot)_F$  or, equivalently, the two vectors that define them are orthogonal.

n = 4:

Assume that  $A \in \mathbb{C}^{4 \times 4}$  is in  $\tau$ . Then, since it is symmetric, is persymmetric, and its second column is a multiple of its first column, it must be (unless a multiplier) of the type

$$A = \begin{bmatrix} 1 & \alpha \\ \alpha & 1 + \beta \\ \beta & \alpha + \gamma \\ \gamma & \beta \end{bmatrix}, \ \alpha^2 = 1 + \beta, \ \alpha\beta = \alpha + \gamma, \ \alpha\gamma = \beta.$$

Observe that  $\gamma \neq 0$  since  $\gamma = 0$  would imply  $\alpha = \beta = 0$ ,  $\beta = -1$ . So, the conditions become  $\gamma \alpha^2 = \gamma + \alpha \gamma^2$ ,  $\alpha^2 \gamma = \alpha + \gamma$ ,  $\alpha \gamma = \beta$ . They imply  $\alpha \gamma^2 = \alpha$ , and thus (since  $\alpha \neq 0$ )  $\gamma^2 = 1$ .

So, we have necessarily either  $\gamma = 1$ ,  $\beta = \alpha$  or  $\gamma = -1$ ,  $\beta = -\alpha$ . Such cases will be referred respectively (1) and (-1):

(1): 
$$A = \begin{bmatrix} 1 \\ \alpha \\ \alpha \\ 1 \end{bmatrix},$$
$$\alpha(\alpha) = 1 + \alpha, \text{ or }$$
$$g_2^+(\alpha) = \alpha(\alpha - 1) - 1 = \alpha^2 - \alpha - 1 = 0, \ \alpha = \frac{1 \pm \sqrt{5}}{2}$$

 $(\alpha(\alpha))$  must be equal to  $1+\alpha$  by the cross-sum condition applied for (i, j) = (2, 1)

$$\begin{aligned} (-1): \quad A = \begin{bmatrix} 1 \\ \alpha \\ -\alpha \\ -1 \end{bmatrix}, \\ \alpha(\alpha) = 1 - \alpha \text{ or } \\ g_2^-(\alpha) = \alpha(\alpha+1) - 1 = \alpha^2 + \alpha - 1 = 0, \ \alpha = \frac{-1\pm\sqrt{5}}{2} \end{aligned}$$

 $(\alpha(\alpha))$  must be equal to  $1 + (-\alpha)$  by the cross-sum condition applied for (i, j) = (2, 1).

Note that the zeros of  $g_2^-$  are the opposite of the zeros of  $g_2^+$ .

So, there are four  $4 \times 4$  rank one  $\tau$  matrices. Note that they are orthogonal with respect the inner product  $(\cdot, \cdot)_F$  or, equivalently, the four vectors that define them are orthogonal.

$$n = 6:$$

$$A = \begin{bmatrix} 1 & \alpha & & \\ \alpha & 1 + \beta & \\ \beta & \alpha + \gamma & \\ \gamma & \beta + \delta & \\ \delta & \gamma + \sigma & \\ \sigma & \delta & \end{bmatrix},$$

$$\alpha^{2} = 1 + \beta, \ \alpha\beta = \alpha + \gamma, \ \alpha\gamma = \beta + \delta, \ \alpha\delta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\sigma = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \sigma, \ \alpha\beta = \beta + \delta, \ \alpha\beta = \gamma + \delta, \ \alpha\beta = \gamma + \delta, \ \alpha\beta = \beta + \delta, \ \beta\beta = \beta + \delta, \$$

The above conditions imply necessarily either  $\sigma = 1$ ,  $\delta = \alpha$ ,  $\gamma = \beta$  or  $\sigma = -1$ ,  $\delta = -\alpha$ ,  $\gamma = -\beta$ . Such cases will be referred respectively (1) and (-1):

δ.

$$(1): \quad A = \begin{bmatrix} 1 \\ \alpha \\ \alpha^2 - 1 \\ \alpha^2 - 1 \\ \alpha \\ 1 \end{bmatrix}, \\ \alpha(\alpha^2 - 1) = \alpha + (\alpha^2 - 1) \text{ or } \\ g_3^+(\alpha) = \alpha(\alpha^2 - \alpha - 1) - (\alpha - 1) = \alpha^3 - \alpha^2 - 2\alpha + 1 = 0$$

 $(\alpha(\alpha^2-1))$  must be equal to  $\alpha + (\alpha^2-1)$  by the cross-sum condition applied for

(i, j) = (3, 1))

$$(-1): \quad A = \begin{bmatrix} 1 \\ \alpha \\ \alpha^2 - 1 \\ -(\alpha^2 - 1) \\ \alpha(\alpha^2 - 1) = \alpha - (\alpha^2 - 1) \text{ or } \\ g_3^-(\alpha) = \alpha(\alpha^2 + \alpha - 1) - (\alpha + 1) = \alpha^3 + \alpha^2 - 2\alpha - 1 = 0 \end{bmatrix},$$

 $(\alpha(\alpha^2-1)$  must be equal to  $\alpha + (-(\alpha^2-1))$  by the cross-sum condition applied for (i, j) = (3, 1).

Note that the zeros of  $g_3^-$  are the opposite of the zeros of  $g_3^+$ . Note, moreover, that the zeros of  $g_3^+$  ( $g_3^-$ ) are distinct.

So, there are six linearly independent  $6 \times 6$  rank one  $\tau$  matrices. Computer says that if  $\alpha_k^{\pm}$ , k = 1, 2, 3, are the zeros of  $g_3^{\pm}$ , then

$$1 + \alpha_k^{\pm} \alpha_s^{\pm} + ((\alpha_k^{\pm})^2 - 1)((\alpha_s^{\pm})^2 - 1) = 0,$$

i.e. the vector  $[1 \ \alpha_k^{\pm} \ (\alpha_k^{\pm})^2 - 1 \ \pm ((\alpha_k^{\pm})^2 - 1) \ \pm \alpha_k^{\pm} \ \pm 1]^T$  is orthogonal to the vector  $[1 \ \alpha_s^{\pm} \ (\alpha_s^{\pm})^2 - 1 \ \pm ((\alpha_s^{\pm})^2 - 1) \ \pm \alpha_s^{\pm} \ \pm 1]^T$ . It follows that such six  $6 \times 6$  rank one  $\tau$  matrices are orthogonal with respect the inner product  $(\cdot, \cdot)_F$  (because the six vectors that define them are orthogonal).

Question:

$$\begin{array}{c} \alpha^{3} - \alpha^{2} - 2\alpha + 1 = 0 \\ \beta^{3} - \beta^{2} - 2\beta + 1 = 0 \end{array} \right\} \ \Rightarrow \ 1 + \alpha\beta + (\alpha^{2} - 1)(\beta^{2} - 1) = 0$$

? Answer: since the roots are: 1.802,  $-1.25,\,0.445$  (Tommaso  $18/11/2010),\,\mathrm{it}$  seems yes

$$\begin{split} n &= 8: \\ A &= \begin{bmatrix} 1 & \alpha & & \\ \alpha & 1 + \beta & \\ \beta & \alpha + \gamma & \\ \gamma & \beta + \delta & \\ \delta & \gamma + \sigma & \\ \sigma & \delta + \rho & \\ \rho & \sigma + x & \\ x & \rho & \\ \alpha^2 &= 1 + \beta, \ \alpha\beta &= \alpha + \gamma, \ \alpha\gamma &= \beta + \delta, \ \alpha\delta &= \gamma + \sigma, \ \alpha\sigma &= \delta + \rho, \ \alpha\rho &= \sigma + x, \ \alpha x = \rho. \end{split}$$

The above conditions imply necessarily either x = 1,  $\rho = \alpha$ ,  $\sigma = \beta$ ,  $\delta = \gamma$  or x = -1,  $\rho = -\alpha$ ,  $\sigma = -\beta$ ,  $\delta = -\gamma$ . Such cases will be referred respectively (1)

and (-1):

 $g_4^-$ 

$$(1): \quad A = \begin{bmatrix} 1 \\ \alpha \\ \alpha^{2} - 1 \\ \alpha^{3} - 2\alpha \\ \alpha^{3} - 2\alpha \\ \alpha^{2} - 1 \\ \alpha \\ 1 \end{bmatrix},$$
$$(1): \quad A = \begin{bmatrix} \alpha^{3} - 2\alpha \\ \alpha^{2} - 1 \\ \alpha \\ 1 \end{bmatrix},$$
$$(1): \quad A = \begin{bmatrix} \alpha^{3} - 2\alpha \\ \alpha^{2} - 1 \\ \alpha \\ 1 \end{bmatrix},$$
$$(1): \quad A = \begin{bmatrix} \alpha^{3} - 2\alpha \\ \alpha^{3} - 2\alpha \\ \alpha^{2} - 1 \\ \alpha \\ 1 \end{bmatrix},$$
$$(1): \quad A = \begin{bmatrix} \alpha^{3} - 2\alpha \\ \alpha^{3} - 2\alpha \\ \alpha^{2} - 1 \\ \alpha \\ 1 \end{bmatrix},$$
$$(1): \quad A = \begin{bmatrix} \alpha^{3} - 2\alpha \\ \alpha^{3} - 2\alpha \\ \alpha^{3} - 2\alpha \\ \alpha^{2} - 1 \\ \alpha \\ 1 \end{bmatrix},$$
$$(1): \quad A = \begin{bmatrix} \alpha^{3} - 2\alpha \\ \alpha^{3} - 2\alpha \\ \alpha^{2} - 1 \\ \alpha \\ 1 \end{bmatrix},$$
$$(1): \quad A = \begin{bmatrix} \alpha^{3} - 2\alpha \\ \alpha^{3} - 2\alpha \\ \alpha^{2} - 1 \\ \alpha \\ 1 \end{bmatrix},$$
$$(1): \quad A = \begin{bmatrix} \alpha^{3} - 2\alpha \\ \alpha^{3} - 2\alpha \\ \alpha^{2} - 1 \\ \alpha \\ 1 \end{bmatrix},$$
$$(1): \quad A = \begin{bmatrix} \alpha^{3} - 2\alpha \\ \alpha^{3}$$

 $(\alpha(\alpha^3-2\alpha)$  must be equal to  $(\alpha^2-1)+(\alpha^3-2\alpha)$  by the cross-sum condition applied for (i,j)=(4,1))

$$(-1): \quad A = \begin{bmatrix} 1 \\ \alpha \\ \alpha^2 - 1 \\ \alpha^3 - 2\alpha \\ -(\alpha^3 - 2\alpha) \\ -(\alpha^2 - 1) \\ \alpha(\alpha^3 - 2\alpha) = (\alpha^2 - 1) - (\alpha^3 - 2\alpha) \text{ or } \\ (\alpha) = \alpha(\alpha^3 + \alpha^2 - 2\alpha - 1) - (\alpha^2 + \alpha - 1) = \alpha^4 + \alpha^3 - 3\alpha^2 - 2\alpha + 1 \\ = (\alpha + 1)(\alpha^3 - 3\alpha + 1) = 0 \end{bmatrix},$$

 $(\alpha(\alpha^3-2\alpha)$  must be equal to  $(\alpha^2-1)+(-(\alpha^3-2\alpha))$  by the cross-sum condition applied for (i,j)=(4,1)).

Note that the zeros of  $g_4^-$  are the opposite of the zeros of  $g_4^+$ . ??? FROM HERE Note, moreover, that the zeros of  $g_4^+$  ( $g_4^-$ ) are distinct.

So, there are eight linearly independent  $8 \times 8$  rank one  $\tau$  matrices. Computer says that if  $\alpha_k^{\pm}$ , k = 1, 2, 3, 4, are the zeros of  $g_4^{\pm}$ , then

$$1 + \alpha_k^{\pm} \alpha_s^{\pm} + ((\alpha_k^{\pm})^2 - 1)((\alpha_s^{\pm})^2 - 1) + ((\alpha_k^{\pm})^3 - 2\alpha_k^{\pm})((\alpha_s^{\pm})^3 - 2\alpha_s^{\pm}) = 0,$$

i.e. the vector  $[1 \ \alpha_k^{\pm} \ (\alpha_k^{\pm})^2 - 1 \ (\alpha_k^{\pm})^3 - 2\alpha_k^{\pm} \ \pm ((\alpha_k^{\pm})^3 - 2\alpha_k^{\pm}) \ \pm ((\alpha_k^{\pm})^2 - 1) \ \pm \alpha_k^{\pm} \ \pm 1]^T$  is orthogonal to the vector  $[1 \ \alpha_s^{\pm} \ (\alpha_s^{\pm})^2 - 1 \ (\alpha_s^{\pm})^3 - 2\alpha_s^{\pm} \ \pm ((\alpha_s^{\pm})^3 - 2\alpha_s^{\pm}) \ \pm ((\alpha_s^{\pm})^2 - 1) \ \pm \alpha_s^{\pm} \ \pm 1]^T$ . It follows that such six  $8 \times 8$  rank one  $\tau$  matrices are orthogonal with respect the inner product  $(\cdot, \cdot)_F$  (because the eight vectors that define them are orthogonal).TO HERE I HAVE TO CHECK ???

Set  $p_0(\alpha) = 1$ ,  $p_1(\alpha) = \alpha$ ,  $p_{i+1}(\alpha) = \alpha p_i(\alpha) - p_{i-1}(\alpha)$ ,  $i = 1, 2, ..., (p_i(\alpha))$  is the characteristic polynomial of the  $i \times i$  upper-left submatrix of  $P_0 + P_0^T$ ). We observe, in general, that for n even generic we have the following n rank one  $\tau$ matrices:

(1): 
$$(\mathbf{u}_{k}^{+})(\mathbf{u}_{k}^{+})^{T}, \ \mathbf{u}_{k}^{+} = \begin{bmatrix} \mathbf{x}_{k}^{+} \\ J\mathbf{x}_{k}^{+} \end{bmatrix}, \ \mathbf{x}_{k}^{+} = \begin{bmatrix} p_{0}(\alpha_{k}^{+}) \\ p_{1}(\alpha_{k}^{+}) \\ \\ p_{2}(\alpha_{k}^{+}) \end{bmatrix}, \ k = 1, \dots, \frac{n}{2},$$
  
 $\alpha_{k}^{+} \operatorname{zeri} \operatorname{di} \alpha p_{\frac{n}{2}-1}(\alpha) = p_{\frac{n}{2}-1}(\alpha) + p_{\frac{n}{2}-2}(\alpha)$ 

If the polynomial

$$g_{\frac{n}{2}}^{+}(\alpha) = \alpha p_{\frac{n}{2}-1}(\alpha) - p_{\frac{n}{2}-1}(\alpha) - p_{\frac{n}{2}-2}(\alpha)$$

has distinct real zeros (I have to check if this is true), then the vectors  $\mathbf{u}_k^+$  are linearly independent. (Such polynomial should coincide with the polynomial obtained in the particular cases (1) n = 2, 4, 6, 8).

$$(-1): (\mathbf{u}_k^-)(\mathbf{u}_k^-)^T, \ \mathbf{u}_k^- = \begin{bmatrix} \mathbf{x}_k^- \\ -J\mathbf{x}_k^- \end{bmatrix}, \ \mathbf{x}_k^- = \begin{bmatrix} p_0(\alpha_k^-) \\ p_1(\alpha_k^-) \\ \\ p_{2-1}(\alpha_k^-) \end{bmatrix}, \ k = 1, \dots, \frac{n}{2},$$
$$\alpha_k^- \text{ zeri di } \alpha p_{\frac{n}{2}-1}(\alpha) = -p_{\frac{n}{2}-1}(\alpha) + p_{\frac{n}{2}-2}(\alpha)$$

If the polynomial

$$g_{\frac{n}{2}}(\alpha) = \alpha p_{\frac{n}{2}-1}(\alpha) + p_{\frac{n}{2}-1}(\alpha) - p_{\frac{n}{2}-2}(\alpha)$$

has distinct real zeros (I have to check if this is true), then the vectors  $\mathbf{u}_k^-$  are linearly independent. (Such polynomial should coincide with the polynomial obtained in the particular cases (-1) n = 2, 4, 6, 8).

Note that the zeros of  $g_{\frac{n}{2}}^+$  are the opposite of the zeros of  $g_{\frac{n}{2}}^-$  (I have to check if this is true).

Note that  $(\mathbf{u}_k^+)^T(\mathbf{u}_s^-) = 0 \forall k, s$ . So, if the "distinct condition" on the zeros is satisfied, then  $\{(\mathbf{u}_k^+)(\mathbf{u}_k^+)^T, (\mathbf{u}_k^-)(\mathbf{u}_k^-)^T : k = 1, \dots, \frac{n}{2}\}$  is a set of n linearly independent rank one  $\tau$  matrices, and  $\tau = \text{Span}\{(\mathbf{u}_k^+)(\mathbf{u}_k^+)^T\} + \text{Span}\{(\mathbf{u}_k^-)(\mathbf{u}_k^-)^T\}$  with  $\text{Span}\{(\mathbf{u}_k^+)(\mathbf{u}_k^+)^T\}$  orthogonal to  $\text{Span}\{(\mathbf{u}_k^-)(\mathbf{u}_k^-)^T\}$ . For n = 4 it also happens that  $(\mathbf{u}_1^+)^T(\mathbf{u}_2^+) = 0 = (\mathbf{u}_1^-)^T(\mathbf{u}_2^-)$ . So, for n = 4

For n = 4 it also happens that  $(\mathbf{u}_1^+)^T(\mathbf{u}_2^+) = 0 = (\mathbf{u}_1^-)^T(\mathbf{u}_2^-)$ . So, for n = 4 we have 4 *orthogonal* rank one  $\tau$  matrices.

Is for n = 6 yet true that  $(\mathbf{u}_k^+)^T(\mathbf{u}_s^+) = 0 = (\mathbf{u}_k^-)^T(\mathbf{u}_s^-) \ s \neq k \ s, k = 1, \dots, 3$ ? In other words, is for n = 6 the basis  $\{(\mathbf{u}_k^+)(\mathbf{u}_k^+)^T, (\mathbf{u}_k^-)(\mathbf{u}_k^-)^T : k = 1, 2, 3\}$  of  $\tau$  an orthogonal basis ? SOLVED with Computer with "yes" (see above).

For n = 8?

Call  $\mathbf{u}_k$ ,  $k = 1, \ldots, n$ , the above (orthogonal) vectors  $\mathbf{u}_k^+$  and  $\mathbf{u}_k^-$ . Let  $\mathbf{v}$  be such that  $\tau_1(\mathbf{v}) = Sd(S^T\mathbf{v})d(S^T\mathbf{e}_1)^{-1}S^{-1}$  is invertible  $(d(S^T\mathbf{v}) \text{ non singular})$ , so that the matrix  $\tau_{\mathbf{v}}(\mathbf{z})$  is well defined. Note that  $\mathbf{v}^T\mathbf{u}_k \neq 0 \forall k \ldots$  (this fact is assured by the the assumption  $d(S^T\mathbf{v})$  non singular, since, we shall see, the  $\mathbf{u}_k$  are nothing else, unless a multiplier, the columns of S).

Since the  $\mathbf{u}_k \mathbf{u}_k^T$  form an (orthogonal) basis for  $\tau$ , there exist  $c_k$  such that  $\tau_{\mathbf{v}}(\mathbf{z}) = \sum_{k=1}^n c_k \mathbf{u}_k \mathbf{u}_k^T$ . We want to give a formula for such  $c_k$ .

$$\begin{aligned} Sd(S^T \mathbf{z})d(S^T \mathbf{v})^{-1}S^{-1} &= \sum_{k=1}^n c_k Sd(S^T (\mathbf{u}_k \mathbf{u}_k^T) \mathbf{v}) d(S^T \mathbf{v})^{-1}S^{-1} \\ &= \sum_{k=1}^n c_k (\mathbf{u}_k^T \mathbf{v}) Sd(S^T \mathbf{u}_k) d(S^T \mathbf{v})^{-1}S^{-1} \end{aligned}$$

if and only if

$$\mathbf{z} = \sum_{k=1}^{n} c_k(\mathbf{u}_k^T \mathbf{v}) \mathbf{u}_k.$$

Note that the  $\mathbf{u}_k^T \mathbf{v}$  must be all non zero; in fact, if one of them is zero, we would have that any vector  $\mathbf{z}$  (with *n* entries) can be written as a linear combination

of only n-1 vectors. For the  $c_k$  in the latter equality we can obtain an explicit formula; in fact, by the orthogonality of the  $\mathbf{u}_k$ ,

$$\mathbf{u}_s^T \mathbf{z} = \sum_k c_k (\mathbf{u}_k^T \mathbf{v}) \mathbf{u}_s^T \mathbf{u}_k = c_s (\mathbf{u}_s^T \mathbf{v}) \mathbf{u}_s^T \mathbf{u}_s.$$

Thus

$$c_s = rac{\mathbf{u}_s^T \mathbf{z}}{(\mathbf{u}_s^T \mathbf{v})(\mathbf{u}_s^T \mathbf{u}_s)}.$$

So

$$\tau_{\mathbf{v}}(\mathbf{z}) = \sum_{k=1}^{n} \frac{\mathbf{u}_{k}^{T} \mathbf{z}}{(\mathbf{u}_{k}^{T} \mathbf{v})(\mathbf{u}_{k}^{T} \mathbf{u}_{k})} \mathbf{u}_{k} \mathbf{u}_{k}^{T}$$

Question:  $\mathbf{u}_k^T \mathbf{u}_k = ?$ Question: is

$$\mathbf{v}^{T} = \begin{bmatrix} 1 \ 1 \ \cdots \ 1 \end{bmatrix} \begin{bmatrix} \mathbf{u}_{1} & \mathbf{u}_{n} \end{bmatrix}^{-1} = \begin{bmatrix} 1 \ 1 \ \cdots \ 1 \end{bmatrix} \begin{bmatrix} \frac{1}{\mathbf{u}_{1}^{T} \mathbf{u}_{1}} \mathbf{u}_{1}^{T} \\ \frac{1}{\mathbf{u}_{n}^{T} \mathbf{u}_{n}} \mathbf{u}_{n}^{T} \end{bmatrix}$$

(which is such that  $\mathbf{v}^T \mathbf{u}_k = 1 \ \forall k$ ) such that  $\tau_1(\mathbf{v})$  is invertible ? ... yes since in such case

$$[S^T \mathbf{v}]_i = [S^T [\frac{1}{\mathbf{u}_1^T \mathbf{u}_1} \mathbf{u}_1 \cdots \frac{1}{\mathbf{u}_n^T \mathbf{u}_n} \mathbf{u}_n] \begin{bmatrix} 1\\ 1 \end{bmatrix}]_i \neq 0, \ \forall i$$

and the  $\mathbf{u}_k$ , we shall see, are nothing else, unless a multiplier, the columns of S.

I think I have simply found again the sine transform S, that is the matrix  $\sin(ij\pi/(n+1)).$ 

More precisely, the vectors  $\mathbf{u}_k^+ \in \mathbf{u}_k^-$  that define my *n* rank one orthogonal  $\tau$  matrices  $((\mathbf{u}_k^+)(\mathbf{u}_k^+)^T \in (\mathbf{u}_k^-)^T, \ k = 1, ..., n/2)$  are nothing else, unless a multiplier, the columns of the sine matrix. I have observed this for small values of n.

Thus, it is obvious that the  $(\mathbf{u}_k^+)(\mathbf{u}_k^+)^T$ ,  $(\mathbf{u}_k^-)(\mathbf{u}_k^-)^T$ , k = 1, ..., n/2, are a basis, orthogonal with respect to  $(\cdot, \cdot)_F$ , of  $\tau$ . proof: if  $A \in \tau$ , then  $A = SDS^T = \sum_i D_{ii}(column \ i \ of \ S)(column \ i \ of \ S)^T$ . Even the fact that such n matrices  $(\mathbf{u}_k^+)(\mathbf{u}_k^+)^T$ ,  $(\mathbf{u}_k^-)(\mathbf{u}_k^-)^T$ , k = 1, ..., n/2,

are  $\tau$  matrices, is not a novelty.

proof: since for any diagonal D the matrix  $SDS^T$  belongs  $\tau$ , it is sufficient to choose  $D = \mathbf{e}_i \mathbf{e}_i^T$  in order to prove that the matrices (column i of S)(column i of S)<sup>T</sup> are in  $\tau$ .

Moreover, rank of  $A \in \tau$  is 1 if and only if rank of D in A = SDS is 1 if and only if  $D = \mathbf{e}_i \mathbf{e}_i^T$  unless a multiplier.

Consider the unitary sine matrix S. Note that the columns  $\mathbf{c}_j$  (j = 1, ..., n)of the matrix

$$S\sqrt{(n+1)/2}$$
 diag  $((1/\sin\frac{j\pi}{n+1}): j = 1,...,n)$ 

are orthogonal and their first entries are  $(\mathbf{c}_j)_1 = 1$ ,  $(\mathbf{c}_j)_2 = 2 \cos \frac{j\pi}{n+1}$ . Note that the  $(\mathbf{c}_j)_2$  coincide with the zeros of the two polynomials

$$g_{\frac{n}{2}}^{+}(\alpha) = \alpha p_{\frac{n}{2}-1}(\alpha) - p_{\frac{n}{2}-1}(\alpha) - p_{\frac{n}{2}-2}(\alpha),$$
  
$$g_{\frac{n}{2}}(\alpha) = \alpha p_{\frac{n}{2}-1}(\alpha) + p_{\frac{n}{2}-1}(\alpha) - p_{\frac{n}{2}-2}(\alpha).$$

The  $(\mathbf{c}_j)_2$  are also the eigenvalues of  $P_0 + P_0^T$ , i.e. the zeros of the polynomial  $p_n(\alpha)$  defined by the sequence

$$p_0(\alpha) = 1, \ p_1(\alpha) = \alpha, \ p_{i+1}(\alpha) = \alpha p_i(\alpha) - p_{i-1}(\alpha), \ i = 1, \dots, n-1.$$

Moreover, they are twice the stationary points of the chebycev polynomial  $T_{n+1}(\alpha)$ . More precisely, if  $T'_{n+1}(\alpha) = \sum_{s=0}^{n} a_s \alpha^s$ , then the  $(\mathbf{c}_j)_2$  are the zeros of the polynomial  $q_n(\alpha) = \sum_{s=0}^{n} \frac{1}{2^s} a_s \alpha^s$ . For example, since  $T_0(\alpha) = 1$ ,  $T_1(\alpha) = \alpha$ ,  $T_2(\alpha) = 2\alpha T_1(\alpha) - T_0(\alpha) = 2\alpha^2 - 1$ ,  $T_3(\alpha) = 2\alpha T_2(\alpha) - T_1(\alpha) = 4\alpha^3 - 3\alpha$ ,  $T_4(\alpha) = 8\alpha^4 - 8\alpha^2 + 1$ ,  $T_5(\alpha) = 16\alpha^5 - 20\alpha^3 + 5\alpha$ , we have  $T'_3(\alpha) = 12\alpha^2 - 3$ ,  $T'_5(\alpha) = 5(16\alpha^4 - 12\alpha^2 + 1)$ , so  $q_2(\alpha) = 3\alpha^2 - 3 = 3(\alpha^2 - 1)$ ,  $q_4(\alpha) = 5(\alpha^4 - 3\alpha^2 + 1)$ .

Verify that  $2\cos\frac{j\pi}{7}$ ,  $j = 1, \ldots, 6$ , are the zeros of

$$g_3^+(\alpha) = \alpha(\alpha^2 - \alpha - 1) - (\alpha - 1) = \alpha^3 - \alpha^2 - 2\alpha + 1 = 0 Tommaso$$
  
$$g_3^-(\alpha) = \alpha(\alpha^2 + \alpha - 1) - (\alpha + 1) = \alpha^3 + \alpha^2 - 2\alpha - 1 = 0$$

Verify that  $2\cos\frac{j\pi}{9}$ ,  $j = 1, \ldots, 8$ , are the zeros of

 $\begin{array}{l}g_{4}^{+}(\alpha) = \alpha(\alpha^{3} - \alpha^{2} - 2\alpha + 1) - (\alpha^{2} - \alpha - 1) = \alpha^{4} - \alpha^{3} - 3\alpha^{2} + 2\alpha + 1 = (\alpha - 1)(\alpha^{3} - 3\alpha - 1)\\g_{4}^{-}(\alpha) = \alpha(\alpha^{3} + \alpha^{2} - 2\alpha - 1) - (\alpha^{2} + \alpha - 1) = \alpha^{4} + \alpha^{3} - 3\alpha^{2} - 2\alpha + 1 = (\alpha + 1)(\alpha^{3} - 3\alpha + 1)\end{array}$ 

Set  $p_{-1}(\alpha) = 0$ , and

$$p_{0}(\alpha) = 1, \ g_{1}^{\pm}(\alpha) = (\alpha \mp 1)p_{0}(\alpha) - p_{-1}(\alpha),$$
  

$$p_{1}(\alpha) = \alpha, \ g_{2}^{\pm}(\alpha) = (\alpha \mp 1)p_{1}(\alpha) - p_{0}(\alpha),$$
  

$$p_{i}(\alpha) = \alpha p_{i-1}(\alpha) - p_{i-2}(\alpha), \ g_{i+1}^{\pm}(\alpha) = (\alpha \mp 1)p_{i}(\alpha) - p_{i-1}(\alpha), \ i = 1, \dots$$

Note that  $p_i(\alpha)$  is the characteristic polynomial of the  $i \times i$  upper-left submatrix of  $P_0 + P_0^T$ .

Introduce also the polynomials:

$$f_0^{\pm}(\alpha) = 1, \ f_1^{\pm}(\alpha) = \alpha \mp 1, \ f_i^{\pm}(\alpha) = \alpha f_{i-1}^{\pm}(\alpha) - f_{i-2}^{\pm}(\alpha), \ i = 2, \dots$$

Note that  $f_i^+(\alpha)$   $(f_i^-(\alpha))$  is the characteristic polynomial of the  $i \times i$  upper-left submatrix of  $P_0 + P_0^T + \mathbf{e_1}\mathbf{e}_1^T$   $(P_0 + P_0^T - \mathbf{e_1}\mathbf{e}_1^T)$ . Then  $g_i^{\pm} = f_i^{\pm}$ . proof: for i = 0, i = 1 it is true:

$$g_1^{\pm}(\alpha) = (\alpha \mp 1)1 - 0 = \alpha \mp 1 = f_1^{\pm}(\alpha), g_2^{\pm}(\alpha) = (\alpha \mp 1)\alpha - 1 = \alpha(\alpha \mp 1) - 1 = f_2^{\pm}(\alpha).$$

Assume the thesis true, and let us show that  $g_{i+1}^{\pm} = f_{i+1}^{\pm}$ :

$$g_{i+1}^{\pm}(\alpha) = (\alpha \mp 1)p_i(\alpha) - p_{i-1}(\alpha) = (\alpha \mp 1)(\alpha p_{i-1}(\alpha) - p_{i-2}(\alpha)) - p_{i-1}(\alpha)$$
  
=  $\alpha(\alpha \mp 1)p_{i-1}(\alpha) - \alpha p_{i-2}(\alpha) \pm p_{i-2}(\alpha) - p_{i-1}(\alpha)$   
=  $\alpha g_i^{\pm}(\alpha) \pm p_{i-2}(\alpha) - \alpha p_{i-2}(\alpha) + p_{i-3}(\alpha)$   
=  $\alpha g_i^{\pm}(\alpha) - (p_{i-2}(\alpha)(\alpha \mp 1) - p_{i-3}(\alpha))$   
=  $\alpha g_i^{\pm}(\alpha) - g_{i-1}^{\pm}(\alpha) = \alpha f_i^{\pm}(\alpha) - f_{i-1}^{\pm}(\alpha)$   
=  $f_{i+1}^{\pm}(\alpha)$ 

For the characteristic polynomial  $p_{2i}$  of the  $2i \times 2i$  matrix  $P_0 + P_0^T$  we have  $p_{2i}(\alpha) = f_i^-(\alpha)f_i^+(\alpha)$  (prove it!). So, the 2i zeros  $2\cos\frac{j\pi}{2i+1}$ ,  $j = 1, \ldots, 2i$ , of  $p_{2i}$  are the zeros of  $f_i^+$  and of  $f_i^-$ .

proof: the proof is by induction. The basis of the induction is true:

$$p_0(\alpha) = 1 = f_0^+(\alpha)f_0^-(\alpha), \ f_0^+(\alpha) = 1, \ f_0^-(\alpha) = 1,$$

$$p_2(\alpha) = \alpha^2 - 1 = f_1^+(\alpha)f_1^-(\alpha), \ f_1^+(\alpha) = \alpha - 1, \ f_1^-(\alpha) = \alpha + 1,$$

 $p_4(\alpha) = \alpha^4 - 3\alpha^2 + 1 = f_2^+(\alpha)f_2^-(\alpha), \ f_2^+(\alpha) = \alpha^2 - \alpha - 1, \ f_2^-(\alpha) = \alpha^2 + \alpha - 1.$ Assume  $p_{2j}(\alpha) = f_j^+(\alpha)f_j^-(\alpha), \ j = 0, 1, \dots, i-1.$  Then

$$p_{2i}(\alpha) = \alpha p_{2i-1}(\alpha) - p_{2i-2}(\alpha) = (\alpha^2 - 1)p_{2i-2}(\alpha) - \alpha p_{2i-3}(\alpha) = *$$

(use the identity:  $p_{2i-1} = \alpha p_{2i-2} - p_{2i-3}$ ). Since the identities

$$p_{2i-4}(\alpha) = \alpha p_{2i-5}(\alpha) - p_{2i-6}(\alpha), \ \alpha p_{2i-3}(\alpha) = \alpha^2 p_{2i-4}(\alpha) - \alpha p_{2i-5}(\alpha)$$

imply  $p_{2i-4}(\alpha) + \alpha p_{2i-3}(\alpha) = \alpha^2 p_{2i-4}(\alpha) - p_{2i-6}(\alpha)$ , we have the equality  $p_{2i-3}(\alpha) = \frac{1}{\alpha}((\alpha^2 - 1)p_{2i-4}(\alpha) - p_{2i-6}(\alpha))$ . So, \* becomes:

$$* = (\alpha^{2} - 1)p_{2i-2}(\alpha) - ((\alpha^{2} - 1)p_{2i-4}(\alpha) - p_{2i-6}(\alpha)) = (\alpha^{2} - 1)(p_{2i-2}(\alpha) - p_{2i-4}(\alpha)) + p_{2i-6}(\alpha)$$

Now, by the inductive hypothesis,

$$\begin{aligned} p_{2i}(\alpha) &= (\alpha^2 - 1)(f_{i-1}^+(\alpha)f_{i-1}^-(\alpha) - f_{i-2}^+(\alpha)f_{i-2}^-(\alpha)) + f_{i-3}^+(\alpha)f_{i-3}^-(\alpha) \\ &= \alpha^2 f_{i-1}^+(\alpha)f_{i-1}^-(\alpha) + f_{i-2}^+(\alpha)f_{i-2}^-(\alpha) - \alpha^2 f_{i-2}^+(\alpha)f_{i-2}^-(\alpha) \\ &\quad -f_{i-1}^+(\alpha)f_{i-1}^-(\alpha) + f_{i-2}^+(\alpha)f_{i-2}^-(\alpha) - \alpha^2 f_{i-2}^+(\alpha)f_{i-2}^-(\alpha) \\ &= \alpha^2 f_{i-1}^+(\alpha)f_{i-1}^-(\alpha) + (\alpha f_{i-2}^+(\alpha) - f_{i-1}^+(\alpha))(\alpha f_{i-2}^-(\alpha) - f_{i-1}^-(\alpha)) \\ &= \alpha^2 f_{i-1}^+(\alpha)f_{i-1}^-(\alpha) + f_{i-2}^+(\alpha)f_{i-2}^-(\alpha) - \alpha(f_{i-2}^+(\alpha)f_{i-1}^-(\alpha) + f_{i-1}^+(\alpha)f_{i-2}^-(\alpha)) \\ &= (\alpha f_{i-1}^+(\alpha) - f_{i-2}^+(\alpha))(\alpha f_{i-1}^-(\alpha) - f_{i-2}^-(\alpha)) \\ &= f_i^+(\alpha)f_i^-(\alpha) \end{aligned}$$

*Exercise.* In n is even, then the eigenvalues of the  $n \times n$  matrix  $P_0 + P_0^T$  are the eigenvalues of the following two  $\frac{n}{2} \times \frac{n}{2}$  matrices:

[ 1	1		-		-1	1		-	
1	0	·			1	0	·		
$\left[\begin{array}{c}1\\1\end{array}\right]$	·	1	$\begin{array}{c}1\\0\end{array}$	,	$\begin{bmatrix} -1\\ 1 \end{bmatrix}$	·	1	$\begin{array}{c}1\\0\end{array}$	

Let us come back to the odd case. Recall:

n = 3

$$A = \begin{bmatrix} 1\\0\\-1 \end{bmatrix}, \begin{bmatrix} 1\\\pm\sqrt{2}\\1 \end{bmatrix}$$

Note on the second entries: 0 is eigenvalue of [0];  $\pm\sqrt{2}$  are eigenvalues of  $\begin{bmatrix} 0 & 2\\ 1 & 0 \end{bmatrix}$ .

$$n = 5$$

$$\begin{bmatrix} 1\\ \pm 1\\ 0\\ \mp 1\\ -1 \end{bmatrix}, \begin{bmatrix} 1\\ 0\\ -1\\ 0\\ 1 \end{bmatrix}, \begin{bmatrix} 1\\ \pm\sqrt{3}\\ 2\\ \pm\sqrt{3}\\ 1 \end{bmatrix}$$
Note on the second entries:  $\pm 1$  are eigenvalues of  $\begin{bmatrix} 0& 1\\ 1& 0 \end{bmatrix}$ ; 0 and  $\pm\sqrt{3}$  are eigenvalues of  $\begin{bmatrix} 0& 2& 0\\ 1& 0& 1\\ 0& 1& 0 \end{bmatrix}$ .  

$$n = 7$$

$$\begin{bmatrix} 1\\ 0\\ -1\\ 0\\ 1\\ 0\\ -1 \end{bmatrix}, \begin{bmatrix} 1\\ \pm\sqrt{2}\\ 1\\ 0\\ -1\\ \mp\sqrt{2}\\ -1 \end{bmatrix}, \begin{bmatrix} 1\\ \alpha\\ \alpha^{2}-1\\ \alpha^{3}-2\alpha\\ \alpha^{2}-1\\ \alpha\\ 1 \end{bmatrix}, \alpha = \pm\sqrt{2-\sqrt{2}}, \alpha = \pm\sqrt{2+\sqrt{2}}$$
Note on the second entries: 0 and  $\pm\sqrt{2}$  are eigenvalues of  $\begin{bmatrix} 0& 1& 0\\ 1& 0& 1\\ 0& 1& 0 \end{bmatrix}$ ;  $\pm\sqrt{2-\sqrt{2}}$ ,  $\pm\sqrt{2+\sqrt{2}}$  are eigenvalues of  $\begin{bmatrix} 0& 1& 0\\ 1& 0& 1\\ 0& 1& 0 \end{bmatrix}$ ;  $\pm\sqrt{2-\sqrt{2}}$ ,  $\pm\sqrt{2+\sqrt{2}}$  are eigenvalues of  $\begin{bmatrix} 0& 2& 0& 0\\ 1& 0& 1& 0\\ 0& 1& 0& 1\\ 0& 0& 1& 0 \end{bmatrix}$ .

For n odd generic we have

$$A = \begin{bmatrix} 1\\ \alpha\\ \alpha^2 - 1\\ p_{\frac{n-3}{2}}(\alpha)\\ p_{\frac{n-1}{2}}(\alpha)\\ \rho p_{\frac{n-3}{2}}(\alpha)\\ \rho(\alpha^2 - 1)\\ \rho \alpha\\ \rho \end{bmatrix}$$

and necessarily  $\rho = \pm 1$ . If  $\rho = 1$  we have the further condition  $\alpha p_{\frac{n-1}{2}}(\alpha) = 2p_{\frac{n-3}{2}}(\alpha)$ , or  $0 = g_{\frac{n+1}{2}}(\alpha) := \alpha p_{\frac{n-1}{2}}(\alpha) - 2p_{\frac{n-3}{2}}(\alpha)$ . If  $\rho = -1$ , we have the further conditions  $p_{\frac{n-1}{2}}(\alpha) + p_{\frac{n-5}{2}}(\alpha) = \alpha p_{\frac{n-3}{2}}(\alpha)$ ,  $p_{\frac{n-1}{2}}(\alpha) - p_{\frac{n-5}{2}}(\alpha) = -\alpha p_{\frac{n-3}{2}}(\alpha)$  which imply  $p_{\frac{n-1}{2}}(\alpha) = 0$ . Viceversa, such conditions  $(0 = g_{\frac{n+1}{2}}(\alpha)$ , in case  $\rho = 1$ , and  $p_{\frac{n-1}{2}}(\alpha) = 0$ , in case  $\rho = -1$ ) imply that the second column of A is  $\alpha$  times the first one.

So, we have the following Rank one  $\tau$  matrices.

$$(\mathbf{u}_{k}^{+})(\mathbf{u}_{k}^{+})^{T}, \ \mathbf{u}_{k}^{+} = \begin{bmatrix} \mathbf{x}_{k}^{+} \\ p_{\frac{n-1}{2}}(\alpha_{k}^{+}) \\ J\mathbf{x}_{k}^{+} \end{bmatrix}, \ \mathbf{x}_{k}^{+} = \begin{bmatrix} 1 \\ \alpha_{k}^{+} \\ (\alpha_{k}^{+})^{2} - 1 \\ p_{\frac{n-3}{2}}(\alpha_{k}^{+}) \end{bmatrix}, \ k = 1, \dots, \frac{n+1}{2},$$
  
$$\alpha_{k}^{+} \text{ roots of } g_{\frac{n+1}{2}}(\alpha) = \alpha p_{\frac{n-1}{2}}(\alpha) - 2p_{\frac{n-3}{2}}(\alpha).$$

Prove that  $g_{\frac{n+1}{2}}(\alpha)$  is the characteristic polynomial of the  $\frac{n+1}{2} \times \frac{n+1}{2}$  matrix

$$\left[\begin{array}{rrrr} 0 & 2 & 0 \\ 1 & 0 & 1 \\ 0 & 1 \\ \end{array}\right]$$

$$(\mathbf{u}_k^-)(\mathbf{u}_k^-)^T, \ \mathbf{u}_k^- = \begin{bmatrix} \mathbf{x}_k^-\\ 0\\ -J\mathbf{x}_k^- \end{bmatrix}, \ \mathbf{x}_k^- = \begin{bmatrix} 1\\ \alpha_k^-\\ (\alpha_k^-)^2 - 1\\ p_{\frac{n-3}{2}}(\alpha_k^-) \end{bmatrix}, \ k = 1, \dots, \frac{n-1}{2},$$

 $\alpha_k^-$  roots of  $p_{\frac{n-1}{2}}(\alpha)$ 

*Exercise*. Prove that  $n \text{ odd} \Rightarrow p_n(\alpha) = p_{\frac{n-1}{2}}(\alpha)g_{\frac{n+1}{2}}(\alpha)$ .

Set  $f_0(\alpha) = 1$ ,  $f_1(\alpha) = \alpha$ ,  $f_2(\alpha) = \alpha^2 - 2$ ,  $f_{i+1}(\alpha) = \alpha f_i(\alpha) - f_{i-1}(\alpha)$ ,  $i = 2, 3, \ldots$  Note that  $f_i(\lambda)$  is the characteristic polynomial of the upper-left  $i \times i$  submatrix of  $P_0 + \mathbf{e}_1 \mathbf{e}_2^T$ .

Set  $g_0(\alpha) = 1$ ,  $g_{i+1}(\alpha) = \alpha p_i(\alpha) - 2p_{i-1}(\alpha)$ ,  $i = 0, 1, \dots$  Then  $g_i(\alpha) = f_i(\alpha)$ ,  $i = 0, 1, \dots$  So the roots of  $g_i(\alpha)$  are those of  $f_i(\alpha)$ :  $\pm \sqrt{2}$  (i = 2),  $0, \pm \sqrt{3}$   $(i = 3), \pm \sqrt{2 - \sqrt{2}}, \pm \sqrt{2 + \sqrt{2}}$   $(i = 4), \dots$  or, more in general,  $2 \cos \frac{j\pi}{2i}$ ,  $j = 1, 3, \dots, 2i - 1$ .

proof: the proof is by induction.  $g_1(\alpha) = \alpha p_0(\alpha) - 2p_{-1}(\alpha) = \alpha = f_1(\alpha),$  $g_2(\alpha) = \alpha p_1(\alpha) - 2p_0(\alpha) = \alpha^2 - 2 = f_2(\alpha),$ 

$$g_{i+1}(\alpha) = \alpha p_i(\alpha) - 2p_{i-1}(\alpha) = \alpha(\alpha p_{i-1}(\alpha) - p_{i-2}(\alpha)) - 2(\alpha p_{i-2}(\alpha) - p_{i-3}(\alpha)) = \alpha(\alpha p_{i-1}(\alpha) - 2p_{i-2}(\alpha)) - \alpha p_{i-2}(\alpha) + 2p_{i-3}(\alpha) = \alpha g_i(\alpha) - g_{i-1}(\alpha) = \alpha f_i(\alpha) - f_{i-1}(\alpha) = f_{i+1}(\alpha).$$

 $p_{2i+1}(\alpha) = p_i(\alpha)f_{i+1}(\alpha), i = 0, 1, \dots$ proof: the proof is by induction. The basis of the induction is true:

$$p_1(\alpha) = \alpha = p_0(\alpha)f_1(\alpha), \ p_3(\alpha) = \alpha^3 - 2\alpha = \alpha(\alpha^2 - 2) = p_1(\alpha)f_2(\alpha), p_5(\alpha) = \alpha^5 - 4\alpha^3 + 3\alpha = (\alpha^2 - 1)(\alpha^3 - 3\alpha) = p_2(\alpha)f_3(\alpha).$$

Assume  $p_{2j+1}(\alpha) = p_j(\alpha) f_{j+1}(\alpha), \ j = 0, 1, ..., i - 1$ . Then

$$p_{2i+1} = \alpha p_{2i} - p_{2i-1} = (\alpha^2 - 1)p_{2i-1} - \alpha p_{2i-2} = (\alpha^2 - 1)(p_{2i-1} - p_{2i-3}) + p_{2i-5}$$

(note that  $p_{2i+1} = \alpha p_{2i} - p_{2i-1}$  and  $\alpha p_{2i+2} = \alpha^2 p_{2i+1} - \alpha p_{2i}$  imply  $p_{2i+1} + \alpha p_{2i+2} = \alpha^2 p_{2i+1} - p_{2i-1}$ , and thus  $p_{2i+2} = \frac{1}{\alpha}((\alpha^2 - 1)p_{2i+1} - p_{2i}))$ . By the inductive assumption:

$$p_{2i+1} = (\alpha^2 - 1)(p_{i-1}f_i - p_{i-2}f_{i-1}) + p_{i-3}f_{i-2}$$

$$= \alpha^2 p_{i-1}f_i + p_{i-2}f_{i-1} - \alpha^2 p_{i-2}f_{i-1}$$

$$-p_{i-1}f_i + p_{i-2}f_{i-1} - \alpha^2 p_{i-2}f_{i-1}$$

$$-p_{i-1}f_i + (\alpha p_{i-2} - p_{i-1})$$

$$(\alpha f_{i-1} - f_i)$$

$$= \alpha^2 p_{i-1}f_i + p_{i-2}f_{i-1}$$

$$-\alpha p_{i-2}f_i - \alpha p_{i-1}f_{i-1}$$

$$= (\alpha p_{i-1} - p_{i-2})(\alpha f_i - f_{i-1})$$

$$= p_i f_{i+1}.$$

First row of the two tridiagonal matrices whose eigenvalues, collected together, give the eigenvalues of the  $n \times n$  matrix  $P_0 + P_0^T$  (the *i*th row,  $i \ge 2$ , is like the *i*th row,  $i \ge 2$ , of  $P_0 + P_0^T$ ):

 $\begin{array}{lll} n \mbox{ even:} \\ n/2 \times n/2 & 1 & 1 \ (+ \mbox{ or } 1 \ \mbox{in the bottom}), \\ n/2 \times n/2 & -1 & 1 \ (- \mbox{ or } -1 \ \mbox{in the bottom}) \\ n \ \mbox{odd:} \\ (n-1)/2 \times (n-1)/2 & 0 & 1 \ (- \ \mbox{or } -1 \ \mbox{in the bottom}), \\ (n+1)/2 \times (n+1)/2 & 0 & 2 \ (+ \ \mbox{or } 1 \ \mbox{in the bottom}) \end{array}$ 

If we assume  $\alpha_k^+$  and  $\alpha_k^-$  both ordered in decreasing order  $(\ldots \leq \alpha_2^+ \leq \alpha_1^+, i.e. \ \alpha_k^+ = 2\cos\frac{(2k-1)\pi}{n+1}, \ldots \leq \alpha_2^- \leq \alpha_1^-, i.e. \ \alpha_k^- = 2\cos\frac{(2k)\pi}{n+1}$ ), then, taking the first, the  $[\frac{n+3}{2}]$ th, the second, the  $[\frac{n+5}{2}]$ th, ..., columns of the following matrix

$$[\mathbf{u}_1^+ \cdots \mathbf{u}_{\lfloor \frac{n+1}{2} \rfloor}^+ \mathbf{u}_1^- \cdots \mathbf{u}_{\lfloor \frac{n}{2} \rfloor}^-]$$

one obtains the sine matrix normalized so that the entries on its first row are all equal to 1.

proof:

$$\frac{\sin \frac{2j\pi}{n+1}}{\sin \frac{3j\pi}{n+1}} = 2\cos \frac{j\pi}{n+1}$$
$$\frac{\sin \frac{3j\pi}{n+1}}{\sin \frac{j\pi}{n+1}} = (2\cos \frac{j\pi}{n+1})^2 - 1$$

Conclusion. Assume n odd. If  $A \in \tau$  is of rank one then there are  $\alpha$  and x such

that

$$A\mathbf{e}_{1} = \begin{bmatrix} 1\\ \alpha\\ \alpha^{2}-1\\ p_{\frac{n-3}{2}}(\alpha)\\ p_{\frac{n-1}{2}}(\alpha) = xp_{\frac{n-1}{2}}(\alpha)\\ xp_{\frac{n-3}{2}}(\alpha)\\ x(\alpha^{2}-1)\\ x\alpha\\ x \end{bmatrix}$$

Moreover, the following two conditions must be satisfied

$$(x-1)p_{\frac{n-1}{2}}(\alpha) = 0, \ \alpha p_{\frac{n-1}{2}}(\alpha) = (x+1)p_{\frac{n-3}{2}}(\alpha).$$

Note that x = 1 and  $p_{\frac{n-1}{2}}(\alpha) = 0$  cannot be simultaneously verified (otherwise we would have  $p_{\frac{n-3}{2}}(\alpha) = 0$  and roots of  $p_{\frac{n-1}{2}}$  are different from those of  $p_{\frac{n-3}{2}}!$ ). So, either

$$x = 1, \ 0 = f_{\frac{n+1}{2}} := \alpha p_{\frac{n-1}{2}}(\alpha) - 2p_{\frac{n-3}{2}}(\alpha)$$

or

$$x = -1, \ p_{\frac{n-1}{2}}(\alpha) = 0.$$

Assume *n* even. If  $A \in \tau$  is of rank one then there are  $\alpha$  and *x* such that

$$A\mathbf{e}_{1} = \begin{bmatrix} 1\\ \alpha\\ \alpha^{2} - 1\\ p_{\frac{n}{2} - 2}(\alpha)\\ p_{\frac{n}{2} - 1}(\alpha)\\ xp_{\frac{n}{2} - 1}(\alpha)\\ xp_{\frac{n}{2} - 2}(\alpha)\\ x(\alpha^{2} - 1)\\ x\alpha\\ x \end{bmatrix}.$$

Moreover, the following two conditions must be satisfied

$$\alpha p_{\frac{n}{2}-1}(\alpha) = p_{\frac{n}{2}-2}(\alpha) + x p_{\frac{n}{2}-1}(\alpha), \ \alpha x p_{\frac{n}{2}-1}(\alpha) = p_{\frac{n}{2}-1}(\alpha) + x p_{\frac{n}{2}-2}(\alpha)$$

which become (since  $x \neq 0$ )

$$\alpha x p_{\frac{n}{2}-1}(\alpha) = x p_{\frac{n}{2}-2}(\alpha) + x^2 p_{\frac{n}{2}-1}(\alpha), \ \alpha x p_{\frac{n}{2}-1}(\alpha) = p_{\frac{n}{2}-1}(\alpha) + x p_{\frac{n}{2}-2}(\alpha)$$

and thus imply  $(x^2-1)p_{\frac{n}{2}-1}(\alpha) = 0$ ; but  $p_{\frac{n}{2}-1}(\alpha) = 0$  would imply  $p_{\frac{n}{2}-2}(\alpha) = 0$  (not possible! see above), so either

$$x = 1, \ f_{\frac{n}{2}}^+(\alpha) := (\alpha - 1)p_{\frac{n}{2}-1}(\alpha) - p_{\frac{n}{2}-2}(\alpha) = 0$$

or

$$x = -1, \ f_{\frac{n}{2}}(\alpha) := (\alpha + 1)p_{\frac{n}{2}-1}(\alpha) - p_{\frac{n}{2}-2}(\alpha) = 0.$$