

## Janossy Densities. II. Pfaffian Ensembles

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We extend the main result of the companion paper *J. Stat. Phys.* **113**:595–610 to the case of the pfaffian ensembles.

**KEY WORDS:** Random matrices; orthogonal polynomials; Janossy densities; pfaffian ensembles.

### 1. INTRODUCTION AND FORMULATION OF RESULTS

Let us consider a  $2n$ -particle pfaffian ensemble introduced by Rains in ref. 9: Let  $(X, \lambda)$  be a measure space,  $\phi_1, \phi_2, \dots, \phi_{2n}$  be complex-valued functions on  $X$ , and  $\epsilon(x, y)$  be an antisymmetric kernel such that

$$p(x_1, \dots, x_{2n}) = (1/Z_{2n}) \det(\phi_j(x_k))_{j, k=1, \dots, 2n} pf(\epsilon(x_j, x_k))_{j, k=1, \dots, 2n} \quad (1)$$

defines the density of a  $2n$ -dimensional probability distribution on  $X^{2n} = X \times \dots \times X$  with respect to the product measure  $\lambda^{\otimes 2n}$ . Ensembles of this form were introduced in refs. 9 and 11. We recall (see, e.g., ref. 5) that the pfaffian of a  $2n \times 2n$  antisymmetric matrix  $A = (a_{jk})$ ,  $j, k = 1, \dots, 2n$ ,  $a_{jk} = -a_{kj}$ , is defined as  $pf(A) = \sum_{\tau} (-1)^{\text{sign}(\tau)} a_{i_1 j_1} \dots a_{i_n j_n}$ , where the summation is over all partitions of the set  $\{1, \dots, 2m\}$  into disjoint pairs  $\{i_1, j_1\}, \dots, \{i_n, j_n\}$  such that  $i_k < j_k$ ,  $k = 1, \dots, n$ , and  $\text{sign}(\tau)$  is the sign of the permutation  $(i_1, j_1, \dots, i_n, j_n)$ . The normalization constant in (1) (usually called the partition function)

$$Z_{2n} = \int_{X^{2n}} \det(\phi_j(x_k))_{j, k=1, \dots, 2n} pf(\epsilon(x_j, x_k))_{j, k=1, \dots, 2n} \quad (2)$$

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can be shown to be equal to  $(2n)! pf(M)$ , where the  $2n \times 2n$  antisymmetric matrix  $M = (M_{jk})_{j,k=1,\dots,2n}$  is defined as

$$M_{jk} = \int_{X^2} \phi_j(x) \epsilon(x, y) \phi_k(y) \lambda(dx) \lambda(dy). \quad (3)$$

For the pfaffian ensemble (1) one can explicitly calculate  $k$ -point correlation functions

$$\begin{aligned} \rho_k(x_1, \dots, x_k) &:= ((2n)! / (2n-k)!) \\ &\times \int_{X^{2n-k}} p(x_1, \dots, x_k, x_{k+1}, \dots, x_{2n}) d\lambda(x_{k+1}) \cdots d\lambda(x_{2n}), \\ &k = 1, \dots, 2n \end{aligned}$$

and show that they have the pfaffian form (ref. 9)

$$\rho_k(x_1, \dots, x_k) = pf(K(x_i, x_j))_{i,j=1,\dots,k}, \quad (4)$$

where  $K(x, y)$  is the antisymmetric matrix kernel

$$\begin{aligned} &K(x, y) \\ &= \begin{pmatrix} \sum_{1 \leq j, k \leq 2n} \phi_j(x) M_{jk}^{-t} \phi_k(y) & \sum_{1 \leq j, k \leq 2n} \phi_j(x) M_{jk}^{-t} (\epsilon \phi_k)(y) \\ \sum_{1 \leq j, k \leq 2n} (\epsilon \phi_j)(x) M_{jk}^{-t} \phi_k(y) & -\epsilon(x, y) + \sum_{1 \leq j, k \leq 2n} (\epsilon \phi_j)(x) M_{jk}^{-t} (\epsilon \phi_k)(y) \end{pmatrix}, \end{aligned} \quad (5)$$

provided the matrix  $M$  is invertible (by definition  $(\epsilon \phi)(x) = \int_X \epsilon(x, y) \phi(y) \lambda(dy)$ ). If  $X \subset \mathbb{R}$  and  $\lambda$  is absolutely continuous with respect to the Lebesgue measure, then the probabilistic meaning of the  $k$ -point correlation functions is that of the density of probability to find an eigenvalue in each infinitesimal interval around points  $x_1, x_2, \dots, x_k$ . In other words

$$\begin{aligned} &\rho_k(x_1, x_2, \dots, x_k) \lambda(dx_1) \cdots \lambda(dx_k) \\ &= \Pr\{\text{there is a particle in each infinitesimal interval } (x_i, x_i + dx_i)\}. \end{aligned}$$

On the other hand, if  $\mu$  is supported by a discrete set of points, then

$$\begin{aligned} &\rho_k(x_1, x_2, \dots, x_k) \lambda(x_1) \cdots \lambda(x_k) \\ &= \Pr\{\text{there is a particle at each of the points } x_i, i = 1, \dots, k\}. \end{aligned}$$

In general, random point processes with the  $k$ -point correlation functions of the pfaffian form (4) are called pfaffian random point processes.<sup>(8)</sup> Pfaffian point processes include determinantal point processes<sup>(10)</sup> as a particular case when the matrix kernel has the form  $\begin{pmatrix} \epsilon & K \\ -K & 0 \end{pmatrix}$  where  $K$  is a scalar kernel and  $\epsilon$  is an antisymmetric kernel.

So-called Janossy densities  $\mathcal{J}_{k,I}(x_1, \dots, x_k)$ ,  $k = 0, 1, 2, \dots$ , describe the distribution of the eigenvalues in any given interval  $I$ . If  $X \subset \mathbb{R}$  and  $\lambda$  is absolutely continuous with respect to the Lebesgue measure then

$$\begin{aligned} \mathcal{J}_{k,I}(x_1, \dots, x_k) \lambda(dx_1) \cdots \lambda(dx_k) \\ = \Pr\{\text{there are exactly } k \text{ particles in } I, \text{ one in each of the } k \text{ distinct} \\ \text{infinitesimal intervals } (x_i, x_i + dx_i)\}. \end{aligned}$$

If  $\lambda$  is discrete then

$$\begin{aligned} \mathcal{J}_{k,I}(x_1, \dots, x_k) \\ = \Pr\{\text{there are exactly } k \text{ particles in } I, \text{ one at each of the } k \text{ points } x_i, \\ i = 1, \dots, k\}. \end{aligned}$$

See refs. 3 and 4 for details and additional discussion. For pfaffian point processes the Janossy densities also have the pfaffian form (see refs. 8 and 9) with an antisymmetric matrix kernel  $L_I$ :

$$\mathcal{J}_{k,I}(x_1, \dots, x_k) = \text{const}(I) \text{pf}(L_I(x_i, x_j))_{i,j=1,\dots,k}, \tag{6}$$

where

$$L_I = K_I(Id + JK_I)^{-1}, \tag{7}$$

$J = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$  and  $\text{const}(I) = \text{pf}(J - K_I)$  is the Fredholm pfaffian of the restriction of the operator  $K$  on the interval  $I$ , i.e.,  $\text{const}(I) = \text{pf}(J - K_I) = (\text{pf}(J + L_I))^{-1} = (\det(Id + JK_I))^{1/2} = (\det(Id - JL_I))^{-1/2}$ . (We refer the reader to ref. 9, Section 8 for the treatment of Fredholm pfaffians).

Let us define three  $2n \times 2n$  matrices  $G^I, M^I, M^{X \setminus I}$ :

$$G^I_{jk} = \int_I \phi_j(x) \int_X \epsilon(x, y) \phi_k(y) \lambda(dy) \lambda(dx), \tag{8}$$

$$M^I_{jk} = \int_{I^2} \phi_j(x) \epsilon(x, y) \phi_k(y) \lambda(dx) \lambda(dy), \tag{9}$$

$$M^{X \setminus I}_{jk} = \int_{(X \setminus I)^2} \phi_j(x) \epsilon(x, y) \phi_k(y) \lambda(dx) \lambda(dy) \tag{10}$$

(please compare (9) and (10) with the above formula (3) for  $M$ ). Throughout the paper we will assume that the matrices  $M^I$  and  $M^{X \setminus I}$  are invertible.

The main result of this paper is

**Theorem 1.1.** The kernel  $L_I$  has a form similar to the formula (5) for  $K$ . Namely,  $L_I$  is equal to

$$L_I(x, y) = \left( \begin{array}{l} \sum_{1 \leq j, k \leq 2n} \phi_j(x) (M^{X \setminus I})_{jk}^{-t} \phi_k(y) \\ \sum_{1 \leq j, k \leq 2n} (\epsilon_{X \setminus I} \phi_j)(x) (M^{X \setminus I})_{jk}^{-t} \phi_k(y) \\ \sum_{1 \leq j, k \leq 2n} \phi_j(x) (M^{X \setminus I})_{jk}^{-t} (\epsilon_{X \setminus I} \phi_k)(y) \\ -\epsilon_{X \setminus I}(x, y) + \sum_{1 \leq j, k \leq 2n} (\epsilon_{X \setminus I} \phi_j)(x) (M^{X \setminus I})_{jk}^{-t} (\epsilon_{X \setminus I} \phi_k)(y) \end{array} \right), \tag{11}$$

where  $\epsilon_{X \setminus I} \phi(x) = \int_{X \setminus I} \epsilon(x, y)^+ \phi(y) \lambda(dy)$ .

Comparing (11) with (5) one can see that the kernel  $L_I$  is constructed in the following way: 1) first it is constructed on  $X \setminus I$  by the same recipe used to construct the kernel  $K$  on the whole  $X$ ; 2) it is extended then to  $I$  (we recall that  $L_I$  acts on  $L^2(I, d\lambda(x))$ , not on  $L^2(X \setminus I, d\lambda(x))$ ).

This result contains as a special case Theorem 1.1 from the companion paper.<sup>(3)</sup> The rest of the paper is organized as follows. We discuss some interesting special cases of the theorem, namely so-called polynomial ensembles ( $\beta = 1, 2$ , and  $4$ ) in Section 2. The proof of the theorem is given in Section 3.

## 2. RANDOM MATRIX ENSEMBLES WITH $\beta = 1, 2, 4$

We follow the discussion in ref. 9 (see also refs. 11 and 12).

### *Biorthogonal Ensembles*

Consider the particle space to be the union of two identical measure spaces  $(V, \mu)$  and  $(W, \mu)$ :  $X = V \cup W, V \cap W = \emptyset$ . The configuration of  $2n$  particles in  $X$  will consist of  $n$  particles  $v_1, \dots, v_n$  in  $V$  and  $n$  particles  $w_1, \dots, w_n$  in  $W$  in such a way that the configurations of particles in  $V$  and  $W$  are identical (i.e.,  $v_j = w_j, j = 1, \dots, n$ ). Let  $\xi_j, \psi_j, j = 1, \dots, n$  be some functions on  $V$ . We define  $\{\phi_j\}$  and  $\epsilon$  in (1) so that  $\phi_j(v) = 0, v \in V, \phi_j(w) = \xi_j(w), w \in W, j = 1, \dots, n, \phi_j(v) = \psi_{j-n-1}(v), v \in V, \phi_j(w) = 0, w \in W, j = n+1, \dots, 2n$ , and  $\epsilon(v_1, v_2) = 0, v_1, v_2 \in V, \epsilon(w_1, w_2) = 0, w_1, w_2 \in W, \epsilon(v, w) = -\epsilon(w, v) = \delta_{vw}, v \in V, w \in W$ . The restriction of the

measure  $\lambda$  on both  $V$  and  $W$  is defined to be equal to  $\mu$ . Then (1) specializes into (see Corollary 1.5. in ref. 9)

$$p(v_1, \dots, v_n) = \text{const}_n \det(\xi_j(v_i))_{i,j=1,\dots,n} \det(\psi_j(v_i))_{i,j=1,\dots,n}. \tag{12}$$

Ensembles of the form (12) are known as biorthogonal ensembles (see refs. 1 and 7). The statement of the Theorem 1.1 in the case (12) has been proven in the companion paper.<sup>(3)</sup> The special case of the biorthogonal ensemble (12) when  $V = \mathbb{R}$ ,  $\xi_j(x) = \psi_j(x) = x^{j-1}$ , and  $W = \{\mathbb{C} \mid |z| = 1\}$ ,  $\xi_j(z) = \bar{\psi}_j(z) = z^{j-1}$ , such ensembles are well known in Random Matrix Theory as *unitary ensembles*, see ref. 6 for details. An ensemble of the form (12) which is different from random matrix ensembles was studied in ref. 7. We specifically want to single out the polynomial ensemble with  $\beta = 2$ .

### Polynomial ( $\beta=2$ ) Ensembles

Let  $X = \mathbb{R}$  or  $\mathbb{Z}$ ,  $\phi_j(x) = x^{j-1}$ ,  $j = 1, \dots, 2n$ , and  $\lambda(dx)$  has a density  $\omega(x)$  with respect to the reference measure on  $X$  (Lebesgue measure in the continuous case, counting measure in the discrete case). Then the formula (12) specializes into

$$p(v_1, \dots, v_n) = \text{const}_n \prod_{1 \leq i < j \leq n} (v_i - v_j)^2 \prod_{1 \leq j \leq n} \omega(v_j). \tag{13}$$

The next two ensembles we want to mention are the polynomial  $\beta = 1$  and 4 ensembles.

### Polynomial ( $\beta=1$ ) Ensembles

Let  $X = \mathbb{R}$  or  $\mathbb{Z}$ ,  $\phi_j(x) = x^{j-1}$ ,  $j = 1, \dots, 2n$ ,  $\epsilon(x, y) = \frac{1}{2} \text{sgn}(y - x)$ , and  $\lambda(dx)$  has a density  $\omega(x)$  with respect to the reference measure on  $X$  (Lebesgue measure in the continuous case, counting measure in the discrete case). Then the formula (1) specializes into the formula for the density of the joint distribution of  $2n$  particles in a so-called  $\beta = 1$  polynomial ensemble (see ref. 9, Remark 1):

$$p(x_1, \dots, x_{2n}) = \text{const}_n \prod_{1 \leq i < j \leq 2n} |x_i - x_j| \prod_{1 \leq j \leq 2n} \omega(x_j). \tag{14}$$

In Random Matrix Theory the ensembles (14) in the continuous case are known as *orthogonal ensembles*, see ref. 6.

### Polynomial ( $\beta=4$ ) Ensembles

Similar to the biorthogonal case ( $\beta = 2$ ) let us consider the particle space to be the union of two identical measure spaces  $(Y, \mu)$ ,  $(Z, \mu)$ ,  $X = Y \cup Z$ ,  $Y \cap Z = \emptyset$ , where  $Y = \mathbb{R}$  or  $Y = \mathbb{Z}$ . The configuration of  $2n$  particles  $x_1, \dots, x_{2n}$ , in  $X$  will consist of  $n$  particles  $y_1, \dots, y_n$  in  $Y$

and  $n$  particles  $z_1, \dots, z_n$ , in  $Z$  in such a way that the configurations of particles in  $Y$  and  $Z$  are identical. We define  $\{\phi_j\}$  and  $\epsilon$  so that  $\phi_j(y) = y^j$ ,  $y \in Y$ ,  $\phi_j(z) = jz^{j-1}$ ,  $z \in Z$ ,  $\epsilon(y_1, y_2) = 0$ ,  $\epsilon(z_1, z_2) = 0$ ,  $\epsilon(y, z) = -\epsilon(z, y) = \delta_{yz}$ . As above we assume that the measure  $\mu$  has a density  $\omega$  with respect to the reference measure on  $Y$ . Then the formula (1) specializes into the formula for the density of the joint distribution of  $n$  particles in a  $\beta = 4$  polynomial ensemble (see Corollary 1.3. in ref. 9)

$$p(y_1, \dots, y_n) = \text{const}_n \prod_{1 \leq i < j \leq n} (y_i - y_j)^4 \prod_{1 \leq j \leq n} \omega(y_j). \quad (15)$$

In Random Matrix Theory the ensembles (15) are known as *symplectic ensembles*, see ref. 6.

### 3. PROOF OF THE MAIN RESULT

Consider matrix kernels

$$\mathcal{K}_I = -JK_I, \quad \mathcal{L}_I = -JL_I. \quad (16)$$

Then the relation (7) simplifies into

$$\mathcal{L}_I = \mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1} \quad (17)$$

which is the same relation that is satisfied by the correlation and Janossy scalar kernels in the determinantal case.<sup>(2,4)</sup> The consideration of  $\mathcal{K}_I$  and  $\mathcal{L}_I$  is motivated by the fact that the pfaiffians of the  $2k \times 2k$  matrices with the antisymmetric matrix kernels  $K_I$  and  $L_I$  are equal to the quaternion determinants<sup>(6)</sup> of  $2k \times 2k$  matrices with the kernels  $\mathcal{K}_I, \mathcal{L}_I$  when the latter matrices are viewed as  $k \times k$  quaternion matrices (i.e., each quaternion entry corresponds to a  $2 \times 2$  block with complex entries). It follows from (5) and (16) that the kernel  $\mathcal{K}_I$  is given by the formula

$$\mathcal{K}_I = \sum_{j, k=1, \dots, 2n} M_{jk}^{-t} \begin{pmatrix} -(\epsilon\phi_j) \otimes \phi_k & -(\epsilon\phi_j) \otimes (\epsilon\phi_k) \\ \phi_j \otimes \phi_k & \phi_j \otimes (\epsilon\phi_k) \end{pmatrix} + \begin{pmatrix} 0 & \epsilon \\ 0 & 0 \end{pmatrix}. \quad (18)$$

Let us denote by  $\tilde{\mathcal{L}}_I$  the following kernel

$$\begin{aligned} \tilde{\mathcal{L}}_I(x, y) = & \sum_{1 \leq j, k \leq 2n} (M^{x \setminus I})_{jk}^{-t} \begin{pmatrix} -(\epsilon_{x \setminus I} \phi_j) \otimes \phi_k & -(\epsilon_{x \setminus I} \phi_j) \otimes (\epsilon_{x \setminus I} \phi_k) \\ \phi_j \otimes \phi_k & \phi_j \otimes (\epsilon_{x \setminus I} \phi_k) \end{pmatrix} \\ & + \begin{pmatrix} 0 & \epsilon \\ 0 & 0 \end{pmatrix}. \end{aligned} \quad (19)$$

As above,  $\epsilon\phi$  stands for  $\int_X \epsilon(x, y) \phi(y)$ . We use the notation  $\phi_j \otimes \phi_k$  as a shorthand for  $\phi_j(x) \phi_k(y)$ . To prove the main result of the paper we will show that  $\tilde{\mathcal{L}}_I = \mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1}$  (in other words we are going to prove that  $\tilde{\mathcal{L}}_I = \mathcal{L}_I$ , where  $\mathcal{L}_I$  is defined in (17)). The proof relies on Lemmas 1 and 2 given below. Let us introduce the notation  $(\epsilon_I \phi)(x) = \int_I \epsilon(x, y) \phi_s(y) d\lambda(y)$ . We will show that the finite-dimensional subspace  $\mathcal{H} = \text{Span}\left\{\begin{pmatrix} \epsilon\phi_s \\ -\phi_s \end{pmatrix}, \begin{pmatrix} \epsilon_I\phi_s \\ 0 \end{pmatrix}\right\}_{s=1, \dots, 2n}$  is invariant under  $\mathcal{K}_I$  and  $\tilde{\mathcal{L}}_I$ . The main part of the proof of the theorem is to show that  $\tilde{\mathcal{L}}_I = \mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1}$  holds on  $\mathcal{H}$ .

**Lemma 3.1.** The operators  $\mathcal{K}_I, \tilde{\mathcal{L}}_I$  leave  $\mathcal{H}$  invariant and  $\tilde{\mathcal{L}}_I = \mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1}$  holds on  $\mathcal{H}$ .

Below we give the proof of the lemma. Using the notations introduced above in (8)–(10) one can easily calculate

$$\mathcal{K}_I \begin{pmatrix} \epsilon\phi_s \\ 0 \end{pmatrix} = \sum_{j=1, \dots, 2n} -((G^I)^t M^{-1})_{sj} \begin{pmatrix} \epsilon\phi_j \\ -\phi_j \end{pmatrix} \tag{20}$$

$$\mathcal{K}_I \begin{pmatrix} 0 \\ -\phi_s \end{pmatrix} = \sum_{j=1, \dots, 2n} (G^I M^{-1})_{sj} \begin{pmatrix} \epsilon\phi_j \\ -\phi_j \end{pmatrix} - \begin{pmatrix} \epsilon_I\phi_s \\ 0 \end{pmatrix}. \tag{21}$$

Defining the  $2n \times 2n$  matrix  $T$  as

$$T_{sk} = \int_I \phi_s(x) \int_{X \setminus I} \epsilon(x, y) \phi_k(y) d\lambda(y) d\lambda(x) \tag{22}$$

we compute

$$\mathcal{K}_I \begin{pmatrix} \epsilon_I\phi_s \\ 0 \end{pmatrix} = \sum_{j=1, \dots, 2n} ((G^I - T) M^{-1})_{sj} \begin{pmatrix} \epsilon\phi_j \\ -\phi_j \end{pmatrix}, \tag{23}$$

where  $(G^I - T)_{sk} = M^I_{sk} = \int_{I^2} \phi_s(x) \epsilon(x, y) \phi_k(y) \lambda(dx) \lambda(dy)$ . One can rewrite Eqs. (20) and (21) as

$$\mathcal{K}_I \begin{pmatrix} \epsilon\phi_s \\ -\phi_s \end{pmatrix} = \sum_{j=1, \dots, 2n} ((G^I - (G^I)^t) M^{-1})^t_{sj} \begin{pmatrix} \epsilon\phi_j \\ -\phi_j \end{pmatrix} - \begin{pmatrix} \epsilon_I\phi_s \\ 0 \end{pmatrix}, \tag{24}$$

$$\mathcal{K}_I \begin{pmatrix} -\epsilon\phi_s \\ -\phi_s \end{pmatrix} = \sum_{j=1, \dots, 2n} ((G^I + (G^I)^t) M^{-1})_{sj} \begin{pmatrix} \epsilon\phi_j \\ -\phi_j \end{pmatrix} - \begin{pmatrix} \epsilon_I\phi_s \\ 0 \end{pmatrix}. \tag{25}$$

We conclude that that the subspace  $\mathcal{H}$  is indeed invariant under  $\mathcal{K}_I$  and the matrix of the restriction of  $\mathcal{K}_I$  on  $\mathcal{H}$  has the following block structure in the basis  $\left\{ \begin{pmatrix} \epsilon_s^{\phi_s} \\ -\phi_s \end{pmatrix}, \begin{pmatrix} -\epsilon_s^{\phi_s} \\ -\phi_s \end{pmatrix}, \begin{pmatrix} \epsilon_s^{\phi_s} \\ 0 \end{pmatrix} \right\}_{s=1, \dots, 2n}$ :

$$\begin{pmatrix} (G^I - (G^I)^t) M^{-1} & (G^I + (G^I)^t) M^{-1} & (G^I - T) M^{-1} \\ 0 & 0 & 0 \\ -\text{Id} & -\text{Id} & 0 \end{pmatrix} \quad (26)$$

(in particular  $\text{Ran}(\mathcal{K}_I|_{\mathcal{H}}) = \text{Span}\left\{ \begin{pmatrix} \epsilon_s^{\phi_s} \\ -\phi_s \end{pmatrix}, \begin{pmatrix} \epsilon_s^{\phi_s} \\ 0 \end{pmatrix} \right\}_{s=1, \dots, 2n}$ ). Let us introduce some additional notations:

$$A = (G^I - (G^I)^t) M^{-1}, \quad (27)$$

$$B = (G^I + (G^I)^t) M^{-1}, \quad (28)$$

$$C = (G^I - T) M^{-1}. \quad (29)$$

When a matrix has a block form

$$\mathcal{M} = \begin{pmatrix} A & B & C \\ 0 & 0 & 0 \\ -\text{Id} & -\text{Id} & 0 \end{pmatrix}$$

(as it is in our case) the matrix  $\mathcal{M}(\text{Id} - \mathcal{M})^{-1}$  has the block form

$$\begin{pmatrix} (\text{Id} - A + C)^{-1} - \text{Id} & (B - C)(\text{Id} - A + C)^{-1} & C(\text{Id} - A + C)^{-1} \\ 0 & 0 & 0 \\ -(\text{Id} - A + C)^{-1} & -\text{Id} - (B - C)(\text{Id} - A + C)^{-1} & -C(\text{Id} - A + C)^{-1} \end{pmatrix}. \quad (30)$$

As one can see from the formulas (31)–(33) the invertibility of  $\text{Id} - \mathcal{M}$  follows from the invertibility of  $M^{X \setminus I}$  which has been assumed throughout the paper. We have

$$(\text{Id} - A + C)^{-1} = M(M + (G^I)^t - T)^{-1} = M(M^{X \setminus I})^{-1} \quad (31)$$

$$C(\text{Id} - A + C)^{-1} = (G^I - T)(M + (G^I)^t - T)^{-1} = M^I(M^{X \setminus I})^{-1} \quad (32)$$

$$\begin{aligned} (B - C)(\text{Id} - A + C)^{-1} &= ((G^I)^t + T)(M + (G^I)^t - T)^{-1} \\ &= ((G^I)^t + T)(M^{X \setminus I})^{-1}. \end{aligned} \quad (33)$$



Let us now compute the matrix of the restriction of  $\tilde{\mathcal{L}}_I$  on  $\mathcal{H}$ . We have

$$\tilde{\mathcal{L}}_I = \sum_{j, k=1, \dots, 2n} (M^{X \setminus I})^t \begin{pmatrix} -(\epsilon_{X \setminus I} \phi_j) \otimes \phi_k & -(\epsilon_{X \setminus I} \phi_j) \otimes (\epsilon_{X \setminus I} \phi_k) \\ \phi_k \otimes \phi_k & \phi_j \otimes (\epsilon_{X \setminus I} \phi_k) \end{pmatrix} + \begin{pmatrix} 0 & \epsilon_{X \setminus I} \\ 0 & 0 \end{pmatrix}. \tag{34}$$

Similarly to the computations above one can see that  $\mathcal{H}$  is invariant under  $\tilde{\mathcal{L}}_I$  and

$$\begin{aligned} \tilde{\mathcal{L}}_I \begin{pmatrix} \epsilon \phi_s \\ -\phi_s \end{pmatrix} &= \sum_{j=1, \dots, 2n} ((T - (G^I)^t)(M^{X \setminus I})^{-1})_{sj} \begin{pmatrix} \epsilon \phi_j \\ -\phi_j \end{pmatrix} \\ &\quad - \sum_{1 \leq j \leq 2n} ((T - (G^I)^t)(M^{X \setminus I})^{-1})_{sj} \begin{pmatrix} \epsilon_I \phi_j \\ 0 \end{pmatrix} - \begin{pmatrix} \epsilon_I \phi_s \\ 0 \end{pmatrix}, \end{aligned} \tag{35}$$

$$\begin{aligned} \tilde{\mathcal{L}}_I \begin{pmatrix} -\epsilon \phi_s \\ -\phi_s \end{pmatrix} &= \sum_{j=1, \dots, 2n} ((T + (G^I)^t)(M^{X \setminus I})^{-1})_{sj} \begin{pmatrix} \epsilon \phi_j \\ -\phi_j \end{pmatrix} \\ &\quad - \sum_{1 \leq j \leq 2n} ((T + (G^I)^t)(M^{X \setminus I})^{-1})_{sj} \begin{pmatrix} \epsilon_I \phi_j \\ 0 \end{pmatrix} - \begin{pmatrix} \epsilon_I \phi_s \\ 0 \end{pmatrix}, \end{aligned} \tag{36}$$

and

$$\begin{aligned} \tilde{\mathcal{L}}_I \begin{pmatrix} \epsilon_I \phi_s \\ 0 \end{pmatrix} &= \sum_{j=1, \dots, 2n} ((G^I - T)(M^{X \setminus I})^{-1})_{sj} \begin{pmatrix} \epsilon \phi_j \\ -\phi_j \end{pmatrix} \\ &\quad - \sum_{j=1, \dots, 2n} ((G^I - T)(M^{X \setminus I})^{-1})_{sj} \begin{pmatrix} \epsilon_I \phi_j \\ 0 \end{pmatrix}. \end{aligned} \tag{37}$$

Therefore the restriction of  $\tilde{\mathcal{L}}_I$  to  $\mathcal{H}$  in the basis  $\{(\begin{smallmatrix} \epsilon \phi_s \\ -\phi_s \end{smallmatrix}), (\begin{smallmatrix} -\epsilon \phi_s \\ -\phi_s \end{smallmatrix}), (\begin{smallmatrix} \epsilon_I \phi_s \\ 0 \end{smallmatrix})\}_{s=1, \dots, 2n}$  has the following block structure

$$\begin{pmatrix} (T - (G^I)^t)(M^{X \setminus I})^{-1} & (T + (G^I)^t)(M^{X \setminus I})^{-1} & (G^I - T)(M^{X \setminus I})^{-1} \\ 0 & 0 & 0 \\ -\text{Id} - (T - (G^I)^t)(M^{X \setminus I})^{-1} & -\text{Id} - (T + (G^I)^t)(M^{X \setminus I})^{-1} & -(G^I - T)(M^{X \setminus I})^{-1} \end{pmatrix}. \tag{38}$$

Comparing (30), (31)–(33), and (38) we see that  $\tilde{\mathcal{L}}_I = \mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1}$  on  $\mathcal{H}$ . Lemma 3.1 is proven.

To show that  $\tilde{\mathcal{L}}_I = \mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1}$  also holds on the complement of  $\mathcal{H}$  it is enough to prove it on the subspaces  $(\begin{smallmatrix} \mathcal{H}_1 \\ 0 \end{smallmatrix})^\perp$ , and  $(\begin{smallmatrix} 0 \\ \mathcal{H}_2 \end{smallmatrix})^\perp$ , where  $\mathcal{H}_1 = \text{Span}(\overline{\epsilon_I \phi_s})_{k=1, \dots, 2n}$  and  $\mathcal{H}_2 = \text{Span}(\overline{\phi_s})_{k=1, \dots, 2n}$ . The invertibility of the matrix  $M_I$  implies that actually it is enough to prove  $\tilde{\mathcal{L}}_I = \mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1}$  on the subspaces  $(\begin{smallmatrix} \mathcal{H}_2 \\ 0 \end{smallmatrix})^\perp$ , and  $(\begin{smallmatrix} 0 \\ \mathcal{H}_1 \end{smallmatrix})^\perp$ . Here we use the standard notation  $(\mathcal{H}_i)^\perp$  for the orthogonal complement in  $L^2(I)$  with the standard scalar product  $(f, g)_I = \int_I f(x) g(x) d\lambda(x)$ . We start with the first subspace.

**Lemma 3.2.** The relation  $\tilde{\mathcal{L}}_I = \mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1}$  holds on  $(\begin{smallmatrix} 0 \\ \mathcal{H}_1 \end{smallmatrix})^\perp$ .

The proof is a straightforward check. The notations are slightly simplified when the functions  $\{\epsilon_I \phi_k, \phi_k, k = 1, \dots, 2n\}$  are linearly independent in  $L^2(I)$ . The degenerate case is left to the reader. Consider  $f_s \in (\mathcal{H}_1)^\perp$ ,  $s = 1, \dots, 2n$  such that

$$(\overline{\epsilon \phi_k}, f_s)_I = (\overline{\epsilon \phi_k}, \phi_s)_I, \quad k = 1, \dots, 2n. \quad (39)$$

We are going to establish the relation for  $(\begin{smallmatrix} 0 \\ f_s \end{smallmatrix})$ , which then immediately extends by linearity to the linear combinations of  $(\begin{smallmatrix} 0 \\ f_s \end{smallmatrix})$ . We write

$$\begin{aligned} \mathcal{K}_I \begin{pmatrix} 0 \\ -f_s \end{pmatrix} &= \sum_{j, k=1, \dots, 2n} M_{jk}^{-t} (\overline{\epsilon \phi_k}, -f_s)_I \begin{pmatrix} -\epsilon \phi_j \\ \phi_j \end{pmatrix} - \begin{pmatrix} \epsilon_I f_s \\ 0 \end{pmatrix} \\ &= \sum_{j, k=1, \dots, 2n} M_{jk}^{-t} (\overline{\epsilon \phi_k}, -\phi_s)_I \begin{pmatrix} -\epsilon \phi_j \\ \phi_j \end{pmatrix} - \begin{pmatrix} \epsilon_I f_s \\ 0 \end{pmatrix} \\ &= \sum_{j=1, \dots, 2n} (G^I M_{sj}^{-1}) \begin{pmatrix} \epsilon \phi_j \\ -\phi_j \end{pmatrix} - \begin{pmatrix} \epsilon_I f_s \\ 0 \end{pmatrix} \end{aligned} \quad (40)$$

(we have used (39) in the second equality) and

$$\mathcal{K}_I \begin{pmatrix} \epsilon_I \phi_s \\ -f_s \end{pmatrix} = \sum_{j=1, \dots, 2n} ((G^I - (G^I)^t) M^{-1})_{sj} \begin{pmatrix} \epsilon \phi_j \\ -\phi_j \end{pmatrix} - \begin{pmatrix} \epsilon_I f_s \\ 0 \end{pmatrix}. \quad (41)$$

Combining (40) and (41) we get

$$\mathcal{K}_I \begin{pmatrix} -\epsilon_I \phi_s \\ -f_s \end{pmatrix} = \sum_{j=1, \dots, 2n} ((G^I + (G^I)^t) M^{-1})_{sj} \begin{pmatrix} \epsilon \phi_j \\ -\phi_j \end{pmatrix} - \begin{pmatrix} \epsilon_I f_s \\ 0 \end{pmatrix}. \quad (42)$$

Similarly to (23) we compute

$$\mathcal{K}_I \begin{pmatrix} \epsilon_I \phi_s \\ 0 \end{pmatrix} = \sum_{j=1, \dots, 2n} ((G^I - T) M^{-1})_{sj} \begin{pmatrix} \epsilon \phi_j \\ -\phi_j \end{pmatrix}. \quad (43)$$

It should be noted that  $\mathcal{K}_I(\epsilon_I f_s) = 0$  because  $\int_I (\epsilon_I \phi_s)(x) \phi_j(x) d\lambda(x) = -\int_I f_s(x)(\epsilon_I \phi_j)(x) d\lambda(x) = 0$  for all  $j = 1, \dots, 2n$ . This together with (39) allows us to conclude that the calculation of  $\mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1} \begin{pmatrix} 0 \\ f_s \end{pmatrix}$  is almost identical to the calculation of  $\mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1} \begin{pmatrix} 0 \\ \phi_s \end{pmatrix}$  with the only difference that in the former one we have to replace the term  $-(\epsilon_I \phi_s)$  by  $-(\epsilon_I f_s)$  (see the last equation of (44)). Namely

$$\begin{aligned} & \mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1} \begin{pmatrix} 0 \\ -f_s \end{pmatrix} \\ &= \mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1} \left( \frac{1}{2} \begin{pmatrix} \epsilon \phi_s \\ -f_s \end{pmatrix} + \begin{pmatrix} -\epsilon \phi_s \\ -f_s \end{pmatrix} \right) \\ &= \sum_{j=1, \dots, 2n} (1/2) ((A+B)(\text{Id} - A + C)^{-1})_{sj} \begin{pmatrix} \epsilon_I \phi_j \\ -\phi_j \end{pmatrix} \\ &\quad - \sum_{j=1, \dots, 2n} (1/2) ((A+B)(\text{Id} - A + C)^{-1})_{sj} \begin{pmatrix} \epsilon_I \phi_j \\ 0 \end{pmatrix} - \begin{pmatrix} \epsilon_I f_s \\ 0 \end{pmatrix} \\ &= \sum_{j=1, \dots, 2n} [G^I (M^{X \setminus I})^{-1}]_{sj} \left[ \begin{pmatrix} \epsilon \phi_j \\ -\phi_j \end{pmatrix} - \begin{pmatrix} \epsilon_I \phi_j \\ 0 \end{pmatrix} \right] - \begin{pmatrix} \epsilon_I f_s \\ 0 \end{pmatrix}. \end{aligned} \tag{44}$$

where  $A, B, C$  are defined in (27)–(29). At the same time

$$\begin{aligned} \tilde{\mathcal{L}}_I \begin{pmatrix} 0 \\ -f_s \end{pmatrix} &= \sum_{j,k=1, \dots, 2n} ((M^{X \setminus I})^{-1})_{jk} \begin{pmatrix} \epsilon_{X \setminus I} \phi_j \\ -\phi_j \end{pmatrix} (\overline{\epsilon_{X \setminus I} \phi_k}, f_s)_I - \begin{pmatrix} \epsilon_I f_s \\ 0 \end{pmatrix} \\ &= \sum_{j,k=1, \dots, 2n} ((M^{X \setminus I})^{-1})_{jk} \begin{pmatrix} \epsilon_{X \setminus I} \phi_j \\ -\phi_j \end{pmatrix} (\overline{\epsilon \phi_k}, f_s)_I - \begin{pmatrix} \epsilon_I f_s \\ 0 \end{pmatrix} \\ &= [G^I (M^{X \setminus I})^{-1}]_{sj} \left[ \begin{pmatrix} \epsilon \phi_j \\ -\phi_j \end{pmatrix} - \begin{pmatrix} \epsilon_I \phi_j \\ 0 \end{pmatrix} \right] - \begin{pmatrix} \epsilon_I f_s \\ 0 \end{pmatrix}. \end{aligned} \tag{45}$$

Therefore  $\tilde{\mathcal{L}}_I \begin{pmatrix} 0 \\ -f_s \end{pmatrix} = \mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1} \begin{pmatrix} 0 \\ -f_s \end{pmatrix}$ ,  $s = 1, \dots, 2n$ . By linearity result follows for all  $\begin{pmatrix} 0 \\ f \end{pmatrix}$  such that  $(\overline{\epsilon_I \phi_k}, f)_I = \int_I (\epsilon_I \phi_k)(x) f(x) d\lambda(x) = 0$ ,  $k, j = 1, \dots, 2n$ . Lemma 3.2 is proven.

To check (17) on  $\begin{pmatrix} \mathcal{K}_0^{\pm} \\ g \end{pmatrix}$  we note that  $\mathcal{K}_I(\text{Id} - \mathcal{K}_I)^{-1} \begin{pmatrix} g \\ 0 \end{pmatrix} = \tilde{\mathcal{L}}_I \begin{pmatrix} g \\ 0 \end{pmatrix} = 0$  for  $g$  such that  $\int_I g(x) \phi_k(x) d\lambda(x) = 0$ ,  $k = 1, \dots, 2n$ , which together with the invertibility of  $M$  finishes the proof. The theorem is proven.

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