Janossy Densities II. Pfaffian Ensembles. *

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Abstract

We extend the main result of the companion paper math-ph/0212063 to the case of the pfaffian ensembles.

1 Introduction and Formulation of Results

Let us consider a 2n-particle pfaffian ensemble introduced by Rains in [9]: Let (X, λ) 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

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

defines the density of a 2n-dimensional probability distribution on $X^{2n} = X \times \cdots \times X$ with respect to the product measure $\lambda^{\otimes 2n}$. Ensembles of this form were introduced in [9] and [11]. We recall (see e.g. [5]) that the pfaffian of a $2n \times 2n$ antisymmetric matrix $A = (a_{jk}), j, k = 1, \ldots, 2n, a_{jk} = -a_{kj}$, is defined as $pf(A) = \sum_{\tau} (-1)^{sign(\tau)} a_{i_1j_1} \times \cdots \times a_{i_n,j_n}$, where the summation is over all partitions of the set $\{1,\ldots,2m\}$ into disjoint pairs $\{i_1,j_1\},\ldots,\{i_n,j_n\}$ such that $i_k < j_k, \ k = 1,\ldots n$, and $sign(\tau)$ is the sign of the permutation (i_1,j_1,\ldots,i_n,j_n) . The normalization constant in (1) (usually called the partition function)

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

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

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

For the pfaffian ensemble (1) one can explicitly calculate k-point correlation functions $\rho_k(x_1,\ldots,x_k):=((2n)!/(2n-k)!)\int_{X^{2n-k}}p(x_1,\ldots,x_k,x_{k+1},\ldots,x_{2n})d\lambda(x_{k+1})\ldots d\lambda(x_{2n}),\ k=1,\ldots,2n$ and show that they have the pfaffian form ([9])

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

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

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

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 λ 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 $x_1, x_2, \ldots x_k$. In other words

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

On the other hand, if μ is supported by a discrete set of points, then

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

In general, random point processes with the k-point correlation functions of the pfaffian form (4) are called pfaffian random point processes. ([8]). Pfaffian point processes include determinantal point processes ([10]) 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 ϵ is an antisymmetric kernel.

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

$$\mathcal{J}_{k,I}(x_1,\ldots x_k)\lambda(dx_1)\cdots\lambda(dx_k) = \Pr\{ \text{ there are exactly } k \text{ particles in } I,$$

one in each of the k distinct infinitesimal intervals $(x_i,x_i+dx_i)\}$

If λ is discrete then

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

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

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

where

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

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

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

$$G_{jk}^{I} = \int_{I} \phi_{j}(x) \int_{X} \epsilon(x, y) \phi_{k}(y) \lambda(dy) \lambda(dx), \tag{8}$$

$$M_{jk}^{I} = \int_{I^{2}} \phi_{j}(x)\epsilon(x,y)\phi_{k}(y)\lambda(dx)\lambda(dy), \qquad (9)$$

$$M_{jk}^{X\backslash I} = \int_{(X\backslash I)^2} \phi_j(x) \epsilon(x, y) \phi_k(y) \lambda(dx) \lambda(dy)$$
 (10)

(please compare (9)-(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

$$L_{I}(x,y) = \left(\begin{array}{cc} \sum_{1 \leq j,k \leq 2n} \phi_{j}(x) (M^{X \setminus I})_{jk}^{-t} \phi_{k}(y) & \sum_{1 \leq j,k \leq 2n} \phi_{j}(x) (M^{X \setminus I})_{jk}^{-t} (\epsilon_{X \setminus I} \phi_{k})(y) \\ \sum_{1 \leq j,k \leq 2n} (\epsilon_{X \setminus I} \phi_{j})(x) (M^{X \setminus I})_{jk}^{-t} \phi_{k}(y) & -\epsilon(x,y) + \sum_{1 \leq j,k \leq 2n} (\epsilon_{X \setminus I} \phi_{j})(x) (M^{X \setminus I})_{jk}^{-t} (\epsilon_{X \setminus I} \phi_{k})(y) \end{array} \right)$$

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 [3]. 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 [9] (see also [11] and [12]).

Biorthogonal Ensembles.

Consider the particle space to be the union of two identical measure spaces (V,μ) and (W,μ) : $X = V \cup W$, V = W. The configuration of 2n particles in X will consist of n particles v_1, \ldots, v_n in V and n particles w_1, \ldots, w_n in W in such a way that the configurations of particles in V and W are identical (i.e. $v_j = w_j, j = 1, \ldots, n$). Let $\xi_j, \ \psi_j, j = 1, \ldots, n$ be some functions on V. We define $\{\phi_j\}$ and ϵ in (1) so that $\phi_j(v) = 0, \ v \in V, \ \phi_j(w) = \xi_j(w), \ w \in W, \ j = 1, \ldots, n, \ \phi_j(v) = \psi_{j-n-1}(v), \ v \in V, \ \phi_j(w) = 0, \ w \in W, \ j = n+1, \ldots, 2n, \ \text{and} \ \epsilon(v_1, v_2) = 0, \ v_1, v_2 \in V, \ \epsilon(w_1, w_2) = 0, \ w_1, w_2 \in W, \ \epsilon(v, w) = -\epsilon(w, v) = \delta_{vw}, \ v \in V, \ w \in W.$ The restriction of the measure λ on both V and W is defined to be equal to μ . Then (1) specializes into (see Corollary 1.5. in [9])

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

Ensembles of the form (12) are known as biorthogonal ensembles (see [7], [1]). The statement of the Theorem 1.1 in the case (12) has been proven in the companion paper [3]. The special case of the biorthogonal ensemble (12) when $V = \mathbb{R}$, $\xi_j(x) = \psi_j(x) = x^{j-1}$, and $V = \{\mathbb{C} | |z| = 1\}$, $\xi_j(z) = \overline{\psi}_j(z) = z^{j-1}$, such ensembles are well known in Random Matrix Theory as unitary ensembles, see [6] for details. An ensemble of the form (12) which is different from random matrix ensembles was studied in [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, ..., 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

$$p(v_1, ..., v_n) = const_n \prod_{1 \le i \le j \le n} (v_i - v_j)^2 \prod_{1 \le j \le n} \omega(v_j).$$
(13)

Polynomial $(\beta = 1)$ Ensembles.

Let $X = \mathbb{R}$ or \mathbb{Z} , $\phi_j(x) = x^{j-1}$, j = 1, ..., 2n, $\epsilon(x, y) = \frac{1}{2}sgn(y - x)$ 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 [9], Remark 1):

$$p(x_1, \dots, x_{2n}) = const_n \prod_{1 \le i < j \le 2n} |x_i - x_j| \prod_{1 \le j \le 2n} \omega(x_j).$$

$$(14)$$

In Random Matrix Theory the ensembles (14) in the continuous case are known as *orthogonal ensembles*, see [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 $(Y,\mu),(Z,\mu),\ X=Y\cup Z,\ Y=Z,$ where $Y=\mathbb{R}$ or $Y=\mathbb{Z}$. The configuration of 2n particles $x_1,\ldots,x_{2n},$ in X will consist of n particles y_1,\ldots,y_n in Y and n particles $z_1,\ldots,z_n,$ in Z in such a way that the configurations of particles in Y and Z are identical. We define $\{\phi_j\}$ and ϵ so that $\phi_j(y)=y^j,\ \in Y,\ \phi_j(z)=jz^{j-1},\ z\in Z,\ \epsilon(y_1,y_2)=0,\ \epsilon(z_1,z_2)=0,\ \epsilon(y,z)=-\epsilon(z,y)=\delta_{yz}.$ As above we assume that the measure μ has a density ω 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 [9]))

$$p(y_1, \dots, y_n) = const_n \prod_{1 \le i \le j \le n} (y_i - y_j)^4 \prod_{1 \le j \le n} \omega(y_j).$$
 (15)

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

3 Proof of the Main Result

Consider matrix kernels

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

The the relation (7) simplifies into

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

which is the same relation that is satisfied by the correlation and Janossy scalar kernels in the determinantal case ([4], [2]). The consideration of \mathcal{K}_I and \mathcal{L}_I is motivated by the fact that the pfaffians of the $2k \times 2k$ matrices with the antisymmetric matrix kernels K_I and L_I are equal to the quaternion determinants ([6]) 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 2×2 block with complex entries). It follows from (5) and (16) that the kernel \mathcal{K}_I is given by the formula

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

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

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

As above, $\epsilon \phi$ stands for $\int_X \epsilon(x,y)\phi(y)$. We use the notation $\phi_j \otimes \phi_k$ is a shorthand for $\phi_j(x)\phi_k(y)$. To prove the main result of the paper we will show that $\widetilde{\mathcal{L}}_I = \mathcal{K}_I(Id - \mathcal{K}_I)^{-1}$ (in other words we are going to prove that $\widetilde{\mathcal{L}}_I = \mathcal{L}_I$, 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 $\mathcal{H} = Span\left\{\begin{pmatrix} \epsilon \phi_s \\ -\phi_s \end{pmatrix}, \begin{pmatrix} -\epsilon \phi_s \\ -\phi_s \end{pmatrix}, \begin{pmatrix} \epsilon_I \phi_s \\ 0 \end{pmatrix}\right\}_{s=1,\dots,2n}$ is invariant under \mathcal{K}_I and $\widetilde{\mathcal{L}}_I$. The main part of the proof of the theorem is to show that $\widetilde{\mathcal{L}}_I = \mathcal{K}_I(Id - \mathcal{K}_I)^{-1}$ holds on \mathcal{H} .

Lemma 1 The operators K_I , $\widetilde{\mathcal{L}}_I$ leave \mathcal{H} invariant and $\widetilde{\mathcal{L}}_I = K_I(Id - K_I)^{-1}$ holds on \mathcal{H} .

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

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

$$T_{sk} = \int_{I} \phi_s(x) \int_{X \setminus I} \epsilon(x, y) \phi_k(y) d\lambda(y) d\lambda(x)$$
 (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 the equations (20)-(21) as

$$\mathcal{K}_{I} \begin{pmatrix} \epsilon \phi_{s} \\ -\phi_{s} \end{pmatrix} = \sum_{j=1,\dots,2n} ((G^{I} - (G^{I})^{t})M^{-1})_{sj}^{t} \begin{pmatrix} \epsilon \phi_{j} \\ -\phi_{j} \end{pmatrix} - \begin{pmatrix} \epsilon_{I}\phi_{s} \\ 0 \end{pmatrix}, \tag{24}$$

$$\mathcal{K}_{I} \begin{pmatrix} -\epsilon \phi_{s} \\ -\phi_{s} \end{pmatrix} = \sum_{j=1,\dots,2n} ((G^{I} + (G^{I})^{t})M^{-1})_{sj} \begin{pmatrix} \epsilon \phi_{j} \\ -\phi_{j} \end{pmatrix} - \begin{pmatrix} \epsilon_{I}\phi_{s} \\ 0 \end{pmatrix}$$
(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 $\left\{ \begin{pmatrix} \epsilon \phi_s \\ -\phi_s \end{pmatrix}, \begin{pmatrix} -\epsilon \phi_s \\ -\phi_s \end{pmatrix}, \begin{pmatrix} \epsilon_I \phi_s \\ 0 \end{pmatrix} \right\}_{s=1,\dots 2n}$

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

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

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

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

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

When a matrix has a block form $\mathcal{M} = \begin{pmatrix} A & B & C \\ 0 & 0 & 0 \\ -Id & -Id & 0 \end{pmatrix}$ (as it is in our case) the matrix $\mathcal{M} \times$

 $(Id - \mathcal{M})^{-1}$ has the block form

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

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

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

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

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

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

$$\widetilde{\mathcal{L}}_{I} = \sum_{\substack{i,k=1,\dots,2n\\j,k=1}} (M_{jk}^{X\backslash I})^{t} \begin{pmatrix} -(\epsilon_{X\backslash I}\phi_{j}) \otimes \phi_{k} & -(\epsilon_{X\backslash I}\phi_{j}) \otimes (\epsilon_{X\backslash I}\phi_{k})\\ \phi_{k} \otimes \phi_{k} & \phi_{j} \otimes (\epsilon_{X\backslash I}\phi_{k}) \end{pmatrix} + \begin{pmatrix} 0 & \epsilon_{X\backslash I}\\ 0 & 0 \end{pmatrix}.$$
(34)

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

$$\widetilde{\mathcal{L}}_{I}\begin{pmatrix} \epsilon \phi_{s} \\ -\phi_{s} \end{pmatrix} = \sum_{j=1,\dots,2n} ((T - (G^{I})^{t})(M^{X \setminus I})^{-1})_{sj} \begin{pmatrix} \epsilon \phi_{j} \\ -\phi_{j} \end{pmatrix} - \sum_{1 \leq j \leq 2n} ((T - (G^{I})^{t})(M^{X \setminus I})^{-1})_{sj} \begin{pmatrix} \epsilon_{I} \phi_{j} \\ 0 \end{pmatrix} - \begin{pmatrix} \epsilon_{I} \phi_{s} \\ 0 \end{pmatrix},$$
(35)

$$\widetilde{\mathcal{L}}_{I}\begin{pmatrix} -\epsilon\phi_{s} \\ -\phi_{s} \end{pmatrix} = \sum_{j=1,\dots,2n} ((T + (G^{I})^{t})(M^{X\backslash I})^{-1})_{sj} \begin{pmatrix} \epsilon\phi_{j} \\ -\phi_{j} \end{pmatrix} - \sum_{1\leq j\leq 2n} ((T + (G^{I})^{t})(M^{X\backslash I})^{-1})_{sj} \begin{pmatrix} \epsilon_{I}\phi_{j} \\ 0 \end{pmatrix} - \begin{pmatrix} \epsilon_{I}\phi_{s} \\ 0 \end{pmatrix}, \tag{36}$$

and

$$\widetilde{\mathcal{L}}_{I}\left(\begin{array}{c} \epsilon_{I}\phi_{s} \\ 0 \end{array}\right) = \sum_{j=1,\dots,2n} ((G^{I} - T)(M^{X\backslash I})^{-1})_{sj} \left(\begin{array}{c} \epsilon\phi_{j} \\ -\phi_{j} \end{array}\right) - \sum_{j=1,\dots,2n} ((G^{I} - T)(M^{X\backslash I})^{-1})_{sj} \left(\begin{array}{c} \epsilon_{I}\phi_{j} \\ 0 \end{array}\right). \tag{37}$$

Therefore the restriction of $\widetilde{\mathcal{L}}_I$ to \mathcal{H} in the basis $\left\{ \begin{pmatrix} \epsilon \phi_s \\ -\phi_s \end{pmatrix}, \begin{pmatrix} -\epsilon \phi_s \\ -\phi_s \end{pmatrix}, \begin{pmatrix} \epsilon_I \phi_s \\ 0 \end{pmatrix} \right\}_{s=1,\dots,2n}$ the following block structure

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

Comparing (30), (31)-(33) and (38) we see that $\widetilde{\mathcal{L}}_I = \mathcal{K}_I (Id - \mathcal{K}_I)^{-1}$ on \mathcal{H} . Lemma is proven. To show that $\widetilde{\mathcal{L}}_I = \mathcal{K}_I (Id - \mathcal{K}_I)^{-1}$ also holds on the complement of \mathcal{H} it is enough to prove it on the subspaces $\begin{pmatrix} (\mathcal{H}_1)^{\perp} \\ 0 \end{pmatrix}$, and $\begin{pmatrix} 0 \\ (\mathcal{H}_2)^{\perp} \end{pmatrix}$, where $\mathcal{H}_1 = Span(\overline{\epsilon_I \phi_s})_{k=1,\dots,2n}$ and $\mathcal{H}_2 = Span(\overline{\epsilon_I \phi_s})_{k=1,\dots,2n}$ $Span(\overline{\phi_s})_{k=1,...,2n}$. The inveribility of the matrix M_I implies that actually it is enough to prove $\widetilde{\mathcal{L}}_I = \mathcal{K}_I (Id - \mathcal{K}_I)^{-1}$ on the subspaces $\begin{pmatrix} (\mathcal{H}_2)^{\perp} \\ 0 \end{pmatrix}$, and $\begin{pmatrix} 0 \\ (\mathcal{H}_1)^{\perp} \end{pmatrix}$. Here we use the standard notation $(\mathcal{H}_i)^{\perp}$ for the orthogonal complement in $L^2(I)$ with the standard scalar product $(f,g)_I = \int_I \times$ $\overline{f(x)}g(x)d\lambda(x)$. We start with the first subspace.

Lemma 2 The relation
$$\widetilde{\mathcal{L}}_I = \mathcal{K}_I (Id - \mathcal{K}_I)^{-1}$$
 holds on $\begin{pmatrix} 0 \\ (\mathcal{H}_1)^{\perp} \end{pmatrix}$.

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

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

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

$$\mathcal{K}_{I}\begin{pmatrix} 0 \\ -f_{s} \end{pmatrix} = \sum_{j,k=1,\dots,2n} M_{jk}^{-t} (\overline{\epsilon\phi_{k}}, -f_{s})_{I} \begin{pmatrix} -\epsilon\phi_{j} \\ \phi_{j} \end{pmatrix} - \begin{pmatrix} \epsilon_{I}f_{s} \\ 0 \end{pmatrix} \\
= \sum_{j,k=1,\dots,2n} M_{jk}^{-t} (\overline{\epsilon\phi_{k}}, -\phi_{s})_{I} \begin{pmatrix} -\epsilon\phi_{j} \\ \phi_{j} \end{pmatrix} - \begin{pmatrix} \epsilon_{I}f_{s} \\ 0 \end{pmatrix} \\
= \sum_{j=1,\dots,2n} (G^{I}M_{sj}^{-1}) \begin{pmatrix} \epsilon\phi_{j} \\ -\phi_{j} \end{pmatrix} - \begin{pmatrix} \epsilon_{I}f_{s} \\ 0 \end{pmatrix} \tag{40}$$

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

$$\mathcal{K}_{I} \begin{pmatrix} \epsilon_{I} \phi_{s} \\ -f_{s} \end{pmatrix} = \sum_{j=1,\dots,2n} ((G^{I} - (G^{I})^{t}) M^{-1})_{sj} \begin{pmatrix} \epsilon \phi_{j} \\ -\phi_{j} \end{pmatrix} - \begin{pmatrix} \epsilon_{I} f_{s} \\ 0 \end{pmatrix}. \tag{41}$$

Combining (40) and (41) we get

$$\mathcal{K}_{I}\left(\begin{array}{c} -\epsilon_{I}\phi_{s} \\ -f_{s} \end{array}\right) = \sum_{j=1,\dots,2n} ((G^{I} + (G^{I})^{t})M^{-1})_{sj} \left(\begin{array}{c} \epsilon\phi_{j} \\ -\phi_{j} \end{array}\right) - \left(\begin{array}{c} \epsilon_{I}f_{s} \\ 0 \end{array}\right), \tag{42}$$

Similarly to (23) we compute

$$\mathcal{K}_{I} \begin{pmatrix} \epsilon_{I} \phi_{s} \\ 0 \end{pmatrix} = \sum_{j=1,\dots,2n} ((G^{I} - T)M^{-1})_{sj} \begin{pmatrix} \epsilon \phi_{j} \\ -\phi_{j} \end{pmatrix}$$
(43)

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

$$\mathcal{K}_{I}(Id - \mathcal{K}_{I})^{-1} \begin{pmatrix} 0 \\ -f_{s} \end{pmatrix} = \mathcal{K}_{I}(Id - \mathcal{K}_{I})^{-1} \begin{pmatrix} \frac{1}{2} \begin{pmatrix} \epsilon \phi_{s} \\ -f_{s} \end{pmatrix} + \begin{pmatrix} -\epsilon \phi_{s} \\ -f_{s} \end{pmatrix} \end{pmatrix}$$

$$= \sum_{j=1,\dots,2n} (1/2) \left((A+B)(Id - A+C)^{-1} \right)_{sj} \begin{pmatrix} \epsilon_{I}\phi_{j} \\ -\phi_{j} \end{pmatrix}$$

$$- \sum_{j=1,\dots,2n} (1/2) \left((A+B)(Id - A+C)^{-1} \right)_{sj} \begin{pmatrix} \epsilon_{I}\phi_{j} \\ 0 \end{pmatrix} - \begin{pmatrix} \epsilon_{I}f_{s} \\ 0 \end{pmatrix}$$

$$= \sum_{j=1,\dots,2n} \left[G^{I}(M^{X\setminus I})^{-1} \right]_{sj} \left[\begin{pmatrix} \epsilon \phi_{j} \\ -\phi_{j} \end{pmatrix} - \begin{pmatrix} \epsilon_{I}\phi_{j} \\ 0 \end{pmatrix} \right] - \begin{pmatrix} \epsilon_{I}f_{s} \\ 0 \end{pmatrix} (44)$$

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

$$\widetilde{\mathcal{L}}_{I}\begin{pmatrix} 0 \\ -f_{s} \end{pmatrix} = \sum_{j,k=1,\dots,2n} (M^{X\backslash I})^{-t})_{jk} \begin{pmatrix} \epsilon_{X\backslash I}\phi_{j} \\ -\phi_{j} \end{pmatrix} (\overline{\epsilon_{X\backslash I}\phi_{k}}, f_{s})_{I} - \begin{pmatrix} \epsilon_{I}f_{s} \\ 0 \end{pmatrix} \\
= \sum_{j,k=1,\dots,2n} (M^{X\backslash I})^{-t})_{jk} \begin{pmatrix} \epsilon_{X\backslash I}\phi_{j} \\ -\phi_{j} \end{pmatrix} (\overline{\epsilon\phi_{k}}, f_{s})_{I} - \begin{pmatrix} \epsilon_{I}f_{s} \\ 0 \end{pmatrix} \\
= \left[G^{I}(M^{X\backslash I})^{-1} \right]_{sj} \left[\begin{pmatrix} \epsilon\phi_{j} \\ -\phi_{j} \end{pmatrix} - \begin{pmatrix} \epsilon_{I}\phi_{j} \\ 0 \end{pmatrix} \right] - \begin{pmatrix} \epsilon_{I}f_{s} \\ 0 \end{pmatrix}. \tag{45}$$

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

To check (17) on $\binom{(\mathcal{H}_2)^{\perp}}{0}$ we note that $\mathcal{K}_I(Id-\mathcal{K}_I)^{-1}\binom{g}{0}=\widetilde{\mathcal{L}}_I\binom{g}{0}=0$ for g such that $\int_I g(x)\phi_k(x)d\lambda(x)=0, \quad k=1,\ldots,2n$, which together with the invertibility of M finishes the proof. Theorem is proven.

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