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ON FREDHOLM INTEGRAL EQUATIONS WITH RANDOM  
DEGENERATE KERNELS<sup>(1)</sup>

1. In this paper we consider the random operator equation

$$(1) \quad (T(\omega) - \lambda I)f = g$$

when the random operator  $T(\omega)$  is a Fredholm integral operator with random kernel of degenerate type. The results presented compliment and extend those announced in [4]. In Sec. 2 we show that a Fredholm integral equation with random degenerate kernel can be reduced to an algebraic system of random linear equations; and consider the question of the existence, uniqueness and measurability of the solution of the algebraic system of random linear equations. In Sec. 3 we investigate the asymptotic distribution of the eigenvalues of the random Fredholm operator.

2. Consider the Fredholm integral equation of second kind

$$(2) \quad \int_0^1 K(x, y)f(y)dy - \lambda f(x) = g(x).$$

It is well-known (cf. Pogorzelski [6], Tricomi [7]) that if the kernel  $K(x, y)$  is *degenerate*<sup>(2)</sup>, that is  $K(x, y)$  is of the form

$$(3) \quad K(x, y) = \sum_{i=1}^n a_i(x)\beta_i(y)$$

where  $\{a_i(x)\}_{i=1}^n$  and  $\{\beta_i(y)\}_{i=1}^n$  are two sets of linearly independent  $L_2(0, 1)$ -functions, then the integral equation can be reduced to an algebraic system of  $n$  linear equations in  $n$  unknowns. If we put

$$(4) \quad \int_0^1 \beta_j(x)f(x)dx = \xi_j, \quad j = 1, 2, \dots, n,$$

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<sup>(2)</sup> Degenerate kernels are also referred to as *separable*, kernels of *finite-rank*, or *Pincherle-Goursat* kernels.

eq. (2) with kernel (3) becomes

$$(5) \quad \sum_{j=1}^n \xi_j \alpha_j(x) - \lambda f(x) = g(x).$$

If we now multiply (5) by  $\beta_i(x)$  ( $i = 1, 2, \dots, n$ ) and then integrate, we obtain

$$\sum_{j=1}^n \xi_j \int_0^1 \alpha_j(x) \beta_i(x) dx - \lambda \xi_i = \int_0^1 \beta_i(x) g(x) dx;$$

that is

$$(6) \quad \sum_{j=1}^n a_{ij} \xi_j - \lambda \xi_i = b_i, \quad i = 1, 2, \dots, n,$$

where

$$(7) \quad a_{ij} = \int_0^1 \alpha_j(x) \beta_i(x) dx,$$

$$(8) \quad b_i = \int_0^1 \beta_i(x) g(x) dx.$$

Rewriting (6) in matrix form, we have

$$(9) \quad (A - \lambda I) \xi = b,$$

where  $A = (a_{ij})$  is an  $n \times n$  matrix, and  $\xi$  and  $b$  are  $n$ -vectors. Hence we see that a Fredholm integral equation with degenerate kernel reduces to an algebraic system of linear equations; and, moreover, the eigenvalues of a degenerate kernel are the roots of the algebraic equation  $|A - \lambda I| = 0$ .

Let  $(\Omega, \mathcal{A}, \mu)$  be a probability measure space; and consider the *random kernel*

$$(10) \quad K(\omega, x, y) = \sum_{i=1}^n \alpha_i(\omega, x) \beta_i(y).$$

In (10)  $\{\alpha_i(\omega, x)\}_{i=1}^n$  is a family of (almost surely) independent  $L_2(0, 1)$ -random functions and  $\{\beta_i(y)\}_{i=1}^n$  is a set of independent  $L_2(0, 1)$ -determinate functions. Clearly, for every fixed  $x, y \in (0, 1)$  the kernel  $K$  is a measurable function of  $\omega$ . Put

$$(11) \quad \xi_i = \int_0^1 \beta_i(x) f(x) dx, \quad i = 1, 2, \dots, n.$$

Then, proceeding as in the deterministic case, we obtain the algebraic system of random linear equations

$$(12) \quad \sum_{j=1}^n a_{ij}(\omega) \xi_j - \lambda \xi_i = b_i, \quad i = 1, 2, \dots, n,$$

where

$$(13) \quad a_{ij}(\omega) = \int_0^1 a_j(\omega, x) \beta_i(x) dx, \quad i, j = 1, 2, \dots, n,$$

$$(14) \quad b_i = \int_0^1 \beta_i(x) g(x) dx, \quad i = 1, 2, \dots, n.$$

The integrals in (11) and (14) are well-defined; and the Riemann-integrability of

$$(15) \quad \int_0^1 \int_0^1 \beta_i(x_1) \beta_i(x_2) R_j(x_1, x_2) dx_1 dx_2,$$

where  $R_j(x_1, x_2) = \mathcal{E}\{a_j(\omega, x_1)a_j(\omega, x_2)\}$  is the covariance kernel associated with the  $a_j(\omega, x)$ -process, is sufficient to insure that the integral in (13) exists in mean square and defines, for every pair  $(i, j)$ , a real-valued random variable.

Equation (12) can be rewritten as the random operator equation

$$(16) \quad (A(\omega) - \lambda I)\xi = b,$$

where  $A(\omega)$  is a  $n \times n$  random matrix with elements  $a_{ij}(\omega)$  defined by (13), and  $\xi$  and  $b$  are  $n$ -vectors. We remark that (16) can be interpreted as a random operator equation in the Euclidean space  $R_n$  or the Hilbert space  $l_2(n)$ .

In [4] the existence, uniqueness and measurability of the solution  $\xi(\omega)$  of (16) was established using the Špaček-Hanš probabilistic analogue of the Banach contraction mapping theorem. We now consider the solution of (16) using the following result of Bharucha-Reid [2], [3] and Hanš [5] on the invertibility of linear random operators of the form  $T(\omega) - \lambda I$ : Let  $T(\omega)$  be a random transformation on a separable Banach space  $\mathcal{X}$  which is for every  $\omega \in \Omega$  linear and bounded. Then, for every real number  $\lambda \neq 0$  the set  $\Omega(\lambda) = \{\omega: \|T(\omega)\| < |\lambda|\} \in \mathcal{A}$ , and the random transformation  $T(\omega) - \lambda I$  is invertible for every  $\omega \in \Omega(\lambda)$ . Furthermore, for every  $\omega \in \Omega(\lambda)$  the solution of the random operator equation  $(T(\omega) - \lambda I)f = g$  is for every random variable  $g(\omega)$  with values in  $\mathcal{X}$  given by  $f(\omega) = (T(\omega) - \lambda I)^{-1}g(\omega)$ , and  $f(\omega)$  is measurable with respect to the  $\sigma$ -algebra  $\Omega(\lambda) \cap \mathcal{A}$ .

A straightforward application of the above theorem enables us to state the following result for the solution of (16): Let  $\lambda \neq 0$  be a real number such that

$$\mu(\Omega(\lambda)) = \mu\left(\left\{\omega: \left(\sum_{i,j=1}^n a_{ij}^2(\omega)\right)^{1/2} < |\lambda|\right\}\right) = 1.$$

Then the random matrix  $(A(\omega) - \lambda I)$  is invertible; and the solution  $\xi(\omega) = (A(\omega) - \lambda I)^{-1}b$  of (16) is  $(\Omega(\lambda) \cap \mathcal{A})$ -measurable.

**3.** In this section we utilize the fact that for the eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_n$  of any  $n \times n$  matrix  $A = (a_{ij})$

$$\sum_{i=1}^n \lambda_i^k = \text{Trace}(A^k) = \sum_{i_1=1}^n \dots \sum_{i_k=1}^n a_{i_1 i_2} a_{i_2 i_3} \dots a_{i_k i_1}$$

for  $k = 1, 2, \dots$

Consider the random kernel

$$(17) \quad K(\omega, x, y) = \sum_{i=1}^n \alpha_i(\omega, x) \alpha_i(\omega, y),$$

where the  $\alpha_i$ 's are independent random functions having the same finite-dimensional probability distributions. For the sake of simplicity, we assume

$$(i) \quad m_k(x) = \mathcal{E}\{|a_i(\omega, x)|^k\} < \infty \quad \text{for all } x \in [0, 1],$$

and

$$(ii) \quad \int_0^1 m_k(x) dx < \infty \quad \text{for every } k = 1, 2, \dots$$

(Stronger results under weaker conditions can be obtained by the technique used in [1].) Put

$$(18) \quad R(x, y) = \mathcal{E}\{\alpha_i(\omega, x) \alpha_i(\omega, y)\},$$

and assume

$$(19) \quad \mathcal{E}\{\alpha_i(\omega, x)\} \equiv 0.$$

In the case we are considering, the kernel (17) is almost surely symmetric and positive-definite, since

$$\int_0^1 \int_0^1 K(\omega, x, y) h(x) \overline{h(y)} dx dy = \sum_{i=1}^n \left| \int_0^1 \alpha_i(\omega, x) h(x) dx \right|^2$$

for any continuous function  $h(x)$ . Therefore, the eigenvalues of (17) are all real and, moreover, non-negative.

The expectation of  $K$ ,

$$(20) \quad \mathcal{E}\{K(\omega, x, y)\} = nR(x, y),$$

is also a symmetric and positive-definite kernel, but in general *not* degenerate.

Consider the corresponding random matrix  $A(\omega)$  with elements

$$(21) \quad a_{ij}(\omega) = \int_0^1 \alpha_i(\omega, x) \alpha_j(\omega, x) dx.$$

Since  $a_{ij}(\omega) = a_{ji}(\omega)$  almost surely,  $A(\omega)$  is symmetric. The diagonal elements  $a_{ii}(\omega)$  are independent and have the same distribution with

$$(22) \quad \mathcal{E}\{a_{ii}(\omega)\} = \int_0^1 R(x, x) dx = \text{Trace}(R).$$

The off-diagonal elements also have the same distribution with

$$(23) \quad \mathcal{E}\{a_{ij}(\omega)\} = 0$$

and

$$(24) \quad \mathcal{E}\{a_{ij}^2(\omega)\} = \int_0^1 \int_0^1 R^2(x, y) dx dy.$$

The only difference between the problem investigated in [1], [8] and [9] and the problem here considered is that in addition to the symmetry condition there are other relations among the matrix elements; hence they are not independent. This difference completely changes the result. However, the elements  $a_{ij}(\omega)$  are independent *except* if they are in the same row or column of  $A(\omega)$ .

Define  $R_1 = R$ , and the  $(k-1)$ -fold iterated kernel  $R_k$  by

$$(25) \quad R_k(x, y) = \int_0^1 R_{k-1}(x, s) R(s, y) ds;$$

and

$$(26) \quad \text{Trace}(R_k) = \int_0^1 R_k(x, x) dx.$$

Then, if the indices  $i_1, i_2, \dots, i_k$  are all different,

$$(27) \quad \mathcal{E}\{a_{i_1 i_2}(\omega) a_{i_2 i_3}(\omega) \dots a_{i_k i_1}(\omega)\} \\ = \underbrace{\int_0^1 \dots \int_0^1}_{k\text{-fold}} R(x_1, x_2) R(x_2, x_3) \dots R(x_k, x_1) dx_1 dx_2 \dots dx_k = \text{Trace}(R_k).$$

Consequently, for all  $k = 1, 2, \dots$ ,

$$(28) \quad \lim_{n \rightarrow \infty} n^{-k} \sum_{i=1}^n \mathcal{E}\{\lambda_i^k(\omega)\} = \text{Trace}(R_k).$$

Let  $N_n(\omega, x)$  denote the number of eigenvalues of the random matrix  $A(\omega)$  which are less than  $x$ . Since  $A(\omega)$  is positive, we have  $N_n(\omega, 0) = 0$  and  $N_n(\omega, \infty) = n$  almost surely. Furthermore,

$$(29) \quad \int_0^1 x^k dN_n(\omega, x) = \sum_{i=1}^n \lambda_i^k(\omega).$$

Relation (28) can now be rewritten as

$$(30) \quad \lim_{n \rightarrow \infty} \int_0^1 x^k d\mathcal{E}\{N_n(\omega, nx)\} = \text{Trace}(R_k).$$

The convergence must be due to the fact that there are, on the average, very many small eigenvalues — but only a few large eigenvalues.

As an example, put  $\alpha_i(\omega, x) = X_i(\omega)$ . Then  $R(x, y) = \mathcal{E}\{X_i^2(\omega)\} = \sigma^2$ , and the eigenvalues of (17) are  $\lambda_1(\omega) = \lambda_2(\omega) = \dots = \lambda_{n-1}(\omega) = 0$  and  $\lambda_n(\omega) = \sum_{i=1}^n X_i^2(\omega)$ . From the strong law of large numbers, it follows that

$$\int_0^1 x^k dN_n(\omega, nx) = \left( \frac{1}{n} \sum_{i=1}^n X_i^2(\omega) \right)^k \rightarrow \text{Trace}(R_k)$$

almost surely, for every  $k = 1, 2, \dots$

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