

EQUIVALENCE AND PERPENDICULARITY OF GAUSSIAN PROCESSES

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1. Introduction. In [6] S. Kakutani showed that if one has equivalent probability measures μ_i and ν_i on the σ -field \mathcal{S}_i of subsets of a set Ω_i , $i = 1, 2, \dots$, and if μ and ν denote respectively the infinite product measures $\bigotimes_{i=1}^{\infty} \mu_i$ and $\bigotimes_{i=1}^{\infty} \nu_i$ on the infinite product σ -ring generated on the infinite product set Ω , then μ and ν are either equivalent or perpendicular, and he obtained necessary and sufficient conditions for equivalence to occur. The theorem here shown may be regarded as a generalization of a case of the Kakutani theorem.

Similar dichotomies have revealed themselves in the study of Gaussian stochastic processes. C. Cameron and W. T. Martin proved in [2] that if one considers the measures induced on path space by a Wiener process on the unit interval, then if the variances of the processes are different the measures are perpendicular. This sort of result was generalized by U. Grenander, starting from the viewpoint of statistical estimation, and utilizing a Karhunen representation for the processes involved. A wider sufficient condition for perpendicularity of the measures induced on path space by continuous Gaussian processes on the unit interval was obtained by G. Baxter in [1]. Cameron and Martin also examined the effect on the induced measure of taking certain types of affine transformations of a Wiener process (see [3], [4]). I. E. Segal extended their results in [8], and made the situation more transparent by use of his notion of "weak distributions", and in a large class of cases got conditions for equivalence.

In the present note it is shown that the equivalence-or-perpendicularity dichotomy holds in general for pairs of measures induced by Gaussian stochastic processes, and Segal's necessary and sufficient conditions for equivalence are extended to cover the case of nonzero mean. It has been pointed out to the author by C. Stein that one could also give a proof, in the case of zero mean, by use of the techniques of statistical testing of hypotheses.

2. Several lemmas. All Hilbert spaces mentioned will be over the reals.

Definition 1. An operator T from Hilbert space \mathbf{H} to Hilbert space \mathbf{K} will be called an *equivalence* operator if

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- (1) T is one-to-one onto, bounded, and has a bounded inverse.
- (2) $\sqrt{T^*T} = I + H$, where H is Hilbert-Schmidt.

LEMMA 1. *If A is a self-adjoint bounded invertible operator on \mathbf{H} then the following statements are equivalent:*

- (a) $A - I$ is Hilbert-Schmidt;
- (b) $(A - I)^2$ is Hilbert-Schmidt;
- (c) $A^{-1} - I$ is Hilbert-Schmidt.

If A, B satisfy (a), then so does ABA .

Proof. The first part is clear from consideration of the eigenvalues of the operators. For the second part: write $A = I + K, B = I + H$. Then $ABA = (I + K)^2 + (I + K)H(I + K)$, and since the sum and product of two Hilbert-Schmidt operators is Hilbert-Schmidt, $ABA - I$ is Hilbert-Schmidt.

DEFINITION 1. An operator T from Hilbert space \mathbf{H} to Hilbert space \mathbf{K} will be called an *equivalence operator* if

- (1) T is one-to-one onto, bounded, and has a bounded inverse;
- (2) $\sqrt{T^*T} - I$ is Hilbert-Schmidt.

LEMMA 2. *Products, conjugates, and inverses of equivalence operators are again equivalence operators.*

Proof. That they are one-to-one onto, and bounded, is clear (in the case of the conjugate operator, use the fact that the nullspace of T^* is the orthogonal complement of the range of T).

Let T be an equivalence operator from \mathbf{H} to \mathbf{K} . Let $Q = \sqrt{T^*T}$. Then $V = TQ^{-1}$ is an isometry from \mathbf{H} onto \mathbf{K} , and $T = VQ$. Thus $T^{-1} = Q^{-1}V^*$, and $(T^{-1})^*(T^{-1}) = VQ^{-2}V^*$. Since Q is the type of operator occurring in Lemma 1, and $(T^{-1})^*T^{-1}$ is a unitary transform of Q^{-2} , we get the result. Similarly, $(T^*)^*T^* = TT^* = VQ^2V^*$. Finally, let S be an equivalence operator from \mathbf{K} to \mathbf{L} , and let $P = \sqrt{S^*S}$, $U = SP^{-1}$. Then

$$(ST)^*(ST) = (VQUP)^*(VQUP) = PU^*QV^*VQUP = PU^*QUP,$$

and again Lemma 1 tells us that $\sqrt{(ST)^*(ST)}$ is of the desired form.

DEFINITION 2. A function x on a measure space with measure μ of total mass 1 is called *Gaussian* if either

- (1) x is almost everywhere a constant, γ .

or

- (2) there are numbers $\sigma > 0$ and γ (depending on x) such that

$$\mu\{\omega \mid x(\omega) \leq \lambda\} = \frac{1}{\sqrt{2\pi\sigma}} \int_{-\infty}^{\lambda} \exp\left\{\frac{1}{2}\left(\frac{t-\gamma}{\sigma}\right)^2\right\} dt .$$

Case 1 may be thought of as Case 2 with $\sigma = 0$. Then in either case we have

$$\int x d\mu = \gamma, \int (x - \gamma)^2 d\mu = \sigma^2$$

(the “ mean ” and “ variance ” of x).

LEMMA 3. Let $\mathcal{S}_1 \subset \mathcal{S}_2 \subset \dots$ be σ -fields of subsets of Ω , \mathcal{S} the smallest σ -field containing their union. Let μ, ν be probability measures in \mathcal{S} such that $\mu_i = \mu \mid \mathcal{S}_i$ is equivalent to $\nu_i = \nu \mid \mathcal{S}_i$. Let A, A_μ, A_ν be sets in \mathcal{S} forming a Hahn decomposition of Ω ; that is, μ is equivalent to ν when both are cut down to subsets of A , and $\mu(A_\nu) = \nu(A_\mu) = 0$. Then $d\mu_i/d\nu_i$ converges almost everywhere with respect to $\mu + \nu$ to $d\mu/d\nu$, if one makes the convention $d\mu/d\nu = 0$ on A_ν and $+\infty$ on A_μ .

Proof. If $A_i \in \mathcal{S}_i$, then

$$\int_{A_i} \frac{d\mu_i}{d(\mu + \nu)_i} d(\mu + \nu) = \mu_i(A_i) = \mu(A) ,$$

so that $d\mu_i/d(\mu + \nu)_i$ is the conditional expectation of $d\mu/d(\mu + \nu)$ with respect to \mathcal{S}_i and the measure $\mu + \nu$. Of course $\mu + \nu$ has total mass 2, but this is inessential; one can always normalize things if so inclined. The Martingale convergence theorem then tells us that

$$\frac{d\mu_i}{d(\mu + \nu)_i} \rightarrow \frac{d\mu}{d(\mu + \nu)}$$

almost everywhere with respect to $\mu + \nu$. Similarly

$$\frac{d\nu_i}{d(\mu + \nu)_i} \rightarrow \frac{d\nu}{d(\mu + \nu)} .$$

Now,

$$\frac{d\mu_i}{d\nu_i} = \frac{d\mu_i}{d(\mu + \nu)_i} \bigg/ \frac{d\nu_i}{d(\mu + \nu)_i} ,$$

so

$$\frac{d\mu_i}{d\nu_i} \rightarrow \frac{d\mu}{d(\mu + \nu)} \bigg/ \frac{d\nu}{d(\mu + \nu)} ,$$

where we understand the right hand side to be $+\infty$ when the

denominator but not the numerator is zero. Now $d\mu/d(\mu + \nu)$ vanishes precisely on $A\mu$, and $d\nu/d(\mu + \nu)$ vanishes precisely on $A\nu$, all statements being up to $(\mu + \nu)$ - measure 0. Whence the lemma.

The following fact is known, and we list it for reference :

LEMMA 4. *If z_1, z_2, \dots are measurable functions with independent Gaussian distributions, mean 0, variance 1, then the product*

$$a_1 \cdots a_k \exp \left\{ \frac{1}{2} \sum_{i=1}^k (1 - a_i^2) Z_i^2 \right\}$$

converges to zero almost everywhere if

$$\sum_{i=1}^{\infty} |1 - a_i^2|^2 = + \infty,$$

and converges to a finite non-zero limit almost everywhere if

$$\sum_{i=1}^{\infty} |1 - a_i^2|^2 < \infty .$$

This can be proven, for example, by applying Kakutani's conditions in [6] for equivalence of product measures.

LEMMA 5. *Let R be a closed densely defined linear operator from the Hilbert space \mathbf{H} to the Hilbert space \mathbf{K} . Then there is an equivalence operator U from \mathbf{H} onto \mathbf{H} such that U^*R^*RU has pure point spectrum.*

Proof. Let $\sqrt{R^*R} = \int_0^{\infty} \lambda dF(\lambda)$. Let $F_i = F(2^i) - F(2^{i-1})$, $i = 0, \pm 1, \pm 2 \dots$, and let $R_i = \sqrt{R^*R} | \mathbf{H}_i$, where \mathbf{H}_i is the range of F_i . By a theorem of von Neumann in [7] there is a self-adjoint Hilbert-Schmidt operator H_i in \mathbf{H}_i whose Hilbert-Schmidt norm $\| H_i \|_2 \leq 2^{-2|i|-3}$, and such that $R_i + H_i$ has pure point spectrum. Now consider the equation $R_i(I + K_i) = R_i + H_i$, that is $R_i K_i = H_i$. Since R_i is invertible in \mathbf{H}_i , and, in fact, $\| R_i \|^{-1} \leq 2^{-i+1}$, we get a solution $K_i = R_i^{-1} H_i$, and

$$\| K_i \|_2 \leq \| R_i^{-1} \| \| H_i \|_2 \leq 2^{-i+1} \cdot 2^{-2|i|-3} \leq 2^{-|i|-2}.$$

Let K be defined on \mathbf{H} by setting $K | \mathbf{H}_i = K_i$. Then

$$\| K \|_2^2 = \sum_{-\infty < i < +\infty} \| K_i \|_2^2 \leq \sum_{-\infty < i < +\infty} 2^{-|i|-2} = 3/4 .$$

so K is Hilbert-Schmidt, and $U = I + K$ is an equivalence. Further, $(I + K_i)^* R_i^2 (I + K_i)$ has a complete set of eigenvectors in \mathbf{H}_i . But this operator is precisely the restriction of U^*R^*RU to \mathbf{H}_i . Therefore U^*R^*RU has a complete set of eigenvectors in \mathbf{H} .

We shall be considering linear spaces of Gaussian functions. In taking the closures of such linear spaces in the $L_2(\mu)$ norm, the functions obtained as limits will again be Gaussian, as is well known and easy to show, the means and variances of a limit being in fact limits of the means and variances of the approximating Gaussian functions. Furthermore, the topology of convergence in measure on Gaussian functions agrees with $L_2(\mu)$ -topology. This is shown in the mean zero case in [8], and the general case can be reduced to this by showing the following :

LEMMA 6. *Let x_i be a net of μ -measurable functions with Gaussian distributions, converging in probability to zero. Then their means γ_i converge to zero.*

Proof. Suppose this does not occur. Then by cutting down to a subnet if necessary, and occasionally using $-x_i$ instead of x_i if necessary, we can assume that there is some $c > 0$ such that $\gamma_i \geq c$ for all i . Now,

$$\mu\{\omega \mid x_i(\omega) - \gamma_i > 0\} = \mu\{\omega \mid x_i(\omega) - \gamma_i < 0\},$$

so that

$$\begin{aligned} \mu\{\omega \mid x_i(\omega) > c\} &\geq \mu\{\omega \mid x_i(\omega) > \gamma_i\} = \mu\{\omega \mid x_i(\omega) < \gamma_i\} \\ &\geq \mu\{\omega \mid x_i(\omega) < c\} \geq \mu\{\omega \mid |x_i(\omega)| < c\}. \end{aligned}$$

The sets on the two ends of the inequality are disjoint, and that on the small end has measure converging to 1, which gives the desired contradiction.

LEMMA 7. *Let μ, ν be nonperpendicular measures. Suppose x_i is Gaussian with respect to μ and ν , $x_i \rightarrow 0$ in μ -measure, and $x_i \rightarrow x$ in ν -measure. Then $x = 0$ a.e. (ν).*

Proof. Since x is Gaussian under ν , the assumption that it is not zero a.e. (ν) implies that it is invertible a.e. (ν). Then $x_i x^{-1} \rightarrow 1$ in ν measure, whereas $x_i \rightarrow 0$ a.e. (μ), which implies $\mu \perp \nu$.

3. The theorem.

THEOREM *Let \mathbf{L} be a linear space of real-valued functions on a set Ω . Let \mathcal{S} be the smallest σ -field of subsets of Ω with respect to which all the functions in \mathbf{L} are measurable. Let μ and ν be probability measures on \mathcal{S} . Suppose all the functions of \mathbf{L} are Gaussian via both measures. Then either $\mu \sim \nu$ or $\mu \perp \nu$. Necessary and sufficient for equivalence is that if we let \mathbf{K} be the linear space generated by \mathbf{L} and the real-valued constant functions, then the μ -equivalence classes of functions in \mathbf{K} are*

the same as the ν -equivalence classes, and the identity correspondence between the two types of equivalence classes in \mathbf{K} is induced by an equivalence operator between the $L_2(\mu)$ -closure of the μ -equivalence classes and the $L_2(\nu)$ -closure of the ν -equivalence classes.

Proof. First, assume μ not $\perp \nu$. Let $\mathbf{J} = \{x - \int x d\mu \mid x \in \mathbf{L}\}$. For any function x , let x^μ (respectively x^ν) denote the equivalence class of x modulo functions which are μ -null (respectively ν -null), and, for a set \mathbf{S} of functions, let $\mathbf{S}_\nu, \mathbf{S}_\mu$ denote the corresponding set of equivalence classes. $\bar{\mathbf{S}}_\mu$ will mean the $L_2(\mu)$ closure of \mathbf{S} .

All elements in \mathbf{K} are Gaussian under μ and ν , and the correspondence $x^\mu \leftrightarrow x^\nu$ between \mathbf{K}_μ and \mathbf{K}_ν is one-to-one and closable, by Lemma 5. So there is a one-to-one closed operator T from a dense subspace \mathbf{D}_T of $\bar{\mathbf{K}}_\mu$ to a dense subspace \mathbf{R}_T of $\bar{\mathbf{K}}_\nu$ such that $Tx^\mu = x^\nu$ for all x in \mathbf{K} . Further, given any ξ in \mathbf{D}_T , there is some \mathcal{S} -measurable x such that $\xi = x^\mu$ and $T\xi = x^\nu$. For choose x_i in \mathbf{K} such that $x_i^\mu \rightarrow \xi, x_i^\nu \rightarrow T\xi$. By taking subsequences, the convergence can be made a.e. (μ) and a.e. (ν) respectively, so that ξ and $T\xi$ must agree a.e. $(\mu \wedge \nu)$.

Let $S = T \mid \mathbf{D}_T \cap \bar{\mathbf{J}}_\mu$. Then S is closed, with dense domain in $\bar{\mathbf{J}}_\mu$ and dense range in $\bar{\mathbf{J}}_\nu$, by Lemma 6. Lemma 5 gives us an equivalence U in $\bar{\mathbf{J}}_\mu$ such that U^*S^*SU has pure point spectrum. Choose y_1, y_2, \dots such that the y_i^μ are a complete orthonormal set of eigenvectors for $U^{-1}S^*U^{-1}$, with eigenvalues a_i^2 . Then the vectors $U^{-1}y_i^\mu$ are again orthogonal, and $\|U^{-1}y_i^\mu\| = a_i^{-1}$. Let \mathcal{S}_N be the sample space of y_1, \dots, y_N . Put a new measure μ' on \mathcal{S} by letting y_1, y_2, \dots be Gaussian, independent, mean 0, variance $1/a_1^2, 1/a_2^2, \dots$ Then

$$\frac{d\mu' \mid \mathcal{S}_N}{d\mu \mid \mathcal{S}_N} = a_1 \cdots a_N \exp\left\{ \frac{1}{2} \sum_{i=1}^N (1 - a_i^2) y_i^2 \right\},$$

which converges almost everywhere $(\mu + \nu)$ to a nonzero limit, so that $\mu' \sim \mu$.

Now we wish to show $\mu' \sim \nu$. We have $x^\nu = Sx^\mu$ for $x \in \mathbf{J}$, so $x^\nu = (SU)U^{-1}x^\mu = (SU)x^{\mu'}$. Let $S' = SU$. Then, taking a.e. limits on both sides, one can for every ξ in $\mathbf{D}_{S'}$ find some \mathcal{S} -measurable x such that $x^{\mu'} = \xi$ and $x^\nu = S'\xi$. Now choose functions Z_1, Z_2, \dots such that $Z_i^{\mu'}$ form a complete orthonormal set of eigenvalues for S'^*S' . The $S'Z_i^{\mu'} = Z_i^\nu$ are also orthogonal and span \mathbf{J}_ν . Define $\int Z_i d\nu = \gamma_i$ and $\int |Z_i|^2 d\nu = \alpha_i^2$. Then α_i is never zero, since S' is nonsingular, and $\int (Z_i - \gamma_i)(Z_j - \gamma_j) d\nu = \alpha_i^2 \delta_{ij} - \gamma_i \gamma_j$, where δ_{ij} is the Kronecker delta function. So the covariance matrix of Z_1, \dots, Z_N in ν -measure is given by $C_N = A_N^2 - \vec{\gamma}_N \otimes \vec{\gamma}_N$, where A_N is the matrix

$$\begin{pmatrix} \alpha_1 & & \\ & \ddots & \\ & & \alpha_N \end{pmatrix},$$

$\vec{\gamma}_N$ is the vector $(\gamma_1, \dots, \gamma_N)$, and the notation $\vec{\gamma}_N \otimes \vec{\gamma}_N$ represents a dyadic operator. C_N is, of course, nonnegative definite. Let $\delta_i = \gamma_i/\alpha_i$, and $\vec{\delta}_N = (\delta_1, \dots, \delta_N)$. So $\vec{\delta}_N = A_N^{-1}\vec{\gamma}_N$, and $A_N^{-1}C_N A_N^{-1} = I - \vec{\delta}_N \otimes \vec{\delta}_N$ is a nonnegative definite matrix. By conjugating with an orthogonal matrix, this can be transformed into the equivalent matrix $I - \|\vec{\delta}_N\|^2 E$, where

$$E = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 0 & \dots & 0 \\ 0 & \dots & \dots & 0 \end{pmatrix}.$$

The determinant of this, and hence of $I - \vec{\delta}_N \otimes \vec{\delta}_N$, is $I - \|\vec{\delta}_N\|^2$. Thus

$$|C_N| = |A_N|^2 |1 - \|\vec{\delta}_N\|^2| = \alpha_1^2 \dots \alpha_N^2 (1 - \sum_{i=1}^N \delta_i^2).$$

Observe that

$$\delta_i = \frac{\gamma_i}{\alpha_i} = \frac{(Z_i, \mathbf{1})}{\sqrt{(Z_i, Z_i)}},$$

where (\cdot, \cdot) denotes the inner product in $L_2(\nu)$, so that, since $\mathbf{1} \notin \mathcal{J}_\nu$, $\sum \delta^2 < (1, 1) = 1$. Thus $|C_N| \neq 0$, and C_N is nonsingular. The inverse matrix to C_N is

$$C_N^{-1} = (A_N^2 - \vec{\gamma}_N \otimes \vec{\gamma}_N)^{-1} = A_N^{-2} + \frac{A_N^{-2}\vec{\gamma}_N \otimes A_N^{-2}\vec{\gamma}_N}{1 - \|A_N^{-1}\vec{\gamma}_N\|^2}.$$

Now let \mathcal{J}_N be the sample space of Z_1, \dots, Z_N and let $\mu_N = \mu' | \mathcal{J}_N$, $\nu_N = \nu | \mathcal{J}_N$. Then $d\nu_N/d\mu_N$ is precisely

$$\sqrt{|C_N|}^{-1} \exp \left\{ \frac{1}{2} [\langle \vec{Z}_N, \vec{Z}_N \rangle - \langle C_N^{-1}(\vec{Z}_N - \vec{\gamma}_N), (\vec{Z}_N - \vec{\gamma}_N) \rangle] \right\},$$

where $\vec{Z}_N = (Z_1, \dots, Z_N)$. A calculation shows that the exponent can be written

$$\frac{1}{2} [\|\vec{Z}_N\|^2 - \|A_N^{-1}\vec{Z}_N\|^2 - \frac{1}{1 - \|\vec{\delta}_N\|^2} (\langle \vec{\delta}_N, A_N^{-1}\vec{Z}_N \rangle - 1)^2 - 1].$$

Consider the convergence of $d\nu_N/d\mu_N$. At this stage one could already conclude from the zero-one law that $\{\omega | d\nu_N/d\mu_N(\omega) \rightarrow 0\}$ has measure 0 or 1, since the set is independent of Z_1, \dots, Z_N for each N . However, we wish to get precise conditions for when this occurs.

(a) Suppose

$$\sum_{i=1}^{\infty} \left| \frac{1}{\alpha_i^2} - 1 \right|^2 = \infty .$$

Consider the factor

$$\begin{aligned} & \sqrt{|C_N|}^{-1} \exp \left\{ \frac{1}{2} [\|\vec{Z}_N\|^2 - \|A_N^{-1}\vec{Z}_N\|^2] \right\} \\ &= \left(1 - \prod_{i=1}^N \delta_i^2 \right)^{-1/2} \left(\frac{1}{\alpha_1} \cdots \frac{1}{\alpha_N} \right) \exp \frac{1}{2} \sum_{i=1}^N \left(1 - \frac{1}{\alpha_i^2} \right) Z_i^2 . \end{aligned}$$

Applying Lemma 3, this factor converges to zero almost everywhere with respect to μ' . The other factor of $d\nu_N/d\mu_N$, namely

$$\exp \frac{1}{2} \left\{ - \left(\frac{\langle \vec{\delta}_N, A_N^{-1}\vec{Z}_N \rangle}{1 - \|\delta_N\|^2} - 1 \right)^2 - 1 \right\} ,$$

is clearly bounded above, so in this case $d\nu_N/d\mu_N \rightarrow 0$.

(b) Suppose

$$\sum_{i=1}^{\infty} \left| \frac{1}{\alpha_i^2} - 1 \right|^2 < \infty .$$

Then, again by Lemma 3, the factor

$$\frac{1}{\alpha_1 \cdots \alpha_N} \exp \frac{1}{2} \sum_{i=1}^N \left(1 - \frac{1}{\alpha_i^2} \right) Z_i^2$$

converges almost everywhere μ' to a finite limit. The remaining factor is, except for a constant,

$$\frac{1}{1 - \|\vec{\delta}_N\|^2} = \exp \frac{1}{2} \left\{ - \frac{1}{1 - \|\delta_N\|^2} \left(\sum_{i=1}^N \frac{\delta_i Z_i}{\alpha_i} - 1 \right)^2 \right\} .$$

Since $\sum_{i=1}^{\infty} \delta_i^2 < 1$, everything in sight converges, because $\sum_{i=1}^{\infty} (\delta_i/\alpha_i)^2 < \infty$. So $\mu' \sim \nu$, and S' is an equivalence from $\bar{\mathbf{J}}_{\mu}$ to $\bar{\mathbf{J}}_{\nu}$. Then S is likewise, and therefore T is an equivalence operator.

Conversely if \mathbf{L} consists of Gaussian functions under μ and ν , and the correspondence $x_{\mu} + c \leftrightarrow x_{\nu} + c$, $x \in \mathbf{L}$, is the restriction of an equivalence operator T from \mathbf{K}_{μ} to \mathbf{K}_{ν} , then again choosing a basis of eigenvectors for $T|_{\bar{\mathbf{J}}_{\mu}}$, we get convergence of the Radon-Nikodym derivatives to a non-zero limit, because of Lemma 3, and therefore get equivalence of the induced measures.

4. **An example.** Let T be a set, Ω the set of all real-valued functions

on T . Let $x_i(\omega) = \omega(t)$, and let \mathcal{S} be the smallest σ -algebra with respect to which all the x_i are measurable. Let μ, ν be measures on \mathcal{S} , by each of which x_i becomes a Gaussian stochastic process. Let

$$m(t) = \int x_i d\mu, \rho(s, t) = \int x_s x_t d\mu - m(s)m(t),$$

$$n(t) = \int x_i d\nu, \sigma(s, t) = \int x_s x_t d\nu - n(s)n(t).$$

Let τ be a measure on T such that ρ, σ, m, n , become measurable. Define, for τ -measurable f

$$[f, f]_\mu = \iint (\rho(s, t) + m(s)m(t))f(s)f(t)d\tau(s)d\tau(t),$$

$$[f, f]_\nu = \iint (\sigma(s, t) + n(s)n(t))f(s)f(t)d\tau(s)d\tau(t),$$

and $[f, f] = [f, f]_\mu + [f, f]_\nu$. Let \mathbf{L}_0 be those f for which $[f, f] < \infty$. Then we get inner products $[f, g]_\mu$, etc. on \mathbf{L}_0 . Define $\int f(t)x_i d\tau(t)$ as an $\mathbf{L}_2(\mu + \nu)$ valued integral, for $f \in \mathbf{L}_0$. This can be done, and in fact

$$[f, g]_\mu = \int \left(\int f(t)x_i d\tau(t) \right) \left(\int g(t)x_s d\tau(s) \right) d\mu,$$

$$[f, g]_\nu = \int \left(\int f(t)x_i d\tau(t) \right) \left(\int g(s)x_s d\tau(s) \right) d\nu.$$

Let $\mathbf{L} = \{x \mid \text{there is some } f \text{ in } \mathbf{L}_0 \text{ for which } x \text{ has as its } \mu + \nu \text{ equivalence class } \int f(t)x_i d\tau(t)\}$. \mathbf{L} is a linear set of functions, all Gaussian with respect to either μ or ν . Let $\mu_0 = \mu \mid \mathcal{S}_0$ and $\nu_0 = \nu \mid \mathcal{S}_0$. We know from our theorem that μ_0 and ν_0 are either equivalent or perpendicular. Let \mathbf{H} and \mathbf{K} be the Hilbert spaces gotten by completing the μ -equivalence classes of \mathbf{L}_0 in $\mathbf{L}_2(\mu)$ and the ν -equivalence classes in $\mathbf{L}_2(\nu)$. The inner products then come from $[f, g]_\mu$ and $[f, g]_\nu$. Necessary for equivalence is that the identity map on \mathbf{L}_0 induce an equivalence operator from \mathbf{H} to \mathbf{K} , and in order to get sufficiency we just have to be sure that if \mathbf{L}_0 is enlarged to include multiples of the identity, the identity map still induces an equivalence operator on the Hilbert spaces. This amounts to requiring that 1 be an $\mathbf{L}_2(\mu)$ limit of functions in \mathbf{L}_0 if and only if it is also an $\mathbf{L}_2(\nu)$ limit.

BIBLIOGRAPHY

1. Baxter : *A strong limit theorem for Gaussian processes*, Proc. Amer. Math. Soc., (1956).
2. Cameron and Martin: *The behavior of measurability under change of scale Wiener space*, Bull. Amer. Math., Soc., (1947).
3. Cameron and Martin: *Transformation of Wiener integrals under translation of Wiener integrals under translation*, Ann. Math., Stat. (1944).

4. Cameron and Martin: *Transformation of Wiener integrals under a general class of linear transformations*, Trans. Amer. Math. Soc., 1945.
5. Grenander: *Stochastic processes and statistical inference*, Arkiv für Matematik, Band 1, nr. 17.
6. S. Kakutani: *On equivalence of infinite product measures*, Ann. Math. Stat., 1948.
7. J. von Neumann: *Charakterisierung des Spektrums eines Integral-operatoren*, Actualités Sci. Ind. 229, Paris (1935).
8. L. E. Segal: *Distributions in Hilbert spaces and canonical systems of operators* (from privately distributed mimeographed notes).

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