

GAUSSIAN RANDOM OPERATORS IN BANACH SPACES (*)

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I. INTRODUCTION

Let X and Y be separable Banach spaces. By a random operator from X into Y we mean a linear continuous operator from X into $L_0(Y)$ where $L_0(Y)$ stands for the set of all Y -valued random variables. For the motivation of the notion of random operator see our recent paper [8] in which the characteristic function, the convergence and the decomposability of random operators have been studied.

This paper which is a continuation of [8] is devoted to the study of Gaussian random operators in Banach spaces. In Section 2 we introduce the definition of covariance operator of Gaussian random operators. This definition extends the notion of covariance operator of Gaussian cylindrical random variables, see [2], [3]. Theorem 2.4 gives the necessary and sufficient condition for an operator to be the covariance operator of some Gaussian random operator. We focus on the problem of π_p -decomposability ($0 < p \leq \infty$) of Gaussian random operators in Section 3. We present conditions for π_p -decomposability of a Gaussian random operator in terms of its covariance operator, which may be considered as an extension of S. A. Chobanian, V. I. Tarieladzes results [1] for Gaussian cylindrical measures.

II. COVARIANCE OPERATOR OF GAUSSIAN RANDOM OPERATORS

Fix a probability space (Ω, \mathcal{F}, P) . Let X and Y be two separable Banach spaces with the duals X' and Y' , respectively. The set of all Y -valued random variables (Y -valued r.v.'s) is denoted by $L_0(Y)$ and is equipped with the topo-

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logy of convergence in probability. By $L_p(Y)$ ($0 < p < \infty$) we denote the space of Y -valued r. v. 's for which $E \|x\|^p < \infty$. When $Y = R$ we write L_p instead of $L_p(R)$

A linear continuous operator A from X into $L_0(Y)$ is called a random operator from X into Y . For some general properties of random operators, see [8].

DEFINITION 2.1. A random operator A from X into Y is called a *Gaussian random operator* if for each $n \in \mathbb{N}$ and x_1, x_2, \dots, x_n in X , the Joint distribution of Ax_1, Ax_2, \dots, Ax_n is Gaussian. Equivalently, a random operator A is Gaussian if and only if the stochastic process (Ax, y) on $X \times Y'$ is Gaussian.

EXAMPLE 1. Let (T, Σ, m) be a finite measurable space. By a Gaussian random measure W on (T, Σ, m) we mean an independently scattered σ -additive setfunction $W: \Sigma \rightarrow L_0$ such that, for each A from Σ , $W(A)$ has a Gaussian distribution with mean 0 and variance $m(A)$. Let Y be a Banach space of type 2. It is known [4] that for each $f \in L_2(T, \Sigma, m; Y)$ the random integral $\int f dW$ is defined. Then a random mapping A from $L_2(T, \Sigma, m; Y)$ into Y given by

$$Af = \int f dW$$

is a Gaussian random operator.

PROPOSITION 2.2. Let A be a Gaussian random operator from X into Y . Then there exists an unique linear continuous operator M from X into Y such that

$$(Mx, y) = E(Ax, y)$$

for each $x \in X, y \in Y'$.

Proof. For each $x \in X$ Ax is an Y -valued Gaussian r. v. so $Ax \in L_1(Y)$. Hence A may be considered as a linear mapping from X into $L_1(Y)$. By using the closed graph theorem it is easy to see that A is continuous. Put

$Mx = E Ax = \int Ax(\omega) P(d\omega)$, M is a linear mapping from X into Y . Moreover

$$\|Mx\| \leq \sup_{\|x\| \leq 1} \|E Ax\| \leq \sup_{\|x\| \leq 1} E \|Ax\| < \infty$$

which shows the continuity of M . By the property of Bochner integral we obtain

$$(Mx, y) = (E Ax, y) = E (Ax, y)$$

The Proposition is proved.

The operator M is called the expectation operator of A . Without loss of generality, from now on we shall always suppose that M is identical to zero.

Let $X \otimes Y'$ be the tensor product of X and Y' . We turn $X \otimes Y'$ into a normed space by considering in it the projective norm defined by

$$\|\Theta\| = \inf \sum \|x_i\| \|y_i\|$$

where the infimum is taken over all finite sets of pairs (x_i, y_i) such that $\theta = \sum x_i \otimes y_i$.

The role of the tensor product is emphasized by the fact that it enables us to replace a bilinear mapping on $X \times Y$ by a linear mapping defined on the tensor product $X \otimes Y$. For more information about the tensor product, we refer to [7].

Let $L(X, Y)$ denote the space of all linear continuous operators from X into Y equipped with the operator norm. For each $u \in L(X, Y)$ we define the following bilinear form on $X \times Y$ $u(x, y) = (ux, y)$.

By the property of the tensor product, u determines a unique linear mapping $u: X \otimes Y \rightarrow R$ such that

$$u(x \otimes y) = (ux, y)$$

From now on, we shall denote by $\langle u, \theta \rangle$ the value of u at the point $\theta \in X \otimes Y$. Note that we have the inequality

$$|\langle u, \theta \rangle| \leq \|u\| \|\theta\| \tag{2-1}$$

Indeed, let $\theta = \sum x_i \otimes y_i$ be an arbitrary representation of θ .

Then

$$\begin{aligned} |\langle u, \theta \rangle| &= |\sum \langle u, x_i \otimes y_i \rangle| \leq \sum |\langle u, x_i \otimes y_i \rangle| = \sum |(x u_i, y_i)| \leq \\ &\sum \|u\| \|x_i\| \|y_i\| = \|u\| \sum \|x_i\| \|y_i\| \end{aligned}$$

From this, the inequality (2-1) follows.

THEOREM 2.3. *Let A be a Gaussian random operator from X into Y . Then there exists a unique linear continuous operator R from $X \otimes Y$ into $L(X, Y)$ such that*

$$\langle R(x \otimes y), s \otimes v \rangle = \text{Cov} \{ (Ax, y), (As, v) \}$$

for all pairs (x, y) and (s, v) in $X \times Y$.

The operator R is called the covariance operator of A .

Proof. Let H_A denote the closed subspace of L_2 spanned by Gaussian $r. v. s$ (Ax, y) . Consider the bilinear mapping T from $X \times Y$ into H_A given by

$$T(x, y) = (Ax, y)$$

By the property of the tensor product T determines a unique linear mapping T from $X \otimes Y$ into H_A such that

$$T(x \otimes y) = (Ax, y)$$

Now we shall show that T is continuous. We have

$$\|T(x \otimes y)\|^2 = \int \|Ax, y\|^2 P(d\omega) \leq \|y\|^2 \int \|Ax(\omega)\|^2 P(d\omega).$$

Because of $Ax \in L_2(Y)$ for each $x \in X$, A may be seen as a linear mapping from X into $L_2(Y)$. By using the closed graph theorem we find that A is continuous. Therefore, there exists a constant C such that

$$\int \|Ax(\omega)\|^2 P(d\omega) \leq C^2 \|x\|^2. \quad (2-2)$$

Thus, we obtain

$$\|T(x \otimes y)\| \leq C \|x\| \|y\|.$$

Let $\theta = \sum x_i \otimes y_i$ be an arbitrary representation of θ . Then $\|T(\theta)\| = \|\sum T(x_i \otimes y_i)\| \leq \sum \|T(x_i \otimes y_i)\| \leq C \sum \|x_i\| \|y_i\|$.

From this we get $\|T(\theta)\| \leq C \|\theta\|$ which shows the continuity of T . Next for each $h \in H_A$ we consider the mapping $Uh: X \rightarrow Y$ given by

$$Uh(x) = \int h(\omega) Ax(\omega) P(d\omega) \quad (2-3)$$

Here the Bochner integral (2-3) exists since

$$\int \|h(\omega) Ax(\omega)\| P(d\omega) \leq [\int |h^2(\omega)| P(d\omega)]^{1/2} [\int \|Ax(\omega)\|^2 P(d\omega)]^{1/2} < \infty. \quad (2-4)$$

Clearly, Uh is linear. From (2-2) and (2-4) we obtain

$$\|Uh(x)\| = \|E hAx\| \leq E \|hAx\| \leq C \|h\| \|x\| \quad (2-5)$$

which shows that Uh is continuous i. e. $Uh \in L(X, Y)$.

Clearly, the mapping $U: H_A \rightarrow L(X, Y)$

$$h \rightarrow Uh$$

is linear. In view of (2-5), we have

$$\|Uh\| = \sup_{\|x\| \leq 1} \|Uh(x)\| \leq C \|h\|,$$

proving that U is continuous.

Moreover, we have the transposition formula

$$(T\theta, h) = \langle Uh, \theta \rangle \quad \text{for } h \in H_A, \theta \in X \otimes Y. \quad (2-6)$$

Indeed, let $\theta = \sum x_i \otimes y_i$. Then

$$\begin{aligned} (T\theta, h) &= \sum (T(x_i \otimes y_i), h) = \sum \int h(\omega) (Ax_i(\omega), y_i) P(d\omega) \\ &= \sum (\int h(\omega) Ax_i(\omega) P(d\omega), y_i) = \sum (Uh(x_i), y_i) = \\ &= \sum \langle Uh, x_i \otimes y_i \rangle = \langle Uh, \theta \rangle \end{aligned}$$

Hence, U is called the transpose of T and denoted by T^* . Set $R = T^*T$. R is a linear continuous operator from $X \otimes Y'$ into $L(X, Y)$. By (2-6) we have

$$\begin{aligned} \langle R(x \otimes y), s \otimes v \rangle &= \langle T^*T(x \otimes y), s \otimes v \rangle = \langle T(x \otimes y), T(s \otimes v) \rangle \\ &= E[(Ax, y)(As, v)] \end{aligned}$$

The proof is thus complete.

Remark. Denote by $X \widehat{\otimes} Y'$ the completion of $X \otimes Y'$. $X \widehat{\otimes} Y'$ is a Banach space and R can be extended to a linear continuous operator from $X \widehat{\otimes} Y'$ into $L(X, Y)$.

The following theorem gives a criterion for a linear continuous operator $R: X \otimes Y' \rightarrow L(X, Y)$ to be the covariance operator of some Gaussian random operator.

THEOREM 2.4 For a linear continuous operator $R: X \otimes Y' \rightarrow L(X, Y)$ to be the covariance operator of some Gaussian random operator, it is necessary and sufficient that

i) R is positive definite i. e.

$$\langle R\theta, \theta \rangle \geq 0 \quad \text{for all } \theta \in X \otimes Y,$$

and symmetric i. e.

$$\langle R\theta_1, \theta_2 \rangle = \langle R\theta_2, \theta_1 \rangle$$

ii) For each $x \in X$ the operator $R_x: Y' \rightarrow Y$ given by

$$R_x(y) = R(x \otimes y)(x)$$

is the covariance operator of some Y -valued Gaussian r. v.

Proof. Suppose that R is the covariance operator of the Gaussian random operator A . Then

$$\langle R\theta, \theta \rangle = \sum_{i,j} E\{(Ax_i, y_i)(Ax_j, y_j)\} = E\{\sum_i (Ax_i, y_i)\}^2 \geq 0,$$

$$\langle R\theta_1, \theta_2 \rangle = \sum_j \sum_i E\{(Ax_i, y_i)(Ax_j, y_j)\} = \langle R\theta_2, \theta_1 \rangle$$

$$\text{where } \theta = \sum_i x_i \otimes y_i, \quad \theta_1 = \sum_i x_i \otimes y_i, \quad \theta_2 = \sum_j x_j \otimes y_j$$

It is clear that R_x is the covariance operator of the Y -valued r. v. Ax

Conversely, suppose that $R: X \otimes Y' \rightarrow L(X, Y)$ is a linear continuous operator satisfying the conditions i) and ii). Consider on $X \otimes Y'$ the function

$$f(\theta) = \exp\{-\langle R\theta, \theta \rangle\}.$$

It is not difficult to check that f satisfies the conditions stated in Theorem 2.3 of [8]. Consequently, there exists a random operator $A: X \rightarrow Y$ such that the joint characteristic function of $(Ax_1, y_1), (Ax_2, y_2), \dots, (Ax_n, y_n)$ is equal to

$$\begin{aligned} E \exp\{i \sum_k t_k (Ax_k, y_k)\} &= f(\sum_k t_k x_k \otimes y_k) = \\ &= \exp\{-\sum_i \sum_j t_i t_j \langle R(x_i \otimes y_i), x_j \otimes y_j \rangle\}. \end{aligned}$$

Thus A is a Gaussian random operator whose covariance operator is precisely R .

COROLLARY 2.5 Let Y be a Hilbert space. A linear continuous operator $R: X \otimes Y' \rightarrow L(X, Y)$ is the covariance operator of some Gaussian random operator $A: X \rightarrow Y$ if and only if R is positive definite, symmetric and $\sum_{i=1}^{\infty} \langle R(x \otimes e_i), x \otimes e_i \rangle < \infty$ where (e_i) is the basis of Y .

PROPOSITION 2.6 Let $R: X \otimes Y' \rightarrow L(X, Y)$ be a covariance operator of a Gaussian random operator. Then there exist a Hilbert space H and a linear continuous operator $T: X \otimes Y' \rightarrow H$ such that R can be factorized as follows

$$\begin{array}{ccc} X \otimes Y' & \xrightarrow{R} & L(X, Y) \\ T \searrow & & \nearrow T^* \\ & H & \end{array}$$

where T^* is the transpose of T in the sense that

$$(T\theta, h) = \langle T^*h, \theta \rangle \quad h \in H, \theta \in X \otimes Y'$$

and H is minimal (i. e. the image of T is dense in H).

Moreover, the operator T is uniquely (up to an isometry equivalence) defined i. e. if R admits a second factorization $R = T_1^* T_1$ where $T_1: X \otimes Y' \rightarrow H_1$ and H_1 is a Hilbert space then there exists an isometry $U: H \rightarrow H_1$ such that $T_1 = U T$.

Proof. In proving Theorem 2.3 we have shown the existence of the desired factorization $R = T^*T$. Suppose that there is another factorization $R = T_1^* T_1$ where $T_1: X \otimes Y' \rightarrow H_1$. Let us define the following mapping $U: T(X \otimes Y') \rightarrow H_1$ by

$$U(T\theta) = T_1\theta$$

Observe that U is well-defined. Indeed if $T\theta = T\theta'$ then $R(\theta - \theta') = T^*T(\theta - \theta') = 0$ which implies $\langle T_1^* T_1(\theta - \theta'), \theta - \theta' \rangle = \|T_1(\theta) - T_1(\theta')\|^2 = 0$

i. e. $T_1\theta = T_1\theta'$. We have $\|UT\theta\|^2 = \|T_1\theta\|^2 = \langle R\theta, \theta \rangle = \|T\theta\|^2$

This means that U is an isometry of $T(X \otimes Y')$ into H_1 . U is extended by continuity to $H = \overline{T(X \otimes Y')}$ and we have $T_1 = UT$.

The operator T in the factorization $R = T^*T$ is denoted by \sqrt{R} and called the square root of R .

DEFINITION 2.7 A Gaussian random operator A is said to be separable if the Hilbert space $H_A \subset L_2$ spanned by Gaussian r. v. s (Ax, y) is separable.

It would be interesting to know when A is separable.

PROPOSITION 2.8 A necessary and sufficient condition for A to be separable is that the image $R(X \otimes Y')$ is separable.

Proof The necessity is clear. Conversely, let $R(X \otimes Y')$ be separable. It is sufficient to prove that the image $T(S) \subset H$ of the unit ball S in $X \otimes Y'$ is separable. Let (y_k) be a sequence in S such that (Ry_k) is dense in $R(S)$. We shall prove that (Ty_k) is dense in $T(S)$. Let $h \in T(S)$. Then there exists an element $y \in S$ such that $h = Ty$. Choosing a subsequence $(y_n) \subset S$ such that (Ry_n) converges to Ry in $L(X, Y)$ we have

$$\|h - Ty_n\|^2 = \|T(y - y_n)\|^2 = \langle R(y - y_n), y - y_n \rangle \leq \|Ry - Ry_n\| \|y - y_n\| \rightarrow 0, \text{ which completes the proof.}$$

THEOREM 2.9. *The random mapping $A : X \rightarrow Y$ is a Gaussian random separable operator if and only if A can be represented in the form*

$$Ax(\omega) = \sum \gamma_n(\omega) B_n x \tag{2-7}$$

where (B_n) is a sequence in $L(X, Y)$, (γ_n) is a sequence of real-valued independent standard Gaussian r. v.'s. The series (2-7) is a.s. convergent in Y .

This representation of A is called the spectral decomposition of A . The sequence (B_n) is called the spectrum of A .

Proof Let A be a Gaussian separable random operator with the covariance operator R . We have the factorization $R = T^*T$ where $T : X \otimes Y' \rightarrow H_A, H_A$ is separable. Take an orthogonal basis in H_A

$$e_n = \gamma_n(\omega) \quad n = 1, 2, \dots$$

representing a sequence of real-valued independent standard Gaussian r. v.'s. Put $B_n = T^*e_n \in L(X, Y)$. For each $x \in X, y \in Y'$ we have

$$\begin{aligned} (Ax, y) &= T(x \otimes y) = \sum (T(x \otimes y), e_n) e_n = \\ &= \sum \langle T^*e_n, x \otimes y \rangle e_n = \sum (B_n x, y) \gamma_n, \end{aligned}$$

where the series converges in L_2 hence in distribution. Thus

$$Ax(\omega) = \sum \gamma_n(\omega) B_n x \text{ for almost every } \omega \text{ by Ito-Nisio's Theorem.}$$

Conversely, if (B_n) is a sequence in $L(X, Y)$ and (γ_n) is a sequence of real-valued independent Gaussian r. v.'s such that for each $x \in X$ the series $\sum \gamma_n B_n x$ is a.s. convergent in Y then by using the Banach-Steinhaus Theorem for random operators ([8]), it is easy to see that the random mapping $A : X \rightarrow Y$ given by.

$$Ax = \sum \gamma_n B_n x$$

is a Gaussian separable random operator.

Remark. If N_x is the set of all ω such that the series (2-7) does not converge to $Ax(\omega)$, then the set on which the convergence fails for at least one $x \in X$ is $N = \bigcup_{x \in X} N_x$, an uncountable union of sets of probability 0, and therefore not

necessarily of probability 0. As we shall see later (Proposition 3-4) the assertion that for ω outside a set of probability 0 the series (2-7) converges to $Ax(\omega)$ for all $x \in X$ holds if and only if A is decomposable.

III. π_p - DECOMPOSABILITY OF GAUSSIAN RANDOM OPERATORS

Recall that a linear operator $u : X \rightarrow Y$ is said to be p -summing ($0 < p < \infty$) if there exists a constant C such that

$$\sum \|ux_n\|^p \leq Cp \sup_{\|x'\| \leq 1} \{|\sum (x_n, x')|^p\} \tag{3-1}$$

for any finite sequence (x_n) in X . Alternatively, u is p -summing if and only if

$\sum \|u x_n\|^p < \infty$ for each sequence $(x_n) \subset X$ such that $\sum |(x_n, x')|^p < \infty$ for all $x' \in X'$.

The minimal C for which the inequality (3-1) holds is denoted by $\pi_p(u)$. The class of all p -summing operators from X into Y is denoted by $\pi_p(X, Y)$. $\pi_p(X, Y)$ is a Banach space equipped with the norm $\|u\| = \pi_p(u)$. If $0 < p < q$ then

$$\pi_p(X, Y) \subset \pi_q(X, Y).$$

One often refers to a linear continuous operator u from X into Y as an ∞ -summing operator.

DEFINITION 3-1. A random operator A from X into Y is said to be π_p -decomposable ($0 < p \leq \infty$) if there exists an $\pi_p(X, Y)$ -valued r.v. B such that $A x(\omega) = B(\omega) x$ for each $x \in X$ and for almost every ω . Instead of saying that A is π_∞ -decomposable, we say that A is decomposable.

PROPOSITION 3.2. For each π_p -decomposable Gaussian random operator A , the decomposing random variable B must be Gaussian. To prove the Proposition 3.2, we need the following

LEMMA 3.1 Suppose that E is a Banach space and M is a linear subspace of E' such that, for all $x \in E$,

$$(x, x') = 0 \text{ for all } x' \in M \text{ implies } x = 0. \quad (3-2)$$

Then an E' -valued r. v. B is Gaussian if for all $x' \in M$, (B, x') is Gaussian.

Proof of Lemma 3.1. We observe that M is dense in E' for the weak topology $\sigma(E', E)$ on E' . Indeed, suppose in the contrary that $M \neq E'$. When E' is equipped with the weak topology, E can be regarded as the dual of E' . By the Hahn-Banach Theorem there exists $x \in E$, $x \neq 0$, such that $(x, x') = 0$ for all $x' \in M$. In view of (3-2) it follows that $x = 0$. A contradiction. Now let x' be an arbitrary element of E' . We have to show that (B, x') is Gaussian. Because M is dense in E' there exists a sequence (x'_n) in M such that (x, x'_n) converges to (x, x') for all $x \in E$. From this $(B(\omega), x'_n)$ converges to $(B(\omega), x')$ for all ω .

As $(B(\omega), x'_n)$ is Gaussian, $(B(\omega), x')$ is Gaussian.

Proof of proposition 3.2. It is clear that every tensor $\theta \in X \otimes Y'$ defines a linear continuous form on $\pi_p(X, Y)$, namely $u \rightarrow \langle u, \theta \rangle$. Moreover, $\langle u, \theta \rangle = 0$ for all $\theta \in X \otimes Y'$ implies $u = 0$. On the other hand, for each $\theta = \sum x_i \otimes y_i$, $\langle B, \theta \rangle = \sum (B x_i, y_i) = \sum (A x_i, y_i)$. Since A is Gaussian, $\langle B, \theta \rangle$ is Gaussian. It then suffices to apply the above lemma.

THEOREM 3.3 Let A be a Gaussian separable random operator with the spectral decomposition

$$Ax = \sum \gamma_n B_n x$$

Then A is π_p -decomposable ($0 < p \leq \infty$) if and only if the sequence (B_n) belongs to $\pi_p(X, Y)$ and the series $\sum \gamma_n B_n$ is a. s. convergent in $\pi_p(X, Y)$.

Proof. The necessity: Suppose that A is π_p -decomposable. By definition there exists an $\pi_p(X, Y)$ -valued r. v. B such that $Ax(\omega) = B(\omega)x$ for each $x \in X$ and for almost every ω . By Proposition 3.2 B is Gaussian. Let \tilde{R} be the covariance operator of B . It is known [9] that \tilde{R} has the factorization $\tilde{R} = \tilde{T}^* \tilde{T}$ where $\tilde{T}: \pi_p(X, Y)' \rightarrow H$ and H is the closed subspace of L_2 spanned by Gaussian r.v.s. (B, u) , $u \in \pi_p(X, Y)'$. When proving Proposition 3.2 we have seen that H is precisely H_A . The sequence (γ_n) represents an orthogonal basis in H_A . It is known [1] that the series $\sum \gamma_n \tilde{T}^* \gamma_n$ is a.s. convergent in $\pi_p(X, Y)$. Now our assertion will follow if we show that $B_n = \tilde{T}^* \gamma_n$. Indeed, for each $x \in X$ and $y \in Y'$ we have

$$\begin{aligned} (\tilde{T}^* \gamma_n(x), y) &= \langle \tilde{T}^* \gamma_n, x \otimes y \rangle = (\gamma_n, \tilde{T}(x \otimes y)) = (\gamma_n, \langle B, x \otimes y \rangle) \\ (\gamma_n, \langle Bx, y \rangle) &= (\gamma_n, \langle Ax, y \rangle) = \gamma_n, \sum (B_k x, y) \gamma_k = (B_n x, y) \end{aligned}$$

Consequently, $B_n = \tilde{T}^* \gamma_n$, as desired.

Conversely, suppose that the series $\sum \gamma_n B_n$ is a.s. convergent in $\pi_p(X, Y)$. Set $B = \sum \gamma_n B_n$. B is an $\pi_p(X, Y)$ -valued r.v. and for each $x \in X$ we have

$$B(\omega)x = \sum \gamma_n(\omega) B_n x \quad \text{for almost every } \omega.$$

So

$$B(\omega)x = Ax(\omega) \quad \text{for almost every } \omega,$$

i. e. A is π_p -decomposable.

PROPOSITION 3.4 *Let A be a Gaussian separable random operator with the spectral decomposition (2-7). Then A is decomposable if and only if there exists set N of probability 0 such that if $\omega \notin N$ then the series*

$$\sum \gamma_n(\omega) B_n x$$

is convergent in Y for all $x \in X$.

Proof. Suppose that A is decomposable. From Theorem 3.3 it follows that there exists an $L(X, Y)$ -valued r.v. B and a set of probability 0 such that if $\omega \notin N$ then

$$B(\omega) = \sum \gamma_n(\omega) B_n \quad \text{in } L(X, Y)$$

Therefore, for all $x \in X$ and $\omega \notin N$ we have

$$B(\omega)x = \sum \gamma_n(\omega) B_n x$$

Conversely, for each $\omega \notin N$ we define a mapping $B(\omega): X \rightarrow Y$

by
$$B(\omega)x = \sum \gamma_n(\omega) B_n x.$$

The Banach-Steinhaus Theorem shows that $B(\omega) \in L(X, Y)$. Thus we have an $L(X, Y)$ -valued r.v. B such that

$$B(\omega)x = \sum \gamma_n(\omega) B_n x \quad \text{for almost every } \omega$$

So

$$B(\omega)x = Ax(\omega) \quad \text{for almost every } \omega,$$

i. e. A is decomposable.

The following result is basis

THEOREM 3.5. Let A and A_I be two Gaussian separable random operators with the covariance operators R and R_I , respectively. Suppose that for all $\theta \in X \otimes Y$ $\langle R_I \theta, \theta \rangle \leq \langle R \theta, \theta \rangle$ and A is π_p -decomposable ($0 < p \leq \infty$) Then A_I is also π_p -decomposable.

We begin with the following

LEMMA. Let R be the covariance operator of the Gaussian separable random operator A . Then we have

$$\langle R \theta, \theta \rangle = \sum |\langle B_n, \theta \rangle|^2$$

where (B_n) is the spectral sequence of A .

Proof. Let $T: X \otimes Y \rightarrow H_A$ be the square root of R and (γ_n) —an orthogonal basis in H_n . We have

$$T\theta = \sum \langle T\theta, \gamma_n \rangle \gamma_n = \sum \langle T^* \gamma_n, \theta \rangle \gamma_n$$

We have seen that (Theorem 2.9) $B_n = T^* \gamma_n$. From this we get $\langle R \theta, \theta \rangle =$
 $= \|T\theta\|^2 = \sum |\langle T^* \gamma_n, \theta \rangle|^2 = \sum |\langle B_n, \theta \rangle|^2.$

Proof of Theorem 3.5. We shall split the proof into two steps.

Step 1. Suppose that $\langle R_I \theta, \theta \rangle = \langle R \theta, \theta \rangle$ for all $\theta \in X \otimes Y$. Let (B_n) and (B_n^I) be two spectral sequences of A and A_I , respectively. By the above lemma we have

$$\langle R \theta, \theta \rangle = \sum |\langle B_n, \theta \rangle|^2 = \langle R_I \theta, \theta \rangle = \sum |\langle B_n^I, \theta \rangle|^2 \quad (3.3)$$

At first we show that $B_n^I \in \pi_p(X, Y)$ ($n=1, 2, \dots$) Suppose that A is π_p -decomposable by $\pi_p(X, Y)$ valued r.v. B . Then for each $\theta \in X \otimes Y$ the r.v. $\langle B, \theta \rangle$ is Gaussian with variance $\langle R \theta, \theta \rangle$. Hence for each $p > 0$ there exists a constant C_p such that

$$|\langle R \theta, \theta \rangle|^{p/2} = C_p \int_{\Omega} |\langle B(\omega), \theta \rangle|^p P(d\omega).$$

In view of (3-3) we have

$$|\langle B_n^I x, y \rangle|^p \leq \langle R(x \otimes y), x \otimes y \rangle^{p/2} = C_p \int |(B(\omega)x, y)|^p P(d\omega) \leq C_p \|y\|^p \int \|B(\omega)x\|^p P(d\omega).$$

Hence

$$\|B_n^I x\|^p \leq C_p \int \|B(\omega)x\|^p P(d\omega)$$

For any finite sequence (x_k) in x we have

$$\sum \|B_n^I x_k\|^p \leq C_p \int \sum \|B(\omega)x_k\|^p P(d\omega) \leq$$

$$C_p \int \|B(\omega)\|^p \sup \{ \sum | \langle x_n, x' \rangle |^p \} P(d\omega) = C_p C \sup \{ \sum | \langle x_n, x' \rangle |^p \}, \text{ where } \|x'\| \leq 1$$

$C = \int \|B(\omega)\|^p P(d\omega) < \infty$ (since B is Gaussian). Consequently, B_n is p -summing.

Let $A_1 x(\omega) = \sum \gamma_n \omega \binom{I}{n} x$ be the spectral decomposition of A_1 . Consider the series $\sum \gamma_n(\omega) B_n^I$ in $\pi_p(X, Y)$. For each $\theta \in X \otimes Y'$ we have

$$E \exp \left\langle \sum_{k=1}^n \gamma_k(\omega) B_k^I, \theta \right\rangle = E \exp \sum_{k=1}^n \gamma_k(\omega) \langle B_k^I, \theta \rangle = \exp \left\{ -\frac{1}{2} \sum_{k=1}^n |\langle B_k^I, \theta \rangle|^2 \right\}$$

converging to $\exp \left\{ -\frac{1}{2} \sum_{k=1}^{\infty} |\langle B_k^I, \theta \rangle|^2 \right\} = \exp \left\{ -\frac{1}{2} \langle R\theta, \theta \rangle \right\} = E \exp \{i \langle B, \theta \rangle\}$.

By using the fact that $X \otimes Y'$ is a total linear subspace of the dual $\pi_p(X, Y)'$ and the same argument as in the proof of the Ito-Nisio theorem we find $\pi_p(X, Y)$ valued *r. v. s* such that

$$\langle S, \theta \rangle = \sum \gamma_n(\omega) \langle B_n^I, \theta \rangle \quad \text{P-a.s.}$$

for each $\theta \in X \otimes Y'$

In particular, taking $\theta = x \otimes y$ we get

$\langle S(\omega) x, y \rangle = \sum \gamma_n(\omega) \langle B_n x, y \rangle = \langle Ax(\omega), y \rangle$ for almost every ω . From this it follows that

$A_1 x(\omega) = S(\omega) x$ for almost every ω , i. e. A_1 is π_p -decomposable.

Step 2. Suppose that $\langle R_1 \theta, \theta \rangle \leq \langle R\theta, \theta \rangle$ for all θ . Put $R_2 = R - R_1$. By using Theorem 2.4 it is easy to show that R_2 is a covariance operator of some Gaussian separable random operator, say A_2 . By the above lemma we have

$$\langle R_1 \theta, \theta \rangle = \sum |\langle B_n^1, \theta \rangle|^2, \quad \langle R_2 \theta, \theta \rangle = \sum |\langle B_n^2, \theta \rangle|^2$$

where (B_n^1) and (B_n^2) are two spectral sequences of A_1 and A_2 , respectively.

Setting

$$C_{2n-1} = B_n^1, \quad C_{2n} = B_n^2 \quad (3-4)$$

we have

$$\langle R\theta, \theta \rangle = \langle R_1 \theta, \theta \rangle + \langle R_2 \theta, \theta \rangle = \sum |\langle C_n, \theta \rangle|^2 \quad (3-5)$$

Now let us consider the spectral decomposition of A

$$Ax = \sum \gamma_n(\omega) B_n x$$

and the series

$$\sum \gamma_n(\omega) C_n x$$

For each $y \in Y'$ we have

$E \exp \left\{ \left\langle \sum_{k=1}^n \gamma_k(\omega) C_k x, y \right\rangle \right\} = \exp \left\{ -\frac{1}{2} \sum_{k=1}^n |\langle C_k x, y \rangle|^2 \right\}$ converging to

$\exp \left\{ -\frac{1}{2} \sum_{k=1}^{\infty} |\langle C_k x, y \rangle|^2 \right\} = \exp \left\{ -\frac{1}{2} \langle R(x \otimes y), x \otimes y \rangle \right\} = E \exp \{i \langle Ax, y \rangle\}$

By the Ito — Nisio theorem we conclude that the series $\sum \gamma_n(\omega) C_n x$ converges a.s. in Y for each $x \in X$. Set

$$\tilde{A} x(\omega) = \sum \gamma_n(\omega) C_n x.$$

\tilde{A} is a Gaussian separable random operator. By lemma and (3—5) we get

$$\langle \tilde{R}\theta, \theta \rangle = \langle R\theta, \theta \rangle$$

where \tilde{R} is the covariance operator of \tilde{A} .

From step 1 it follows that \tilde{A} is π_p -decomposable. Hence, by Theorem 3.3 the series $\sum \gamma_n(\omega) C_n$ is a.s. convergent in $\pi_p(X, Y)$. For any bounded sequence of real numbers (t_k) the series $\sum t_k \gamma_k(\omega) C_k$ is also a.s. convergent in $\pi_p(X, Y)$. If we put

$$t_{2n-1} = 1, t_{2n} = 0 \quad \tilde{\gamma}_n = \gamma_{2n-1},$$

then by (3—4) the series $\sum \tilde{\gamma}_n B_n^1$ is a.s. convergent in $\pi_p(X, Y)$.

Put

$$\tilde{A}_1 x(\omega) = \sum \tilde{\gamma}_n(\omega) B_n^1 x.$$

\tilde{A}_1 is π_p -decomposable and we have

$$\langle \tilde{R}_1 \theta, \theta \rangle = \sum |\langle B_n^1, \theta \rangle|^2 = \langle R_1 \theta, \theta \rangle$$

From step 1 it follows that \tilde{A}_1 is π_p -decomposable. The proof of the Theorem is thus complete.

COROLLARY 3.6. *Let A and B be two independent Gaussian separable random operators with the expectation operators zero. If $A + B$ is π_p -decomposable then both A and B are π_p -decomposable.*

In the sequel we shall find conditions on the covariance operator R such that the corresponding Gaussian random operator A in π_p -decomposable.

DEFINITION 3.7 Let Z be a Banach space. A linear operator T from $X \otimes Y'$ into Z is said to be (r, π_p) -summing ($0 < r < \infty, 0 < p \leq \infty$) if for each sequence (θ_n) in $X \otimes Y'$ such that $\sum |\langle u, \theta_n \rangle|^r < \infty$ for all $u \in \pi_p(X, Y)$ we have $\sum \|T \theta_n\|^r < \infty$. Equivalently, T is (r, π_p) -summing if and only if there exists a constant C such that

$$\sum \|T \theta_n\|^r \leq C^r \sup_{\substack{\|u\| \leq 1 \\ \pi_p}} \{ \sum |\langle u, \theta_n \rangle|^r \}$$

for any finite sequence (θ_n) in $X \otimes Y'$. Because of $\pi_p(X, Y) \subset \pi_q(X, Y)$ whenever $p < q$, the (r, π_p) -summing operators are (r, π_q) -summing if $p < q$. As we shall see later, the converse is not generally true, unless Y is finite-dimensional.

By the same argument as in the proof of Pietsch's theorem we get

THEOREM 3.8 A continuous linear operator $T : X \otimes Y' \rightarrow Z$ is (r, π_p) -summing if and only if there exists a finite measure μ on the unit ball U of $\pi_p(X, Y)$ such that

$$\|T \theta\|^r \leq \int_U |\langle u, \theta \rangle|^r \mu(du)$$

for all $\theta \in X \otimes Y'$.

THEOREM 3.9 Let A be a Gaussian random operator from X into Y with the covariance operator R . If A is π_p -decomposable then the operator $T = \sqrt[r]{R}$ is (r, π_p) -summing for all $r > 0$.

Proof Suppose that A is π_p -decomposable by an $\pi_p(X, Y)$ -valued r.v. B . Then for each $\theta \in X \otimes Y'$ the r.v. $\langle B, \theta \rangle$ is Gaussian with variance $\langle R \theta, \theta \rangle$. Hence for each $r > 0$ there exists a constant C_r such that

$$\|T \theta\|^r = C_r \int_{\Omega} |\langle B(\omega), \theta \rangle|^r P(d\omega)$$

For each finite sequence (θ_n) in $X \otimes Y'$ we have

$$\begin{aligned} \sum \|T \theta_n\|^r &= C_r \int \sum |\langle B(\omega), \theta_n \rangle|^r P(d\omega) = \\ &C_r \int \|B(\omega)\|^r \sum \left\| \frac{B(\omega)}{\|B(\omega)\|}, \theta_n \right\|^r P(d\omega) \leq \\ &C_r \int \|B(\omega)\|^r P(d\omega) \left\{ \sup_{\|u\| \leq 1} \sum |\langle u, \theta_n \rangle|^r \right\} = C_r C \sup_{\|u\| \leq 1} \sum |\langle u, \theta_n \rangle|^r \end{aligned}$$

where $C = \int \|B(\omega)\|^r P(d\omega) < \infty$ (since B is Gaussian).

Thus T is (r, π_p) -summing

THEOREM 3.10. Suppose that $\pi_p(X, Y)$ is of type 2. Then a Gaussian separable random operator $A : X \rightarrow Y$ is π_p -decomposable if and only if the operator $T = \sqrt[r]{R}$ is $(2, \pi_p)$ -summing, where R is the covariance operator of A .

Proof. By Theorem 3.9 it remains to prove the «if» part. Assume that T is $(2, \pi_p)$ -summing. By Theorem 3.8 there exists a finite measure μ on the unit ball U of $\pi_p(X, Y)$ such that

$$\|T \theta\|^2 \leq \int_U |\langle u, \theta \rangle|^2 \mu(du). \quad (3-6)$$

Since $\pi_p(X, Y)$ is of type 2 by a result in [4], there exists an $\pi_p(X, Y)$ -valued Gaussian r. v. B such that

$$E \exp \{i \langle B, u' \rangle\} = \exp \left\{ -\frac{1}{2} \int_U \left| \langle u, u' \rangle \right|^2 \mu(du) \right\}$$

for all $u' \in \pi_p(X, Y)$. Taking $\theta = u'$ we get

$$E \exp \{i \langle B, \theta \rangle\} = \exp \left\{ -\frac{1}{2} \int_U \left| \langle u, \theta \rangle \right|^2 \mu(du) \right\}. \quad (3-7)$$

Let \tilde{A} be the Gaussian random operator generated by B , i. e.

$$\tilde{A}x(\omega) = B(\omega)x.$$

From (3-7) we have

$$\langle \tilde{R}\theta, \theta \rangle = \int_U |\langle u, \theta \rangle|^2 \mu(du), \text{ where } \tilde{R} \text{ is the covariance operator of } \tilde{A}.$$

From (3-6) we get $\langle R\theta, \theta \rangle \leq \langle \tilde{R}\theta, \theta \rangle$. By Theorem 3.5 we conclude that A is π_p -decomposable.

THEOREM 3.11. *Suppose that $\pi_p(X, Y)$ is of cotype 2. Then a Gaussian separable random operator $A: X \rightarrow Y$ is π_p -decomposable if and only if the transpose T^* of $T = \sqrt{R}$ is an 2-summing operator from H_A into $\pi_p(X, Y)$, where R is the covariance operator of A .*

Proof Suppose that A is π_p -decomposable. Then there exists an $\pi_p(X, Y)$ -valued Gaussian r.v. B such that

$$Ax(\omega) = B(\omega)x \quad \text{for almost every } \omega$$

Let $\tilde{R}: \pi_p(X, Y) \rightarrow \pi_p(X, Y)$ be the covariance operator of B . \tilde{R} has the factorization $\tilde{R} = \tilde{T}^* \tilde{T}$ where $\tilde{T}: \pi_p(X, Y)' \rightarrow H_A$ (see the proof of Theorem 3.3). Since $\pi_p(X, Y)$ is of cotype 2 by the result of [1], the operator $\tilde{T}: H_A \rightarrow \pi_p(X, Y)$ is 2-summing. Now our assertion will follow if we show that $T^* \equiv \tilde{T}^*$. Indeed for each $h \in H_A$, $x \in X$ and $y \in Y'$ we have

$$\begin{aligned} \langle \tilde{T}^*h(x), y \rangle &= \langle \tilde{T}^*h, x \otimes y \rangle = (h, \tilde{T}(x \otimes y)) = (h, \langle B, x \otimes y \rangle) \\ &= (h, \langle B(\omega)x, y \rangle) = (h, \langle Ax, y \rangle) = (h, T(x \otimes y)) = \langle T^*h, x \otimes y \rangle \\ &= T^*(h(x), y). \end{aligned}$$

Hence $\tilde{T}^* \equiv T^*$ as desired.

Conversely, assume that $T^*: H_A \rightarrow \pi_p(X, Y)$ is 2-summing. In view of the Schwartz Radonification theorem [6] $\mu = T^*(\gamma_2)$ is a Radon measure on $\pi_p(X, Y)$, where γ_2 is the Gaussian cylindrical measure on H_A with the characteristic

function $\exp \left\{ -\frac{\|h\|^2}{2} \right\}$. The characteristic function $\hat{\mu}$ is equal to

$$\hat{\mu}(u') = \exp \left\{ -\frac{1}{2} \|(T^*)^*u'\|^2 \right\} \quad (u' \in \pi_p(X, Y)')$$

where $(T^*)^*: \pi_p(X, Y)' \rightarrow H_A$.

For each $h \in H_A$ and $\theta \in X \otimes Y'$ we have

$$\langle (T^*)^*\theta, h \rangle = \langle \theta, T^*h \rangle = \langle T\theta, h \rangle$$

which shows that $(T^*)^*\theta = T\theta$. Hence

$$\hat{\mu}(\theta) = \exp \left\{ -\frac{\|T\theta\|^2}{2} \right\}.$$

For each $\theta \in X \otimes Y'$ we have

$$E \exp \left\langle \sum_{k=1}^n \gamma_k(\omega) B_k, \theta \right\rangle = \exp \left\{ -\frac{1}{2} \sum_{k=1}^n |\langle B_k, \theta \rangle|^2 \right\} \text{converging to}$$

$$\exp \left\{ -\frac{1}{2} \sum_{n=1}^{\infty} |\langle B_n, \theta \rangle|^2 \right\} = \exp \left\{ -\frac{1}{2} \|T\theta\|^2 \right\} = \widehat{\mu}(\theta),$$

where (B_n) is the spectral sequence of A .

By the same argument as in proving Step 1 of Theorem 3.5 we conclude that A is π_p -decomposable.

Remark. Let us observe that the «if» part is always true without any additional assumption on $\pi_p(X, Y)$.

Theorems 3.10 and 3.11 lead us to the question of which Banach spaces X and Y have the property that $\pi_p(X, Y)$ is of type 2 or is of cotype 2.

THEOREM 3.12. *Let H be a Hilbert space. Then*

- 1) $\pi_2(X, H)$ is of type 2 if X is the dual of a Banach space of type 2.
- 2) $\pi_2(H, Y)$ is of cotype 2 if Y is of cotype 2.

Proof 1) Assume that $X = E'$. Let $\Lambda_2(E', H)$ denote the set of all operators T from E' into H for which the function

$$y \rightarrow \exp \{ -\|Ty\|^2 \}$$

is a characteristic function of some E -valued Gaussian r. v. X_T . It is known [5] that $\Lambda_2(E', H)$ is a Banach space equipped with the norm.

$$\|T\|_{\Lambda_2}^2 = E \|X_T\|^2.$$

The correspondence $T \rightarrow X_T$ is an isometry of $\Lambda_2(E', H)$ into $L_2(E)$

When E is of type 2, it is known [1] that $\Lambda_2(E', H) = \pi_2(E', H)$. Hence there exist two constants C_1 and C_2 such that

$$C_2 \|T\|_{\Lambda_2}^2 \leq \|T\|_{\pi_2}^2 \leq C_1 \|T\|_{\Lambda_2}^2$$

Let r_1, r_2, \dots be the Rademacher sequence on the probability space $([0, 1], \mathcal{B}, dt)$ and let T_1, T_2, \dots, T_n be a finite sequence in $\Lambda_2(E', H)$. Then

$$\int_0^1 \|\sum_{n=1}^n T_n r_n(t)\|_{\Lambda_2}^2 dt = \int_0^1 E \|\sum_{n=1}^n X_{T_n} r_n(t)\|^2 dt =$$

$$E \int_0^1 \|\sum_{n=1}^n X_{T_n} r_n(t)\|^2 dt \leq K E \{ \sum_{n=1}^n \|X_{T_n}\|^2 \} = K \sum_{n=1}^n \|T_n\|_{\Lambda_2}^2$$

The last inequality used the fact that E is of type 2.

Hence $\Lambda_2(E', H)$ is of type 2 so is $\pi_2(E', H)$

2) Suppose Y is of cotype 2. Then by the results of [1] $T \in \pi_2(H, Y)$ if and only if $T^* \in \Lambda_2(Y', H)$ and we have the constants C_1, C_2 such that \circledast

$$\bullet \quad C_2 \|T^*\|_{\Lambda_2}^2 \leq \|T\|_{\pi_2}^2 \leq C_1 \|T^*\|_{\Lambda_2}^2.$$

Let T_1, T_2, \dots, T_n be a finite sequence in $\pi_2(H, Y)$. Then

$$\begin{aligned} \int_0^1 \|\Sigma T_n r_n(t)\|_{\pi_2}^2 dt &\geq C_2 \int_0^1 \|\Sigma T_n^* r_n(t)\|_{\Lambda_2}^2 dt = \\ C_2 \int_0^1 E \|\Sigma X_{T_n^*} r_n(t)\|^2 dt &= C_2 E \int_0^1 \|\Sigma X_{T_n^*} r_n(t)\|^2 dt \geq \\ C_2 C E \{\Sigma \|X_{T_n^*}\|^2\} &= C_2 C \Sigma \|T_n^*\|_{\Lambda_2}^2 \geq C_1^{-1} C C_2 \Sigma \|T_n\|_{\pi_2}^2 \end{aligned}$$

This proves that $\pi_2(H, Y)$ is of cotype 2.

PROPOSITION. 3.13 Let X, Y be two separable Hilbert spaces and $A: X \rightarrow Y$ a Gaussian separable random operator. Then the following assertions are equivalent

- 1) A is π_p - decomposable for $0 < p < \infty$
- 2) A is π_2 - decomposable.
- 3) $\sum_j \sum_i \langle R(e_i \otimes f_j), e_i \otimes f_j \rangle < \infty$
- 4) $\sum_j \sum_i \|B_j e_i\|^2 < \infty$.

Here R is the covariance operator of A , (B_j) is the spectral sequence of A and $(e_i), (f_j)$ are the bases in X, Y , respectively.

Proof 1) \leftrightarrow 2) is trivial because all the classes $\pi_p(X, Y)$ coincide ($0 < p < \infty$).

2) \rightarrow 3): By Theorem 3.10 the operator $T = \sqrt{R}$ is $(2, \pi_2)$ -summing. For each $u \in \pi_2(X, Y)$ we have

$$\Sigma \sum_i |\langle u, e_i \otimes f_j \rangle|^2 = \Sigma \sum_{i,j} |\langle u e_i, f_j \rangle|^2 = \Sigma \|u e_i\|^2 < \infty.$$

So

$$\Sigma \sum \langle R(e_i \otimes f_j), e_i \otimes f_j \rangle = \Sigma \sum \|T(e_i \otimes f_j)\|^2 < \infty.$$

3) \rightarrow 4): It is not difficult to check that

$$\Sigma \sum \|B_j e_i\|^2 = \Sigma \sum \langle R(e_i \otimes f_j), e_i \otimes f_j \rangle.$$

4) \rightarrow 2): We have $\Sigma \|B_n\|_{\pi_2}^2 = \Sigma \sum_{i,j} \|B_j e_i\|^2 < \infty$.

Because $\pi_2(X, Y)$ is a Hilbert space it follows that the series $\Sigma \gamma_n B_n$ is a.s. convergent in $\pi_2(X, Y)$. By Theorem 3.3 A is π_2 -decomposable.

Finally, we give an example showing that there exists an (r, π_∞) -summing operator which is not (r, π_p) -summing for $p < \infty$.

Let H be a Hilbert space with the basis (e_n) and (s_n) be a sequence of numbers such that $\sup |s_n| < \infty$.

We define a Gaussian separable random operator A from H into H by means of.

$$Ax(\omega) = \sum (x, e_n) e_n s_n \gamma_n(\omega)$$

By Proposition 3.13 A is π_p -decomposable ($p < \infty$) if and only if.

$$\sum_{i,j} \|B_j e_i\|^2 = \sum |s_n|^2 < \infty. \quad (3-8)$$

On the other hand, as shown in [8], A is decomposable if and only if.

$$\sum \exp \left\{ -\frac{t}{|s_n|^2} \right\} < \infty \text{ for some } t > 0. \quad (3-9)$$

So if the sequence (s_n) is chosen such that (3-9) holds but (3-8) fails then by Theorem 3.10 and Theorem 3.9 the operator $T = \sqrt{R} : H \otimes H \rightarrow L_2$ is (r, π_∞) -summing but not (r, π_p) -summing ($p < \infty$).

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