

## INVERSION OF CONDITIONAL EXPECTATIONS

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**By its definition a conditional expectation is the Radon-Nikodym derivative of a finite signed measure. In this paper an inversion formula is given for recapturing  $E(Y|X)$  as an inverse Fourier transform of the function  $E(e^{i(u,X)}Y)$ ,  $u \in R^n$ , where  $X$  is a random vector and  $Y$  is a random variable satisfying some regularity conditions.**

1. Introduction. Let  $(\Omega, \mathfrak{B}, P)$  be a probability space and let  $X$  be a  $k$ -dimensional random vector on  $(\Omega, \mathfrak{B}, P)$ , i.e., a measurable transformation of  $(\Omega, \mathfrak{B})$  into  $(R^k, \mathfrak{B}^k)$  where  $\mathfrak{B}^k$  is the  $\sigma$ -algebra of Borel sets in the  $k$ -dimensional Euclidean space  $R^k$ . Assume that the probability distribution  $X$  is absolutely continuous with respect to the Lebesgue measure  $m_L$  on  $(R^k, \mathfrak{B}^k)$ . For a real valued random variable  $Y$  on  $(\Omega, \mathfrak{B}, P)$  with  $E(|Y|) < \infty$  let  $E(Y|X)$  be the conditional expectation of  $Y$  given  $X$  which is given as a function on the value space  $R^k$  of  $X$ . For  $u \in R^k$  let  $(u, X) = \sum_{j=1}^k u_j X_j$ . In this paper we show that if  $E[e^{i(u,X)}Y]$  is a  $m_L$ -integrable function of  $u$  on  $R^k$  then a version of  $E(Y|X)$  is given by

$$(1.1) \quad E(Y|X)(\xi) = \left( \frac{dP_X}{dm_L}(\xi) \right)^{-1} \frac{1}{(2\pi)^k} \int_{R^k} e^{-i(u,\xi)} E[e^{i(u,X)}Y] m_L(du)$$

for  $\xi \in R^k$  assuming that  $(dP_X/dm_L)(\xi) > 0$  for a.e.  $\xi$  in  $(R^k, \mathfrak{B}^k, m_L)$ . (Our conditional expectation  $E(Y|X)$  given as a function on  $R^k$  rather than one on  $\Omega$  is the "conditional expectation in the wide sense" in the terminology of [2].)

In preparation for (1.1) which is given in Theorem 2 in §3 we show in Theorem 1 in §3 that if the characteristic function (i.e., the Fourier transform)  $\varphi$  of a finite measure  $\Phi$  on  $(R^k, \mathfrak{B}^k)$  is  $m_L$ -integrable on  $R^k$  then  $\Phi$  is absolutely continuous with respect to  $m_L$  on  $(R^k, \mathfrak{B}^k)$  and a version of the Radon-Nikodym derivative of  $\Phi$  with respect to  $m_L$  is given by the inverse Fourier transform of  $\varphi$ . We base this result on the Lévy-Haviland Theorem for the inversion of Fourier transforms of finite measures on  $(R^k, \mathfrak{B}^k)$ .

The substance of Propositions 1, 2, and 3 in §2 concerning conditional probabilities, conditional expectations and regular conditional distributions given as functions on the value space of  $X$  is well-known. We included them here in order to state them in a convenient form.

This research is an attempt at justifying a calculus of Wiener integral originated by M. D. Donsker. Its applications to conditional

function space integrals will appear in a subsequent paper.

2. **Integration of conditional expectations.** Throughout §2 we write  $(\Omega, \mathfrak{B}, P)$  for a probability space and  $X$  and  $Y$  for two measurable transformations of  $(\Omega, \mathfrak{B})$  into two arbitrary measurable spaces  $(S, \mathfrak{F})$  and  $(T, \mathfrak{G})$  respectively unless further specified. We write  $P_X$  and  $P_Y$  for the probability measures on  $(S, \mathfrak{F})$  and  $(T, \mathfrak{G})$  determined by  $X$  and  $Y$  respectively, i.e.,

$$(2.1) \quad P_X(F) = P(X^{-1}(F)) \quad \text{for } F \in \mathfrak{F}$$

and similarly for  $P_Y$ .

**DEFINITION 1.** For  $G \in \mathfrak{G}$  fixed, the conditional probability of  $Y$  being in  $G$  given  $X$ , written  $P(Y \in G | X)$ , is defined to be any real valued  $\mathfrak{F}$ -measurable and  $P_X$ -integrable function  $\psi$  on  $S$  such that

$$P(Y^{-1}(G) \cap X^{-1}(F)) = \int_F \psi(\xi) P_X(d\xi) \quad \text{for } F \in \mathfrak{F}.$$

From the Radon-Nikodym Theorem follows that such a function  $\psi$  always exists and is determined uniquely up to a null set of  $(S, \mathfrak{F}, P_X)$ . We shall use  $P(Y \in G | X)$  to mean either the class of all such functions  $\psi$  or a particular member in it depending on the context. Thus

$$(2.2) \quad P(Y^{-1}(G) \cap X^{-1}(F)) = \int_F P(Y \in G | X)(\xi) P_X(d\xi) \quad \text{for } F \in \mathfrak{F}.$$

**DEFINITION 2.** Let  $Z$  be a real valued random variable on  $(\Omega, \mathfrak{B}, P)$  with  $E(|Z|) < \infty$ . The conditional expectation of  $Z$  given  $X$ , written  $E(Z | X)$ , is defined to be any real valued  $\mathfrak{F}$ -measurable and  $P_X$ -integrable function  $\psi$  on  $S$  such that

$$\int_{X^{-1}(F)} Z(\omega) P(d\omega) = \int_F \psi(\xi) P_X(d\xi) \quad \text{for } F \in \mathfrak{F}.$$

The same remark as the one following Definition 1 holds here too and we have

$$(2.3) \quad \int_{X^{-1}(F)} Z(\omega) P(d\omega) = \int_F E(Z | X)(\xi) P_X(d\xi) \quad \text{for } F \in \mathfrak{F}.$$

**DEFINITION 3.** By the regular conditional distribution of  $Y$  given  $X$ , written  $P(Y | X)$ , we mean a real valued function  $\psi$  on  $\mathfrak{G} \times S$  such that

- 1° for every  $G \in \mathfrak{G}$ ,  $\psi(G, \cdot)$  is a version of  $P(Y \in G | X)$
- 2° for every  $\xi \in S$ ,  $\psi(\cdot, \xi)$  is a probability measure on  $(T, \mathfrak{G})$ .

Thus when we need to indicate the arguments of  $P(Y|X)$ , we write  $P(Y|X)(G, \xi)$  for  $(G, \xi) \in \mathfrak{G} \times S$ . It is known that a function  $\psi$  satisfying the conditions 1° and 2° of Definition 3 always exists whenever the value space  $(T, \mathfrak{G})$  of  $Y$  is a Borel space and in particular when  $(T, \mathfrak{G}) = (R^k, \mathfrak{B}^k)$ . For a proof of this statement see [1]. The proof of Proposition 1 below which relates the regular conditional distribution to the conditional expectation is parallel to the proof of the corresponding theorem in which these two are given as functions on  $\Omega$  rather than as functions on the value space  $S$  of  $X$ . (See for instance Proposition 4.28 in [1].) We give the proof here for the sake of completeness.

PROPOSITION 1. *Let  $f$  be a measurable transformation of  $(T, \mathfrak{G})$  into  $(R^1, \mathfrak{B}^1)$  and  $f \in L_1(T, \mathfrak{G}, P_Y)$ . If  $P(Y|X)$  exists then*

$$(2.4) \quad E(f \circ Y | X)(\xi) = \int_T f(\eta)P(Y|X)(d\eta, \xi) \quad \text{for a.e. } \xi \in (S, \mathfrak{F}, P_X) .$$

*Proof.* Consider the case where  $f = \chi_G$  for some  $G \in \mathfrak{G}$ . Then for every  $F \in \mathfrak{F}$  we have by (2.3)

$$(2.5) \quad \int_F E(f \circ Y | X)(\xi)P_X(d\xi) = \int_{X^{-1}(F)} \chi_G(Y(\omega))P(d\omega) \\ = P(Y^{-1}(G) \cap X^{-1}(F)) .$$

On the other hand, by 2° and then 1° of Definition 3,

$$\int_T f(\eta)P(Y|X)(d\eta, \xi) = \int_T \chi_G(\eta)P(Y|X)(d\eta, \xi) \\ = P(Y|X)(G, \xi) = P(Y \in G | X)(\xi)$$

so that by (2.2) we have

$$(2.6) \quad \int_F \left\{ \int_T f(\eta)P(Y|X)(d\eta, \xi) \right\} P_X(d\xi) = P(Y^{-1}(G) \cap X^{-1}(F)) .$$

Thus the left side of (2.5) is equal to that of (2.6) for every  $F \in \mathfrak{F}$  so that (2.4) holds in this case.

Now that (2.4) holds when  $f$  is the characteristic function of a member of  $\mathfrak{G}$  we can follow the usual procedure in integration theory to show that (2.4) holds for nonnegative simple functions on  $T$ , nonnegative  $\mathfrak{G}$ -measurable function on  $T$  and finally real valued  $\mathfrak{G}$ -measurable functions on  $T$ . Since  $f \in L_1(T, \mathfrak{G}, P_Y)$ , both sides of (2.4) always exist and are finite. In passing from nonnegative simple functions on  $T$  to nonnegative  $\mathfrak{G}$ -measurable function on  $T$  we use the Monotone Convergence Theorem for the conditional expectation which states that if  $\{Z_n, n = 1, 2, \dots\} \subset L_1(\Omega, \mathfrak{B}, P)$  and  $Z_n(\omega) \uparrow Z_0(\omega)$  for a.e.

$\omega \in (\Omega, \mathfrak{B}, P)$  then  $E(Z_n|X)(\xi) \uparrow E(Z_0|X)(\xi)$  for a.e.  $\xi \in (S, \mathfrak{F}, P_X)$  and which can be proved readily.

**PROPOSITION 2.** *Let  $\sigma(\mathfrak{F} \times \mathfrak{G})$  be the  $\sigma$ -algebra of subsets of  $S \times T$  generated by the semialgebra  $\mathfrak{F} \times \mathfrak{G}$  and let  $P_{[X, Y]}$  be the probability measure on  $(S \times T, \sigma(\mathfrak{F} \times \mathfrak{G}))$  determined by the measurable transformation  $[X, Y]$  of  $(\Omega, \mathfrak{B})$  into  $(S \times T, \sigma(\mathfrak{F} \times \mathfrak{G}))$ . Let  $f$  be a measurable transformation of  $(S \times T, \sigma(\mathfrak{F} \times \mathfrak{G}))$  into  $(R^1, \mathfrak{B}^1)$ . If  $P(Y|X)$  exists then*

$$(2.7) \quad \begin{aligned} E(f \circ [X, Y]) &= \int_{S \times T} f(\xi, \eta) P_{[X, Y]}(d(\xi, \eta)) \\ &= \int_S \left\{ \int_T f(\xi, \eta) P(Y|X)(d\eta, \xi) \right\} P_X(d\xi) \end{aligned}$$

in the sense that the existence of any member in (2.7) implies that of the other and the equality of all.

*Proof.* The first equality in (2.7) is standard. Let us prove the second. Consider the case where

$$f(\xi, \eta) = \chi_{F \times G}(\xi, \eta) = \chi_F(\xi) \chi_G(\eta) \quad \text{for } (\xi, \eta) \in S \times T$$

where  $F \in \mathfrak{F}$  and  $G \in \mathfrak{G}$ . Then by 2° and 1° of Definition 3 and by (2.2)

$$\begin{aligned} &\int_S \left\{ \int_T f(\xi, \eta) P(Y|X)(d\eta, \xi) \right\} P_X(d\xi) \\ &= \int_S \chi_F(\xi) \left\{ \int_T \chi_G(\eta) P(Y|X)(d\eta, \xi) \right\} P_X(d\xi) \\ &= \int_S \chi_F(\xi) P(Y|X)(G, \xi) P_X(d\xi) \\ &= \int_F P(Y \in G | X)(\xi) P_X(d\xi) \\ &= P(Y^{-1}(G) \cap X^{-1}(F)) \end{aligned}$$

while

$$\begin{aligned} &\int_{S \times T} f(\xi, \eta) P_{[X, Y]}(d(\xi, \eta)) \\ &= \int_{S \times T} \chi_{F \times G}(\xi, \eta) P_{[X, Y]}(d(\xi, \eta)) \\ &= P_{[X, Y]}(F \times G) \\ &= P\{\omega \in \Omega; X(\omega) \in F \text{ and } Y(\omega) \in G\} \\ &= P(Y^{-1}(G) \cap X^{-1}(F)) \end{aligned}$$

so that the second equality in (2.7) holds for this particular case.

We then proceed as in the proof of the Fubini Theorem to an arbitrary real valued  $\sigma(\mathfrak{F} \times \mathfrak{G})$ -measurable function  $f$  on  $S \times T$  to complete the proof.

**PROPOSITION 3.** *Let  $Z$  be a real valued random variable on  $(\Omega, \mathfrak{B}, P)$  with  $E(|Z|) < \infty$  and let  $g$  be a measurable transformation of  $(S, \mathfrak{F})$  into  $(R^1, \mathfrak{B}^1)$ . Then*

$$(2.8) \quad E[(g \circ X)Z] = \int_S g(\xi)E(Z|X)(\xi)P_X(d\xi)$$

*in the sense that the existence of one side implies that of the other and the equality of the two.*

*Proof.* Let us define a set function  $\Phi$  on  $\mathfrak{B}$  by

$$\Phi(B) = \int_B Z(\omega)P(d\omega) \quad \text{for } B \in \mathfrak{B}.$$

Since  $E(|Z|) < \infty$ ,  $\Phi$  is a finite signed measure on  $(\Omega, \mathfrak{B})$  which is absolutely continuous with respect to  $P$  and has  $Z$  as its Radon-Nikodym derivative with respect to  $P$ . Thus for the real valued random variables  $g \circ X$  and  $Z$  on  $(\Omega, \mathfrak{B}, P)$  we have

$$E[(g \circ X)Z] = \int_{\Omega} g(X(\omega))Z(\omega)P(d\omega) = \int_{\Omega} g(X(\omega))\Phi(d\omega)$$

in the sense that the existence of one member implies that of the others and the equality of all. Then, to prove (2.8) it suffices to show that

$$(2.9) \quad \int_{\Omega} g(X(\omega))\Phi(d\omega) = \int_S g(\xi)E(Z|X)(\xi)P_X(d\xi)$$

in the sense that the existence of one side implies that of the other and the equality of the two.

Let us consider the case where  $g = \chi_F$  for some  $F \in \mathfrak{F}$ . Then

$$\begin{aligned} \int_{\Omega} g(X(\omega))\Phi(d\omega) &= \int_{\Omega} \chi_F(X(\omega))\Phi(d\omega) = \int_{X^{-1}(F)} \Phi(d\omega) \\ &= \int_{X^{-1}(F)} Z(\omega)P(d\omega) = \int_F E(Z|X)(\xi)P_X(d\xi) \\ &= \int_S g(\xi)E(Z|X)(\xi)P_X(d\xi) \end{aligned}$$

by (2.3) so that (2.9) holds. Following the standard procedure in integration theory we proceed from this particular case to nonnegative simple functions on  $S$ , nonnegative  $\mathfrak{F}$ -measurable functions on  $S$  and finally real valued  $\mathfrak{F}$ -measurable functions on  $S$  to complete

the proof.

**3. Inversion of conditional expectations.** It is well-known that if the characteristic function  $\varphi$  of a distribution function  $F$  on  $R^1$  is  $m_L$ -integrable on  $R^1$ , then  $F$  is absolutely continuous and

$$F'(\xi) = \frac{1}{2\pi} \int_{R^1} e^{-i(\xi, \eta)} \varphi(\eta) m_L(d\eta) \quad \text{for a.e. } \xi \in (R^1, \mathfrak{B}^1, m_L).$$

Let  $\Phi$  be a finite measure on  $(R^k, \mathfrak{B}^k)$  and let  $\varphi$  be its characteristic function, i.e.,

$$(3.1) \quad \varphi(\eta) = \int_{R^k} e^{i(\xi, \eta)} \Phi(d\xi) \quad \text{for } \eta \in R^k$$

where  $(\xi, \eta) = \sum_{j=1}^k \xi_j \eta_j$ . According to the Lévy-Haviland Inversion Theorem (see [3] and [4])

$$(3.2) \quad \int_{R^k} \prod_{j=1}^k \tilde{\chi}_{a_j, b_j}(\xi_j) \Phi(d\xi) \\ = \lim_{h \rightarrow \infty} \frac{1}{(2\pi)^k} \int_{C_h} \prod_{j=1}^k \frac{e^{-ib_j \eta_j} - e^{-ia_j \eta_j}}{-i\eta_j} \varphi(\eta) m_L(d\eta)$$

for any  $a_j, b_j \in R^1$ ,  $a_j < b_j$ ,  $j = 1, 2, \dots, k$ , where

$$(3.3) \quad C_h = (-h, h) \times \dots \times (-h, h) \subset R^k \quad \text{with } h > 0,$$

and the modified characteristic function  $\tilde{\chi}_{a_j, b_j}$  is defined by

$$(3.4) \quad \tilde{\chi}_{a_j, b_j}(\eta_j) = \begin{cases} 1 & \text{for } \eta_j \in (a_j, b_j) \\ 0 & \text{for } \eta_j \in [a_j, b_j]^c \\ \frac{1}{2} & \text{for } \eta_j = a_j \text{ and for } \eta_j = b_j. \end{cases}$$

From (3.2) we derive the following:

**THEOREM 1.** *If the characteristic function  $\varphi$  of a finite measure  $\Phi$  on  $(R^k, \mathfrak{B}^k)$  is  $m_L$ -integrable on  $R^k$ , then  $\Phi$  is absolutely continuous with respect to  $m_L$  on  $(R^k, \mathfrak{B}^k)$  and a version of the Radon-Nikodym derivative of  $\Phi$  with respect to  $m_L$  is given by*

$$(3.5) \quad \frac{d\Phi}{dm_L}(\xi) = \frac{1}{(2\pi)^k} \int_{R^k} e^{-i(\xi, \eta)} \varphi(\eta) m_L(d\eta) \quad \text{for } \xi \in R^k.$$

*Proof.* Since the  $j$ th factor of the product in the integrand on the right side of (3.2) is a bounded continuous function of  $\eta_j \in R^1$ , if we assume the  $m_L$ -integrability of  $\varphi$  on  $R^k$  then the integrand on the right side of (3.2) is  $m_L$ -integrable on  $R^k$  so that (3.2) reduces to

$$(3.6) \quad \int_{R^k} \prod_{j=1}^k \tilde{X}_{a_j, b_j}(\xi_j) \Phi(d\xi) = \frac{1}{(2\pi)^k} \int_{R^k} \prod_{j=1}^k \frac{e^{-ib_j\eta_j} - e^{-ia_j\eta_j}}{-i\eta_j} \varphi(\eta) m_L(d\eta).$$

To show that  $\Phi$  is absolutely continuous with respect to  $m_L$  on  $(R^k, \mathfrak{B}^k)$  let  $A \in \mathfrak{B}^k$  and  $m_L(A) = 0$ . We proceed to show that  $\Phi(A) = 0$ . Let  $\mathfrak{A}$  be the algebra of subsets of  $R^k$  which are unions of finitely many disjoint half open and half closed intervals  $(a_1, b_1] \times \cdots \times (a_k, b_k]$  in  $R^k$ . Then the  $\sigma$ -algebra of subsets of  $R^k$  generated by  $\mathfrak{A}$  is precisely our  $\mathfrak{B}^k$ . Let  $\varepsilon > 0$  be arbitrarily given. Since  $\Phi$  is a finite measure on  $\mathfrak{B}^k$  and since  $m_L(A)$  is finite (in fact equal to zero),  $(\Phi + m_L)(A)$  is finite so that there exists some  $B \in \mathfrak{A}$  such that

$$(3.7) \quad (\Phi + m_L)(A \Delta B) < \varepsilon$$

where  $A \Delta B$  is the symmetric difference between  $A$  and  $B$ . Now (3.7) implies that  $\Phi(A \Delta B) < \varepsilon$  so that

$$(3.8) \quad \Phi(A) \leq \Phi(B) + \Phi(A \Delta B) < \Phi(B) + \varepsilon.$$

It also implies that  $m_L(A \Delta B) < \varepsilon$  so that in view of  $m_L(A) = 0$  we have

$$m_L(B) < \varepsilon.$$

Since  $B$  is the union of finitely many, say  $m$ , disjoint half open half closed intervals, there exist  $m$  open intervals  $B^{(n)}$ ,  $n = 1, 2, \dots, m$ , such that

$$(3.9) \quad B \subset \bigcup_{n=1}^m B^{(n)} \quad \text{and} \quad m_L(B) < \sum_{n=1}^m m_L(B^{(n)}) < \varepsilon.$$

Let each  $B^{(n)}$  be given as

$$(3.10) \quad B^{(n)} = \zeta^{(n)} + C^{(n)}$$

where

$$(3.11) \quad \zeta^{(n)} \in R^k \quad \text{and} \quad C^{(n)} = (-h_1^{(n)}, h_1^{(n)}) \times \cdots \times (-h_k^{(n)}, h_k^{(n)}) \subset R^k.$$

In view of the openness of  $C^{(n)}$  and the definition of  $\tilde{\chi}_{a_j, b_j}$  by (3.4) we have from (3.6)

$$(3.12) \quad \begin{aligned} & \Phi(\zeta^{(n)} + C^{(n)}) \\ & \leq \frac{1}{(2\pi)^k} \int_{R^k} \prod_{j=1}^k \frac{e^{-i(\zeta_j^{(n)} + h_j^{(n)})\eta_j} - e^{-i(\zeta_j^{(n)} - h_j^{(n)})\eta_j}}{-i\eta_j} \\ & = \frac{m_L(C^{(n)})}{(2\pi)^k} \int_{R^k} \prod_{j=1}^k \frac{\sin \eta_j h_j^{(n)}}{\eta_j h_j^{(n)}} e^{-i\zeta_j^{(n)}\eta_j} \varphi(\eta) m_L(d\eta). \end{aligned}$$

Since  $|(\eta_j h_j^{(n)})^{-1} \sin \eta_j h_j^{(n)}| \leq 1$  for  $\eta_j \in R^1$  and since  $m_L(C^{(n)}) = m_L(B^{(n)})$  we have

$$(3.13) \quad \Phi(B^{(n)}) = \Phi(\zeta^{(n)} + C^{(n)}) \leq \frac{m_L(B^{(n)})}{(2\pi)^k} \int_{R^k} |\varphi(\eta)| m_L(d\eta).$$

From (3.9) and (3.13) we obtain

$$(3.14) \quad \Phi(B) \leq \sum_{n=1}^m \Phi(B^{(n)}) \leq \frac{\varepsilon}{(2\pi)^k} \int_{R^k} |\varphi(\eta)| m_L(d\eta).$$

Using (3.14) in (3.8) we have

$$\Phi(A) < \varepsilon \left\{ \frac{1}{(2\pi)^k} \int_{R^k} |\varphi(\eta)| m_L(d\eta) + 1 \right\}.$$

From the arbitrariness of  $\varepsilon > 0$  we have  $\Phi(A) = 0$ . This proves the absolute continuity of  $\Phi$  with respect to  $m_L$  on  $(R^k, \mathfrak{B}^k)$ .

To obtain the Radon-Nikodym derivative of  $\Phi$  with respect to  $m_L$  on  $(R^k, \mathfrak{B}^k)$ , let us observe first that the absolute continuity of  $\Phi$  with respect to  $m_L$  implies that the  $\Phi$  measure of the boundary of the open interval  $C^{(n)}$  in (3.11) is equal to zero. Thus in (3.12) the strict equality actually holds. If we apply this improved (3.12) to  $\zeta + C_h$  where  $\zeta$  is an arbitrary point in  $R^k$  and  $C_h$  is an open interval in  $R^k$  as given by (3.3) then we have

$$(3.15) \quad \begin{aligned} \int_{\zeta + C_h} \frac{d\Phi}{dm_L}(\xi) m_L(d\xi) &= \Phi(\zeta + C_h) \\ &= \frac{m_L(C_h)}{(2\pi)^k} \int_{R^k} \prod_{j=1}^k \frac{\sin \eta_j h}{\eta_j h} e^{-i\zeta_j \eta_j} \varphi(\eta) m_L(d\eta). \end{aligned}$$

Let  $h \rightarrow 0$  on both sides of (3.15). On the one hand we have

$$(3.16) \quad \begin{aligned} \lim_{h \rightarrow 0} \frac{1}{m_L(C_h)} \int_{\zeta + C_h} \frac{d\Phi}{dm_L}(\xi) m_L(d\xi) \\ = \frac{d\Phi}{dm_L}(\zeta), \text{ for a.e. } \zeta \in (R^k, \mathfrak{B}^k, m_L) \end{aligned}$$

and on the other hand by the Dominated Convergence Theorem

$$(3.17) \quad \begin{aligned} \lim_{h \rightarrow 0} \frac{1}{(2\pi)^k} \int_{R^k} \prod_{j=1}^k \frac{\sin \eta_j h}{\eta_j h} e^{-i\zeta_j \eta_j} \varphi(\eta) m_L(d\eta) \\ = \frac{1}{(2\pi)^k} \int_{R^k} e^{-i(\zeta, \eta)} \varphi(\eta) m_L(d\eta). \end{aligned}$$

Using (3.16) and (3.17) in (3.15) we have

$$\frac{d\Phi}{dm_L}(\zeta) = \frac{1}{(2\pi)^k} \int_{R^k} e^{-i(\zeta, \eta)} \varphi(\eta) m_L(d\eta) \text{ for a.e. } \zeta \in (R^k, \mathfrak{B}^k, m_L).$$

This completes the proof of the theorem.

By means of Proposition 2 and Theorem 1 our inversion theorem for conditional expectation can be derived now.

**THEOREM 2.** *Let  $Y$  be a real valued random variable on a probability space  $(\Omega, \mathfrak{B}, P)$  with  $E(|Y|) < \infty$  and let  $X$  be a  $k$ -dimensional random vector i.e., a measurable transformation of  $(\Omega, \mathfrak{B})$  into  $(R^k, \mathfrak{B}^k)$ . Assume that the probability distribution  $P_X$  of  $X$  is absolutely continuous with respect to  $m_L$  on  $(R^k, \mathfrak{B}^k)$ . If  $E[e^{i(u, X)} Y]$  is a  $m_L$ -integrable function of  $u$  on  $R^k$  then a version of the conditional expectation of  $Y$  given  $X$ ,  $E(Y|X)$ , is given by*

$$(3.18) \quad E(Y|X)(\xi) \frac{dP_X}{dm_L}(\xi) = \frac{1}{(2\pi)^k} \int_{R^k} e^{-i(u, \xi)} E[e^{i(u, X)} Y] m_L(du)$$

for  $\xi \in R^k$ .

*Proof.* Since  $Y$  is a measurable transformation of  $(\Omega, \mathfrak{B})$  into  $(R^1, \mathfrak{B}^1)$  which is a Borel space, the regular conditional distribution of  $Y$  given  $X$ ,  $P(Y|X)$ , exists. With fixed  $u \in R^k$  consider a complex valued function  $f$  on  $R^k \times R^1$  defined by

$$f(\xi, \eta) = e^{i(u, \xi)} \eta \quad \text{for } \xi \in R^k \text{ and } \eta \in R^1.$$

Applying (2.7) of Proposition 2 and (2.4) of Proposition 1 to the real and the imaginary parts of  $f$  we obtain

$$(3.19) \quad E[e^{i(u, X)} Y] = \int_{R^k} e^{i(u, \xi)} \left\{ \int_{R^1} \eta P(Y|X)(d\eta, \xi) \right\} P_X(d\xi) \\ \int_{R^k} e^{i(u, \xi)} E(Y|X)(\xi) P_X(d\xi).$$

Consider a set function  $\Phi$  defined on  $\mathfrak{B}^k$  by

$$(3.20) \quad \Phi(F) = \int_F E(Y|X)(\xi) P_X(d\xi) \quad \text{for } F \in \mathfrak{B}^k.$$

Since  $E(Y|X)$  is  $P_X$  integrable on  $R^k$ ,  $\Phi$  is a finite signed measure on  $(R^k, \mathfrak{B}^k)$  which is absolutely continuous with respect to  $P_X$  on  $(R^k, \mathfrak{B}^k)$  and has  $E(Y|X)$  as its Radon-Nikodym derivative with respect to  $P_X$ . According to (3.19),  $E[e^{i(u, X)} Y]$ ,  $u \in R^k$ , is the characteristic function of  $\Phi$ . Under the hypothesis of the theorem, this characteristic function is  $m_L$ -integrable over  $R^k$ . Applying Theorem 1 to the positive and the negative part of  $\Phi$  we obtain

$$(3.21) \quad \frac{d\Phi}{dm_L}(\xi) = \frac{1}{(2\pi)^k} \int_{R^k} e^{-i(u,\xi)} E[e^{i(u,X)} Y] m_L(du) \text{ for a.e. } \xi \in (R^k, \mathfrak{B}^k, m_L) \dots$$

Since by (3.20)

$$\frac{d\Phi}{dm_L}(\xi) = E(Y|X)(\xi) \frac{dP_X}{dm_L}(\xi) \quad \text{for a.e. } \xi \in (R^k, \mathfrak{B}^k, m_L)$$

we have (3.18).

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Received August 13, 1973.

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