

## ON JENSEN'S FUNCTIONAL EQUATION

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**Abstract.** Our aim is to give a martingale characterization to the general measurable solution of the Jensen functional equation. We show that the function  $f = (f(x), x \in R)$  is a measurable solution of Jensen's functional equation if and only if the process  $f(x + W_t)$  is a martingale for all  $x \in R$ , where  $W$  is a standard Brownian Motion.

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**1 Introduction.** Let  $W = (W_t, t \geq 0)$  be a standard Brownian Motion, defined on a complete probability space  $(\Omega, \mathcal{F}, \mathcal{P})$ . Let  $F = (\mathcal{F}_t, t \geq 0)$  be a filtration generated by the Brownian Motion  $\mathcal{F}_t^W = \sigma(W_s, s \leq t)$ .

We consider the Jensen functional equation

$$f\left(\frac{x+y}{2}\right) = \frac{f(x) + f(y)}{2}, \quad x, y \in R. \quad (1)$$

It is well known that the general measurable solution of (1) is linear (see e.g. [1]). In the following theorem, which is our main result, we give an equivalent characterization of the general measurable solution of (1) in terms of martingale transformation of the Brownian motion.

**Theorem.** *The function  $f = (f(x), x \in R)$  is a measurable solution of equation (1) if and only if the process  $f(x + W_t)$  is a martingale for all  $x \in R$ .*

**2 The proof of the main result.** We shall often use the equality

$$f\left(\frac{x}{2}\right) = \frac{f(x) + f(0)}{2}, \quad (2)$$

which follows from (1) taking  $y = 0$ . Before we start off with the proof of the main result let us bring the next

**Lemma.** *Let  $\xi$  be a random variable having normal distribution with  $E\xi = 0$ . If  $f(x)$  is a measurable solution of Jensen's functional equation (1) then  $f(\xi)$  is also normally distributed and*

$$Ef(\xi) = f(0).$$

*Proof.* To show that  $f(\xi)$  is normally distributed, as in [3], we use the Bernstein theorem. Let  $\xi$  and  $\eta$  be independent random variables, having normal distribution with zero mean and equal variances. Let

$$X = f(\xi) \quad \text{and} \quad Y = f(\eta).$$

It follows from (1) and (2) that

$$\frac{X + Y}{2} = \frac{f(\xi) + f(\eta)}{2} = f\left(\frac{\xi + \eta}{2}\right) = \frac{f(\xi + \eta) + f(0)}{2}$$

and hence

$$X + Y = f(\xi + \eta) + f(0). \quad (3)$$

Putting  $x = \xi - \eta$  and  $y = \eta$  in (1)

$$\frac{f(\xi - \eta) + f(\eta)}{2} = f\left(\frac{\xi}{2}\right) = \frac{f(\xi) + f(0)}{2},$$

which implies that

$$X - Y = f(\xi - \eta) - f(0). \quad (4)$$

Since  $\xi + \eta$  and  $\xi - \eta$  are independent random variables, it follows from (3) and (4) that the random variables  $X + Y$  and  $X - Y$  are also independent. Therefore, by Bernstein's theorem [2] we conclude that  $f(\xi)$  is normally distributed. This implies that

$$E|f(\xi)| < \infty.$$

Now in equation (1) we can put  $x = \xi$  and  $y = -\xi$ , which gives us

$$f(\xi) + f(-\xi) = 2f(0).$$

Taking expectations in the last equation on both sides and considering the fact that  $\xi$  has symmetrical distribution we get

$$Ef(\xi) = f(0).$$

□

**The proof of the Theorem.** Assume that the function  $f = (f(x), x \in R)$  is a measurable solution of equation (1). We need to show that  $f(W_t)$  is a martingale. By the Lemma  $f(W_t)$  is integrable for any  $t \geq 0$  and we have to show the martingale equality

$$E(f(W_t) | \mathcal{F}_s) = f(W_s) \quad \text{a.s.} \quad (5)$$

Taking  $x = W_t - W_s$  and  $y = W_s$  in equation (1) and using also equation (2) for  $x = W_t$  we have

$$f\left(\frac{W_t}{2}\right) = \frac{f(W_t - W_s) + f(W_s)}{2} = \frac{f(W_t) + f(0)}{2},$$

from where we obtain

$$f(W_t) = f(W_t - W_s) + f(W_s) - f(0).$$

If we take conditional expectations in the last equation we get

$$E(f(W_t) | \mathcal{F}_s) = E(f(W_t - W_s) + f(W_s) - f(0) | \mathcal{F}_s) \quad a.s.$$

As long as  $W_t - W_s$  is independent of  $\mathcal{F}_s$  filtration and  $W_s$  is  $\mathcal{F}_s$  measurable, by the properties of conditional expectation we will have

$$E(f(W_t) | \mathcal{F}_s) = Ef(W_t - W_s) + f(W_s) - f(0) \quad a.s. \quad (6)$$

Since by the Lemma  $Ef(W_t - W_s) = f(0)$ , the martingale equality (5) follows from equation (6). Now it is easy to show that  $f(x + W_t)$  is also a martingale. From (1) and (2) we will have

$$f\left(\frac{x + W_t}{2}\right) = \frac{f(x) + f(W_t)}{2} = \frac{f(x + W_t) + f(0)}{2}, \quad (7)$$

which implies that from (7) we have

$$f(x + W_t) = f(x) + f(W_t) - f(0),$$

from where it is evident that  $f(x + W_t)$  is also a martingale.

Now, if we assume that  $f(x + W_t)$  is a martingale for every  $x \in R$ , then Theorem A1 from [4] implies that the function  $f$  is of the form

$$f(x) = ax + b, \quad a, b \in R, \quad (8)$$

which satisfies equation (1). □

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