

# Brownian Motion Approximation by Neural Networks

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## Abstract

The first author recently derived several approximation results by neural network operators see the monograph [16]. There, the activation functions are induced by the arc-tangent, algebraic, Gudermannian and generalized symmetrical sigmoid functions. The results we apply here are univariate on a compact interval, regular and fractional. The outcome is the quantitative approximation of Brownian motion over the two dimensional sphere. We derive several Jackson type inequalities estimating the degree of convergence of our neural network operators to a general expectation function of Brownian motion. We give a detailed list of approximation applications regarding the expectation of well known functions of Brownian motion. Smoothness of our functions is taken into account producing higher speeds of approximation.

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## 1 Introduction

The first author in [1] and [2], see chapter 2 – 5, was the first to establish neural network approximation to continuous functions with rated by very specifically defined neural network operators of Cardaliagnet-Euvrard and 'Squasing' types, by employing the modulus of continuity of the engaged function or its high order derivative, and producing very tight Jacson type inequalities. He treats there both the univariate and multivariate cases. The defining these operators 'bell-shaped' and 'squashing' functions are assumed to be compact support. Also the first author inspired by [17], continued his studies on neural networks approximation

by introducing and using the proper quasi-interpolation operators of sigmoidal and hyperbolic tangent type which resulted into [4],[5],[6],[7],[9], by treating both the univariate and multivariate cases. He did also the corresponding fractional cases [8], [10],[12].

In [16] the first author continued similar studies for Banach space valued functions for activation functions deriving from the arctangent, algebraic, Gudermanian and generalized symmetrical sigmoid functions. The authors based and inspired by [22] perform here neural network quantitative approximations to Brownian motion over the two dimensional sphere.

They present a series of Jackson type inequalities estimating the error of approximation to a general expectation function of the Brownian motion and its derivative. They produce regular and fractional calculus results. They finish with a lot of important applications.

## 2 About Neural Network Operators

### 2.1 About the arctangent activation function neural networks

We consider the function

$$\arctan(x) = \int_0^x \frac{dz}{1+z^2}, x \in \mathbb{R}. \quad (1)$$

We will be using the function

$$h(x) := \frac{2}{\pi} \arctan\left(\frac{\pi}{2}x\right) = \frac{2}{\pi} \int_0^{\frac{\pi x}{2}} \frac{dz}{1+z^2}, x \in \mathbb{R}. \quad (2)$$

which is a sigmoid type function and it is strictly increasing. We have that

$$h(0) = 0, h(-x) = -h(x), h(+\infty) = 1, h(-\infty) = -1$$

and

$$h'(x) = \frac{4}{4 + \pi^2 x^2}, \text{ for all } x \in \mathbb{R}. \quad (3)$$

We consider the activation function

$$\psi_1(x) = \frac{1}{4} (h(x+1) - h(x-1)), x \in \mathbb{R}, \quad (4)$$

and we notice that

$$\psi_1(x) = \psi_1(-x), \quad (5)$$

i.e. it is an even function.

Since  $x+1 > x-1$ , then  $h(x+1) > h(x-1)$ , and  $\psi_1(x) > 0$ , for all  $x \in \mathbb{R}$ .

We see that

$$\psi_1(0) = \frac{1}{\pi} \arctan\left(\frac{\pi}{2}\right) \cong 0.319. \quad (6)$$

Let  $x > 0$ , we have that

$$\begin{aligned} \psi_1'(x) &= \frac{1}{4} (h'(x+1) - h'(x-1)) = \\ &= \frac{-4\pi^2 x}{(4 + \pi^2 (x+1)^2)(4 + \pi^2 (x-1)^2)} < 0 \end{aligned} \quad (7)$$

That is

$$\psi_1'(x) < 0, \text{ for } x > 0 \quad (8)$$

That is  $\psi_1$  is strictly decreasing on  $[0, +\infty)$ , strictly increasing on  $(-\infty, 0]$  and  $\psi_1'(0) = 0$ .  
Observe that

$$\lim_{x \rightarrow +\infty} \psi_1(x) = \frac{1}{4} (h(+\infty) - h(+\infty)) = 0,$$

and

$$\lim_{x \rightarrow -\infty} \psi_1(x) = \frac{1}{4} (h(-\infty) - h(-\infty)) = 0$$
(9)

That is the  $x$ -axis is the horizontal asymptote on  $\psi_1$ .

All in all,  $\psi_1$  is a bell symmetric function with maximum  $\psi_1(0) \cong 0.319$ . We need

**Theorem 1.** ([11], p. 286) *We have that*

$$\sum_{i=-\infty}^{+\infty} \psi_1(x - i) = 1, \text{ for every } x \in \mathbb{R}$$
(10)

**Theorem 2.** ([11], p. 287) *It holds*

$$\int_{-\infty}^{+\infty} \psi_1(x) dx = 1.$$
(11)

So that  $\psi_1(x)$  is a density function on  $\mathbb{R}$ .

**Definition 1.** *Let  $f \in C([a, b])$ , we call the first modulus of continuity of  $f$  at  $\delta > 0$  the following*

$$\omega_1(f, \delta) = \sup_{x, y \in [a, b]: |x - y| \leq \delta} |f(x) - f(y)|.$$

*We have that*

$$\omega_1(f, \delta) \longrightarrow 0 \text{ if and only if } \delta \longrightarrow 0$$

Denote by  $[\cdot]$  the integral part of the number and by  $\lceil \cdot \rceil$  the ceiling of the number. We mention

**Definition 2.** ([11]) *Let  $f \in C([a, b])$  and  $n \in \mathbb{N} : \lceil na \rceil \leq \lfloor nb \rfloor$ . We define the real positive linear network operator*

$${}_1A_n(f, x) := \frac{\sum_{k=\lceil na \rceil}^{\lfloor nb \rfloor} f\left(\frac{k}{n}\right) \psi_1(nx - k)}{\sum_{k=\lceil na \rceil}^{\lfloor nb \rfloor} \psi_1(nx - k)}, x \in [a, b].$$
(12)

Clearly here  ${}_1A_n(f, x) \in C([a, b])$ . In [11] we studied the pointwise and uniform convergence of  ${}_1A_n(f, x)$  to  $f(x)$  with rates.

We also mention

**Theorem 3.** ([11]) *Let  $f \in C([a, b])$ ,  $0 < \alpha < 1$ ,  $n \in \mathbb{N} : n^{1-\alpha} > 2$ ,  $x \in [a, b]$ .*

*Then*

1. 
$$|{}_1A_n(f, x) - f(x)| \leq 4.9737 \left[ \omega_1\left(f, \frac{1}{n^\alpha}\right) + \frac{4 \|f\|_\infty}{\pi^2 (n^{1-\alpha} - 2)} \right] =: \rho_1(f),$$
(13)

*and*

2.

$$\|{}_1A_n(f) - f\|_\infty \leq \rho_1(f). \quad (14)$$

We notice that

$$\lim_{n \rightarrow \infty} {}_1A_n(f) = f,$$

pointwise and uniformly. We need

**Definition 3.** Let  $\nu \geq 0, n = \lceil \nu \rceil, f \in AC^n([a, b])$  (space of functions  $f$  with  $f^{(n-1)} \in AC([a, b])$ , absolutely continuous functions). We call left Caputo functional derivative (see [18], pp. 49-52, [21], [23]) the function

$$D_{*\alpha}^\nu f(x) = \frac{1}{\Gamma(n - \nu)} \int_\alpha^x (x - t)^{n-\nu-1} f^{(n)}(t) dt \quad (15)$$

for every  $x \in [a, b]$ , where  $\Gamma$  is the gamma function  $\Gamma(\nu) = \int_0^\infty e^{-t} t^{\nu-1} dt, \nu > 0$ .

Notice  $D_{*\alpha}^\nu f \in L_1([a, b])$  and  $D_{*\alpha}^\nu f(x)$  exists a.e. on  $[a, b]$ .

We set  $D_{*\alpha}^0 f(x) = f(x)$  for every  $x \in [a, b]$ .

**Definition 4.** (see also [3], [20], [21]) Let  $f \in AC^m([a, b]), m = \lceil \beta \rceil, \beta > 0$ .

The right Caputo functional derivative of order  $\beta > 0$  is given by

$$D_{b-}^\beta f(x) = \frac{(-1)^m}{\Gamma(m - \beta)} \int_x^b (\zeta - x)^{m-\beta-1} f^{(m)}(\zeta) d\zeta, \quad (16)$$

for every  $x \in [a, b]$ . We set  $D_{b-}^0 f(x) = f(x)$ . Notice that  $D_{b-}^\beta f \in L_1([a, b])$  and  $D_{b-}^\beta f$  exists a.e. on  $[a, b]$ .

We also mention the real valued fractional approximation result by neural networks.

**Theorem 4.** ([11]) Let  $0 < \alpha < 1, f \in C^1([a, b]), 0 < \beta < 1, x \in [a, b], n \in \mathbb{N} : n^{1-\beta} > 2$ . Then

$$\begin{aligned} |{}_1A_n(f, x) - f(x)| &\leq \frac{4.9737}{\Gamma(\alpha + 1)} \\ &\left\{ \frac{\omega_1 (D_{x-}^\alpha f, \frac{1}{n^\beta})_{[a, x]} + \omega_1 (D_{*x}^\alpha f, \frac{1}{n^\beta})_{[x, b]}}{n^{\alpha\beta}} + \right. \\ &\left. + \frac{2}{\pi^2 (n^{1-\beta} - 2)} \left( \|D_{x-}^\alpha f\|_{\infty, [a, x]} (x - a)^\alpha + \|D_{*x}^\alpha f\|_{\infty, [x, b]} (b - x)^\alpha \right) \right\}. \quad (17) \end{aligned}$$

As we see here we obtain the real valued fractionally type pointwise convergence with rates of  ${}_1A_n \rightarrow I$  the unit operator as  $n \rightarrow \infty$ .

## 2.2 About the algebraic activation function neural networks

Here see also [13].

We consider the generator algebraic function

$$\varphi(x) = \frac{x}{\sqrt[2m]{1 + x^{2m}}}, m \in \mathbb{N}, x \in \mathbb{R}, \quad (18)$$

which is a sigmoidal type of function and is a strictly increasing function. We see that  $\varphi(-x) = -\varphi(x)$  with  $\varphi(0) = 0$ . We get that

$$\varphi'(x) = \frac{1}{(1+x^{2m})^{\frac{2m+1}{2m}}} > 0, \text{ for every } x \in \mathbb{R}, \quad (19)$$

proving  $\varphi$  as strictly increasing over  $\mathbb{R}$ ,  $\varphi'(-x) = \varphi'(x)$ . We easily find that

$$\lim_{x \rightarrow +\infty} \varphi(x) = 1, \varphi(+\infty) = 1, \text{ and } \lim_{x \rightarrow -\infty} \varphi(x) = -1, \varphi(-\infty) = -1.$$

We consider the activation function

$$\psi_2(x) = \frac{1}{4} [\varphi(x+1) - \varphi(x-1)]. \quad (20)$$

Clearly it is  $\psi_2(x) = \psi_2(-x)$  for every  $x \in \mathbb{R}$ , so that  $\psi_2$  is an even function and symmetric with respect to the  $y$ -axis. Clearly  $\psi_2(x) > 0$ , for every  $x \in \mathbb{R}$ . Also it is

$$\psi_2(0) = \frac{1}{2^{2m\sqrt{2}}}, m \in \mathbb{N}. \quad (21)$$

By [13] we have that  $\psi_2'(x) < 0$ , for  $x > 0$ . That is  $\psi_2$  is strictly decreasing over  $[0, +\infty)$ . Clearly  $\psi_2$  is strictly increasing over  $(-\infty, 0]$  and  $\psi_2'(0) = 0$ .

Furthermore we obtain that

$$\lim_{x \rightarrow +\infty} \psi_1(x) = \frac{1}{4} (\varphi(+\infty) - \varphi(+\infty)) = 0, \quad (22)$$

and

$$\lim_{x \rightarrow -\infty} \psi_2(x) = \frac{1}{4} (\varphi(-\infty) - \varphi(-\infty)) = 0 \quad (23)$$

That is the  $x$ -axis is the horizontal asymptote on  $\psi_2$ .

Conclusion,  $\psi_2$  is a bell shape symmetric function with maximum

$$\psi_2(0) = \frac{1}{2^{2m\sqrt{2}}}, m \in \mathbb{N}.$$

We need

**Theorem 5.** ([13]) *We have that*

$$\sum_{i=-\infty}^{+\infty} \psi_2(x-i) = 1, \text{ for every } x \in \mathbb{R}. \quad (24)$$

**Theorem 6.** ([13], p. 287) *It holds*

$$\int_{-\infty}^{+\infty} \psi_2(x) dx = 1. \quad (25)$$

So  $\psi_2$  is a density function.

We mention

**Definition 5.** ([13]) *Let  $f \in C([a, b])$  and  $n \in \mathbb{N} : [na] \leq [nb]$ . We introduce and define the real positive valued linear network operator*

$${}_2A_n(f, x) := \frac{\sum_{k=[na]}^{[nb]} f\left(\frac{k}{n}\right) \psi_2(nx-k)}{\sum_{k=[na]}^{[nb]} \psi_2(nx-k)}, x \in [a, b]. \quad (26)$$

Clearly here  ${}_2A_n(f, x) \in C([a, b])$ .

We mention here about the pointwise and uniform convergence of  ${}_2A_n(f, x)$  to  $f(x)$  with rates.

**Theorem 7.** ([13]) Let  $f \in C([a, b])$ ,  $0 < \alpha < 1$ ,  $n \in \mathbb{N} : n^{1-\alpha} > 2$ ,  $x \in [a, b]$ ,  $m \in \mathbb{N}$ . Then

1.

$$|{}_2A_n(f, x) - f(x)| \leq 2 \left( \sqrt[2m]{1+4m} \right) \left[ \omega_1 \left( f, \frac{1}{n^\alpha} \right) + \frac{\|f\|_\infty}{2m(n^{1-\alpha} - 2)^{2m}} \right] =: \rho_2(f), \quad (27)$$

and

2.

$$\|{}_2A_n(f) - f\|_\infty \leq \rho_2(f). \quad (28)$$

We get that  $\lim_{n \rightarrow +\infty} {}_2A_n(f) = f$ , pointwise and uniformly.

**Theorem 8.** ([13])

Let  $0 < \alpha < 1$ ,  $f \in C^1([a, b])$ ,  $0 < \beta < 1$ ,  $m \in \mathbb{N}$ ,  $x \in [a, b]$ ,  $n \in \mathbb{N} : n^{1-\beta} > 2$ . Then

$$\begin{aligned} |{}_2A_n(f, x) - f(x)| &\leq \frac{2 \sqrt[2m]{1+4m}}{\Gamma(\alpha+1)} \\ &\left\{ \frac{\omega_1 \left( D_{x^-}^\alpha f, \frac{1}{n^\beta} \right)_{[a,x]} + \omega_1 \left( D_{*x}^\alpha f, \frac{1}{n^\beta} \right)_{[x,b]}}{n^{\alpha\beta}} + \right. \\ &\left. + \frac{1}{4m(n^{1-\beta} - 2)^{2m}} \left( \|D_{x^-}^\alpha f\|_{\infty, [a,x]} (x-a)^\alpha + \|D_{*x}^\alpha f\|_{\infty, [x,b]} (b-x)^\alpha \right) \right\}. \quad (29) \end{aligned}$$

As we see here we obtain the real valued functionally type pointwise convergence with rates of  ${}_2A_n \rightarrow I$  the unit operator as  $n \rightarrow \infty$ .

### 2.3 About the Gudermannian activation function neural networks

See also [14], [24].

Here we consider  $gd(x)$  the Gudermannian function [14], which is a sigmoid function, as a generator function:

$$\sigma(x) = 2 \arctan \left( \tanh \left( \frac{x}{2} \right) \right) = \int_0^x \frac{dt}{\cosh t} =: gd(x), \quad x \in \mathbb{R}. \quad (30)$$

Let the normalized generator sigmoid function

$$f(x) := \frac{2}{\pi} \sigma(x) = \frac{2}{\pi} \int_0^x \frac{dt}{\cosh t} = \frac{4}{\pi} \int_0^x \frac{1}{e^t + e^{-t}} dt, \quad x \in \mathbb{R}. \quad (31)$$

Here

$$f'(x) = \frac{2}{\pi \cosh(x)} > 0, \quad \text{for every } x \in \mathbb{R},$$

hence  $f$  is strictly increasing on  $\mathbb{R}$ .

Notice that  $\tanh(-x) = -\tanh(x)$  and  $\arctan(-x) = -\arctan(x)$ ,  $x \in \mathbb{R}$ .

So, here the neural network activation function will be:

$$\psi_3(x) = \frac{1}{4} [f(x+1) - f(x-1)], x \in \mathbb{R}. \quad (32)$$

By [14], we get that

$$\psi_3(-x) = \psi_3(x) \text{ for every } x \in \mathbb{R}, \quad (33)$$

i.e. it is even and symmetric with respect to the  $y$ -axis. Here we have  $f(+\infty) = 1$ ,  $f(-\infty) = 1$  and  $f(0) = 0$ . Clearly it is

$$f(-x) = -f(x) \text{ for every } x \in \mathbb{R}, \quad (34)$$

an odd function, symmetric with respect to the origin. Since  $x+1 > x-1$ , and  $f(x+1) > f(x-1)$ , we obtain  $\psi_3(x) > 0$ , for all  $x \in \mathbb{R}$ .

By [14], we have that

$$\psi_3(0) = \frac{1}{\pi} g d(1) \cong 0.2757. \quad (35)$$

By [14]  $\psi_3$  is strictly decreasing on  $[0, +\infty)$ , strictly increasing on  $(-\infty, 0]$  and  $\psi_3'(0) = 0$ .

Also we have that

$$\lim_{x \rightarrow +\infty} \psi_3(x) = \lim_{x \rightarrow -\infty} \psi_3(x) = 0, \quad (36)$$

That is the  $x$ -axis is the horizontal asymptote on  $\psi_3$ .

Conclusion,  $\psi_3$  is a bell symmetric function with maximum  $\psi_3(0) \cong 0.2757$ .

We mention

**Theorem 9.** ([14]) *It holds that*

$$\sum_{i=-\infty}^{+\infty} \psi_3(x-i) = 1, \text{ for every } x \in \mathbb{R} \quad (37)$$

**Theorem 10.** ([14]) *We have that*

$$\int_{-\infty}^{+\infty} \psi_3(x) dx = 1. \quad (38)$$

So  $\psi_3(x)$  is a density function.

We make

**Definition 6.** ([14]) *Let  $f \in C([a, b])$  and  $n \in \mathbb{N} : [na] \leq [nb]$ . We define the real positive valued linear network operator*

$${}_3A_n(f, x) := \frac{\sum_{k=[na]}^{[nb]} f\left(\frac{k}{n}\right) \psi_3(nx-k)}{\sum_{k=[na]}^{[nb]} \psi_3(nx-k)}, x \in [a, b]. \quad (39)$$

Clearly here  ${}_3A_n(f, x) \in C([a, b])$ .

We mention here about the pointwise and uniform convergence of  ${}_3A_n(f, x)$  to  $f(x)$  with rates.

**Theorem 11.** ([14]) Let  $f \in C([a, b])$ ,  $0 < \alpha < 1$ ,  $n \in \mathbb{N} : n^{1-\alpha} > 2$ ,  $x \in [a, b]$ .

Then

1.

$$|{}_3A_n(f, x) - f(x)| \leq 4.824 \left[ \omega_1 \left( f, \frac{1}{n^\alpha} \right) + \frac{4 \|f\|_\infty}{\pi e^{(n^{1-\alpha}-2)}} \right] =: \rho_3(f), \quad (40)$$

and

2.

$$\|{}_3A_n(f) - f\|_\infty \leq \rho_3(f). \quad (41)$$

We get that  $\lim_{n \rightarrow +\infty} {}_3A_n(f) = f$ , pointwise and uniformly.

We also mention the following real valued functional approximation result by neural networks.

**Theorem 12.** ([14])

Let  $0 < \alpha < 1$ ,  $f \in C^1([a, b])$ ,  $0 < \beta < 1$ ,  $x \in [a, b]$ ,  $n \in \mathbb{N} : n^{1-\beta} > 2$ . Then

$$\begin{aligned} |{}_3A_n(f, x) - f(x)| &\leq \frac{4.824}{\Gamma(\alpha + 1)} \\ &\left\{ \frac{\omega_1(D_{x^-}^\alpha f, \frac{1}{n^\beta})_{[a,x]} + \omega_1(D_{*x}^\alpha f, \frac{1}{n^\beta})_{[x,b]}}{n^{\alpha\beta}} + \right. \\ &\left. + \frac{2}{\pi e^{(n^{1-\beta}-2)}} \left( \|D_{x^-}^\alpha f\|_{\infty, [a,x]} (x-a)^\alpha + \|D_{*x}^\alpha f\|_{\infty, [x,b]} (b-x)^\alpha \right) \right\}. \end{aligned} \quad (42)$$

As we see here we obtain the real valued functionally type pointwise convergence with rates of  ${}_3A_n \rightarrow I$  the unit operator as  $n \rightarrow \infty$ .

## 2.4 About the generalized symmetrical activation function neural networks

Here we consider the generalized symmetrical sigmoid function ([15], [19])

$$f_1(x) = \frac{x}{(1 + |x|^\mu)^{\frac{1}{\mu}}}, \mu > 0, x \in \mathbb{R}. \quad (43)$$

This has applications in immunology and protection from disease together with probability theory. It is also called a symmetrical protection curve. The parameter  $\mu$  is a shape parameter controlling how fast the curve approaches the asymptotes for a given slope at the inflection point. When  $\mu = 1$   $f_1$  is the absolute sigmoid function, and when  $\mu = 2$   $f_1$  is the square root sigmoid function. When  $\mu = 1.5$  the function approximates the arctangent function, when  $\mu = 2.9$  it approximates the logistic function, and when  $\mu = 3.4$  it approximates the error function. Parameter  $\mu$  is estimated in the likelihood maximization ([19]) For more see [19].

Next we study the particular generator sigmoid function

$$f_2(x) = \frac{x}{(1 + |x|^\lambda)^{\frac{1}{\lambda}}}, \lambda \text{ is an odd number, } x \in \mathbb{R} \quad (44)$$

We have that  $f_2(0) = 0$ , and

$$f_2(-x) = -f_2(x). \quad (45)$$



So  $f_2$  is symmetric with respect to zero.

When  $x \geq 0$ , we get that ([15])

$$f_2'(x) = \frac{1}{(1+x^\lambda)^{\frac{\lambda+1}{\lambda}}} > 0, \quad (46)$$

that is  $f_2$  is strictly increasing on  $[0, +\infty)$ , strictly increasing on  $(-\infty, 0]$ . Hence  $f_2$  is strictly increasing on  $\mathbb{R}$ .

We have that  $f_2(+\infty) = f_2(-\infty) = 1$ .

Let us consider the activation function ([15]):

$$\begin{aligned} \psi_4(x) &= \frac{1}{4} [f_2(x+1) - f_2(x-1)] = \\ \psi_4(x) &= \frac{1}{4} \left[ \frac{x+1}{(1+|x+1|^\lambda)^{\frac{1}{\lambda}}} - \frac{x-1}{(1+|x-1|^\lambda)^{\frac{1}{\lambda}}} \right]. \end{aligned} \quad (47)$$

Clearly it holds ([15])

$$\psi_4(-x) = \psi_4(x), \text{ for every } x \in \mathbb{R}, \quad (48)$$

$$\psi_4(0) = \frac{1}{2\sqrt[2]{2}}, \quad (49)$$

and  $\psi_4(x) > 0$ , for every  $x \in \mathbb{R}$ .

Following [15], we have that  $\psi_4$  is strictly decreasing on  $[0, +\infty)$ , strictly increasing on  $(-\infty, 0]$ , by  $\psi_4$ -symmetry with respect to  $y$ -axis, and  $\psi_4'(0) = 0$

Clearly it is

$$\lim_{x \rightarrow +\infty} \psi_4(x) = \lim_{x \rightarrow -\infty} \psi_4(x) = 0, \quad (50)$$

Therefore the  $x$ -axis is the horizontal asymptote on  $\psi_4(x)$ .

The value

$$\psi_4(0) = \frac{1}{2\sqrt[2]{2}}, \lambda \text{ is an odd number,}$$

is the maximum of  $\psi_4$ , which is a bell shaped function.

We need

**Theorem 13.** ([15]) *It holds that*

$$\sum_{i=-\infty}^{+\infty} \psi_4(x-i) = 1, \text{ for every } x \in \mathbb{R} \quad (51)$$

**Theorem 14.** ([15]) *We have that*

$$\int_{-\infty}^{+\infty} \psi_4(x) dx = 1. \quad (52)$$

So  $\psi_4(x)$  is a density function on  $\mathbb{R}$ .

We mention

**Definition 7.** ([15]) Let  $f \in C([a, b])$  and  $n \in \mathbb{N} : \lceil na \rceil \leq \lfloor nb \rfloor$ . We define the real positive valued linear network operator

$${}_4A_n(f, x) := \frac{\sum_{k=\lceil na \rceil}^{\lfloor nb \rfloor} f\left(\frac{k}{n}\right) \psi_4(nx - k)}{\sum_{k=\lceil na \rceil}^{\lfloor nb \rfloor} \psi_4(nx - k)}, x \in [a, b]. \quad (53)$$

Clearly here  ${}_4A_n(f, x) \in C([a, b])$ .

We mention here about the pointwise and uniform convergence of  ${}_4A_n(f, x)$  to  $f(x)$  with rates.

**Theorem 15.** ([15]) Let  $f \in C([a, b])$ ,  $0 < \alpha < 1$ ,  $n \in \mathbb{N} : n^{1-\alpha} > 2$ ,  $x \in [a, b]$ ,  $\lambda \in \mathbb{N}$  is odd. Then

1. 
$$|{}_4A_n(f, x) - f(x)| \leq 2 \sqrt[\lambda]{1 + 2^\lambda} \left[ \omega_1\left(f, \frac{1}{n^\alpha}\right) + \frac{\|f\|_\infty}{\lambda(n^{1-\alpha} - 2)^\lambda} \right] =: \rho_4(f), \quad (54)$$

and

2. 
$$\|{}_4A_n(f) - f\|_\infty \leq \rho_4(f). \quad (55)$$

We get that  $\lim_{n \rightarrow \infty} {}_4A_n(f) = f$ , pointwise and uniformly.

Next we mention the corresponding real valued fractional approximation result by neural networks.

**Theorem 16.** ([15]) Let  $0 < \alpha < 1$ ,  $f \in C^1([a, b])$ ,  $0 < \beta < 1$ ,  $\lambda$  is odd  $x \in [a, b]$ ,  $n \in \mathbb{N} : n^{1-\beta} > 2$ . Then

$$\begin{aligned} |{}_4A_n(f, x) - f(x)| &\leq \frac{2 \sqrt[\lambda]{1 + 2^\lambda}}{\Gamma(\alpha + 1)} \\ &\left\{ \frac{\omega_1(D_{x^-}^\alpha f, \frac{1}{n^\beta})_{[a, x]} + \omega_1(D_{*x}^\alpha f, \frac{1}{n^\beta})_{[x, b]}}{n^{\alpha\beta}} + \right. \\ &\left. + \frac{1}{2\lambda(n^{1-\beta} - 2)^\lambda} \left( \|D_{x^-}^\alpha f\|_{\infty, [a, x]} (x - a)^\alpha + \|D_{*x}^\alpha f\|_{\infty, [x, b]} (b - x)^\alpha \right) \right\}. \quad (56) \end{aligned}$$

As we see here we obtain real valued fractionally type pointwise convergence with rates of  ${}_4A_n \rightarrow I$  the unit operator, as  $n \rightarrow \infty$ .

We give a unified definition:

**Definition 8.** Let  $f \in C([a, b])$  and  $n \in \mathbb{N} : \lceil na \rceil \leq \lfloor nb \rfloor$ . We define the following positive linear network operators ( $j = 1, 2, 3, 4$ )

$${}_jA_n(f, x) := \frac{\sum_{k=\lceil na \rceil}^{\lfloor nb \rfloor} f\left(\frac{k}{n}\right) \psi_j(nx - k)}{\sum_{k=\lceil na \rceil}^{\lfloor nb \rfloor} \psi_j(nx - k)}, x \in [a, b]. \quad (57)$$

We make

**Remark 1.** We set

$$c_j := \begin{cases} 4.9737, & j = 1 \\ 2 \sqrt[2m]{1 + 4^m}, & j = 2, m \in \mathbb{N} \\ 4.824, & j = 3 \\ 2 \sqrt[\lambda]{1 + 2^\lambda}, & j = 4, \lambda \in \mathbb{N} \text{ is odd} \end{cases}. \quad (58)$$

Furthermore, let  $n, m \in \mathbb{N}, 0 < \alpha < 1$  and  $\lambda \in \mathbb{N}$  is odd.

We define,

$$\gamma_j(n) := \begin{cases} \frac{4}{\pi^2(n^{1-\alpha}-2)}, & j = 1 \\ \frac{1}{2m(n^{1-\alpha}-2)^{2m}}, & j = 2 \\ \frac{4}{\pi e^{(n^{1-\alpha}-2)}}, & j = 3 \\ \frac{1}{\lambda(n^{1-\alpha}-2)^\lambda}, & j = 4 \end{cases}. \quad (59)$$

We present the unified basic result

**Theorem 17.** Let  $f \in C([a, b]), 0 < \alpha < 1, n \in \mathbb{N} : n^{1-\alpha} > 2, x \in [a, b], j = 1, 2, 3, 4$ . Then

1.

$$|{}_jA_n(f, x) - f(x)| \leq c_j \left[ \omega_1 \left( f, \frac{1}{n^\alpha} \right) + \gamma_j(n) \|f\|_\infty \right] =: \rho_j(f), \quad (60)$$

and

2.

$$\|{}_jA_n(f) - f\|_\infty \leq \rho_j(f). \quad (61)$$

We get that  $\lim_{n \rightarrow \infty} {}_jA_n(f) = f$ , pointwise and uniformly.

**Proof.** By Theorems 3, 7, 11 and 15.  $\square$

We make

**Remark 2.** Let  $m, n \in \mathbb{N}, 0 < \beta < 1$  and  $\lambda \in \mathbb{N}$  is odd. We set

$$\epsilon_j(n) := \begin{cases} \frac{2}{\pi^2(n^{1-\beta}-2)}, & j = 1 \\ \frac{1}{4m(n^{1-\beta}-2)^{2m}}, & j = 2 \\ \frac{2}{\pi e^{(n^{1-\beta}-2)}}, & j = 3 \\ \frac{1}{2\lambda(n^{1-\beta}-2)^\lambda}, & j = 4. \end{cases}. \quad (62)$$

Next we present the following unified result

**Theorem 18.** Let  $0 < \alpha < 1, f \in C^1([a, b]), 0 < \beta < 1, x \in [a, b], n \in \mathbb{N} : n^{1-\beta} > 2, j = 1, 2, 3, 4$ . Then

$$|{}_jA_n(f, x) - f(x)| \leq \frac{c_j}{\Gamma(\alpha + 1)}$$

$$\left\{ \frac{\omega_1 (D_{x^-}^\alpha f, \frac{1}{n^\beta})_{[a,x]} + \omega_1 (D_{*x}^\alpha f, \frac{1}{n^\beta})_{[x,b]}}{n^{\alpha\beta}} + \right. \\ \left. + \epsilon_j(n) \left( \|D_{x^-}^\alpha f\|_{\infty,[a,x]} (x-a)^\alpha + \|D_{*x}^\alpha f\|_{\infty,[x,b]} (b-x)^\alpha \right) \right\}. \quad (63)$$

where  $c_j$  as in (58).

We have

$${}_j A_n(f, x) \rightarrow f(x), \text{ as } n \rightarrow \infty$$

**Proof.** By Theorems 4, 8, 12 and 16.  $\square$

### 3 About Brownian Motion on 2-Dimensional Sphere

#### 3.1 Describing the Brownian Motion on $S^2$

([22]) The Brownian motion on  $S^n$  is a diffusion (Markov) process  $W_t, t \geq 0$ , on  $S^n$  whose transition density is a function  $P(t, x, y)$  on  $(0, \infty) \times S^n \times S^n$  satisfying

$$\frac{\partial P}{\partial t} = \frac{1}{2} \Delta_n P, \quad (64)$$

and

$$P(t, x, y) \rightarrow \delta_x(y) \text{ as } t \rightarrow 0^+ \quad (65)$$

where  $\Delta_n$  is the Laplace-Beltrami operator of  $S^n$  acting on the x-variables and  $\delta_x(y)$  is the delta mass at  $x$ , i.e.  $P(t, x, y)$  is the **heat kernel** of  $S^n$ . The heat kernel exists, it is unique, positive, and smooth in  $(t, x, y)$ .

**Remark 3.** The heat kernel  $P(t, x, y)$  satisfying the following properties

1. *Symmetry:*  $P(t, x, y) = P(t, y, x)$
2. *The semigroup identity:* For any  $s \in (0, t)$ ,

$$P(t, x, y) = \int_{S^n} P(s, x, z) P(t-s, z, y) d\mu(z)$$

where  $d\mu$  is the area measure element of  $S^n$ .

3. For all  $t > 0$  and  $x \in S^n$

$$\int_{S^n} P(t, x, y) d\mu(y) = 1$$

4. As  $t \rightarrow \infty$ ,  $P(t, x, y)$  approaches the uniform density on  $S^n$ , i.e.

$$\lim_{t \rightarrow \infty} P(t, x, y) = \frac{1}{A_n}$$

where  $A_n$  is the area of the  $S^n$  with radius  $a$ . It is also well known that

$$A_n = \frac{2\pi^{\frac{n+1}{2}} a^n}{(\frac{n-1}{2})!} \text{ for } n \text{ odd,}$$

$$A_n = \frac{2^n (\frac{n}{2} - 1)! \pi^{\frac{n}{2}} a^n}{(n-1)!} \text{ for } n \text{ even.}$$

Finally, the symmetry of  $S^n$  implies that  $P(t, x, y)$  depends only on  $t$  and  $d(x, y)$ , the distance between  $x$  and  $y$ . Thus in spherical coordinates it depends on  $t$  and the angle  $\varphi$  between  $x$  and  $y$ . Hence

$$P(t, x, y) = p(t, \varphi)$$

where  $p(t, \varphi)$  satisfies

$$\frac{\partial p}{\partial t} = \frac{1}{2} \Delta_n p = \frac{1}{2a^2} \left[ (n-1) \cot \varphi \cdot \frac{\partial p}{\partial \varphi} + \frac{\partial^2 p}{\partial \varphi^2} \right] \quad (66)$$

and

$$\lim_{t \rightarrow 0^+} a A_{n-1} p(t, \varphi) \cdot \sin^{n-1} \varphi = \delta(\varphi) \quad (67)$$

Here  $\delta(\cdot)$  is the standard Dirac delta function on  $\mathbb{R}$ .

### 3.2 Explicit form of the heat kernel of $S^2$

Let  $W_t$ ,  $t \geq 0$  be the Brownian motion on a 2-dimensional sphere  $S^2$  of radius  $a$ . The transition density function  $p(t, \varphi)$  of  $X_t$  is the unique solution of

$$\frac{\partial p}{\partial t} = \frac{1}{2a^2 \sin \varphi} \left( \frac{\partial^2 p(t, \varphi)}{\partial \varphi^2} \sin \varphi + \frac{\partial p}{\partial \varphi} \cos \varphi \right) \quad (68)$$

and

$$\lim_{t \rightarrow 0^+} 2\pi a^2 \sin \varphi \cdot p(t, \varphi) = \delta(\varphi). \quad (69)$$

The solution to the diffusion equation

$$\frac{\partial K(t, \varphi)}{\partial t} = \frac{1}{\sin \varphi} \left( \cos \varphi \frac{\partial K(t, \varphi)}{\partial \varphi} + \sin \varphi \frac{\partial^2 K(t, \varphi)}{\partial \varphi^2} \right) \quad (70)$$

with initial condition

$$\lim_{t \rightarrow 0^+} 2\pi \sin(\varphi) K(t, \varphi) = \delta(\varphi) \quad (71)$$

is given by the function

$$K(t, \varphi) = \frac{1}{4\pi} \sum_{n \in \mathbb{N}} (2n+1) \exp\left(-n(n+1)\sqrt{2t}\right) P_n^0(\cos \varphi). \quad (72)$$

Here  $P_n^0$ ,  $n = 0, 1, 2, \dots$  is the associated Legendre polynomials of order zero, i.e.

$$P_n^0(x) = \frac{1}{2^n n!} \cdot \frac{d^n}{dx^n} [(x^2 - 1)^n] \quad (73)$$

This fact implies the following

**Proposition 1.** ([22]) *The transition density function of the Brownian motion  $W_t$ ,  $t \geq 0$  on  $S^2$  with radius  $a$  it is given by the function*

$$p(t, \varphi) = \frac{1}{4\pi a^2} \sum_{n \in \mathbb{N}} (2n+1) \exp\left(-\frac{n(n+1)\sqrt{t}}{a}\right) P_n^0(\cos \varphi) \quad (74)$$

**Theorem 19.** Consider function  $g : \mathbb{R} \rightarrow \mathbb{R}$ , which is bounded on  $[0, \pi]$ , i.e. there exists  $M > 0$  such that  $|g(\phi)| \leq M$ , for every  $\phi \in [0, \pi]$ , and Lebesgue measurable on  $\mathbb{R}$ . Let also  $W(t, \phi)$  be the Brownian motion on  $S^2$ . Then the expectation

$$E(|g(W)|)(t) = \int_0^\pi |g(\phi)| p(t, \phi) d\phi$$

is continuous in  $t$ , and

$$E(|g(W)|)(t) \leq \pi M p(t_o, \phi_o), \quad (75)$$

where

$$p(t_o, \phi_o) = \max_{(t, \phi) \in [t_1, t_2] \times [0, \pi]} p(t, \phi), \text{ with } 0 < t_1 < t_2.$$

Here  $p(t, \phi)$  is the transition density function of the Brownian motion  $W_t$ ,  $t \geq 0$  on  $S^2$  given by (74).

**Proof.** It is known that the transition density function of the Brownian motion  $W_t$ ,  $t \geq 0$  on  $S^2$ ,  $p(t, \phi)$  is continuous in  $(t, \phi) \in [t_1, t_2] \times [0, \pi]$ ,  $t_1 > 0$ .

By the extreme value theorem there exists  $(t_o, \phi_o) \in [t_1, t_2] \times [0, \pi]$  such that

$$p(t_o, \phi_o) = \max_{(t, \phi) \in [t_1, t_2] \times [0, \pi]} p(t, \phi).$$

So we have

$$0 \leq p(t, \phi) \leq p(t_o, \phi_o), \text{ for every } (t, \phi) \in [t_1, t_2] \times [0, \pi].$$

Let  $N \in \mathbb{N}$ ,  $t_N, t \in [t_1, t_2] : t_N \rightarrow t$ , as  $N \rightarrow \infty$ .

Then,  $p(t_N, \phi) \rightarrow p(t, \phi)$  for every  $\phi \in [0, \pi]$ .

The function  $g : \mathbb{R} \rightarrow \mathbb{R}$ , is bounded on  $[0, \pi]$ , i.e. there is a  $M > 0$  such that  $|g(\phi)| \leq M$ , for every  $\phi \in [0, \pi]$ , and Lebesgue measurable on  $\mathbb{R}$ .

Furthermore we have that,

$$|g(\phi)| p(t_N, \phi) \rightarrow |g(\phi)| p(t, \phi), \text{ as } N \rightarrow \infty$$

and

$$|g(\phi)| p(t_N, \phi) \leq |g(\phi)| p(t_o, \phi_o), \text{ for all } \phi \in [0, \pi] \text{ and } N \in \mathbb{N}.$$

So, by dominated convergence theorem we get that

$$E(|g(W)|)(t_N) \rightarrow E(|g(W)|)(t) \text{ as } N \rightarrow \infty.$$

Thus  $E(|g(W)|)(t)$  is proved to be continuous in  $t$ .

Moreover,

$$|g(\phi)| p(t, \phi) \leq M p(t_o, \phi_o) \text{ for all } t \in [t_1, t_2] \text{ and } \phi \in [0, \pi].$$

Thus,

$$E(|g(W)|)(t) = \int_0^\pi |g(\phi)| p(t, \phi) d\phi \leq \pi M p(t_o, \phi_o).$$

□

**Proposition 2.** Consider function  $g : \mathbb{R} \rightarrow \mathbb{R}$ , which is bounded on  $[0, \pi]$  and Lebesgue measurable on  $\mathbb{R}$ . Let also  $W(t, \phi)$  be the Brownian motion on  $S^2$ . Then the expectation

$$E(|g(W)|)(t) = \int_0^\pi |g(\phi)| p(t, \phi) d\phi$$

is differentiable in  $t$ , and

$$\frac{\partial E(|g(W)|)}{\partial t} = \int_0^\pi |g(\phi)| \frac{\partial(p(t, \phi))}{\partial t} d\phi, \quad (76)$$

which is continuous in  $t$ .

**Proof.** As we said before, the transition density function of the Brownian motion  $W_t, t \geq 0$  on  $S^2$ ,  $p(t, \phi)$  is continuous in  $(t, \phi) \in [t_1, t_2] \times [0, \pi], t_1 > 0$ .

We have

$$E(|g(W)|) = \int_0^\pi |g(\phi)| p(t, \phi) d\phi, \text{ for every } t \in [t_1, t_2].$$

We apply differentiation under the integral sign.

We notice

$$|g(\phi)| \frac{\partial p(t, \phi)}{\partial t} \leq M \left\| \frac{\partial p(t, \phi)}{\partial t} \right\|_{\infty, [t_1, t_2] \times [0, \pi]}.$$

Therefore there exists

$$\frac{\partial E(|g(W)|)}{\partial t} = \int_0^\pi |g(\phi)| \frac{\partial(p(t, \phi))}{\partial t} d\phi,$$

which is continuous in  $t$  (same proof as in Theorem 19).  $\square$

## 4 Main Results

We present the following general approximation results of Brownian Motion by neural network operators.

**Theorem 20.** Let  $0 < \alpha < 1, n \in \mathbb{N} : n^{1-\alpha} > 2, t \in [t_1, t_2]$ , where  $t_1 > 0, j = 1, 2, 3, 4$ . Then

1.

$$\begin{aligned} |{}_j A_n(E(|g(W)|))(t) - (E(|g(W)|))(t)| &\leq c_j \left[ \omega_1 \left( E(|g(W)|), \frac{1}{n^\alpha} \right) + \gamma_j(n) \|E(|g(W)|)\|_{\infty, [t_1, t_2]} \right] = \\ &=: \rho_j(E(|g(W)|)), \end{aligned} \quad (77)$$

and

2.

$$\|{}_j A_n(E(|g(W)|)) - E(|g(W)|)\|_{\infty, [t_1, t_2]} \leq \rho_j(E(|g(W)|)). \quad (78)$$

We have that  $\lim_{n \rightarrow \infty} {}_j A_n(E(|g(W)|)) = E(|g(W)|)$ , pointwise and uniformly.

**Proof.** By Theorem 17.  $\square$

**Theorem 21.** Let  $0 < \alpha < 1, n \in \mathbb{N} : n^{1-\alpha} > 2, t \in [t_1, t_2],$  where  $t_1 > 0, j = 1, 2, 3, 4.$  Then

1.

$$\left| {}_j A_n \left( \frac{\partial E(|g(W)|)}{\partial t} \right) (t) - \left( \frac{\partial E(|g(W)|)}{\partial t} \right) (t) \right| \leq c_j \left[ \omega_1 \left( \left( \frac{\partial E(|g(W)|)}{\partial t} \right), \frac{1}{n^\alpha} \right) + \gamma_j(n) \left\| \frac{\partial E(|g(W)|)}{\partial t} \right\|_{\infty, [t_1, t_2]} \right] =: \rho_j \left( \frac{\partial E(|g(W)|)}{\partial t} \right), \quad (79)$$

and

2.

$$\left\| {}_j A_n \left( \frac{\partial E(|g(W)|)}{\partial t} \right) - \frac{\partial E(|g(W)|)}{\partial t} \right\|_{\infty, [t_1, t_2]} \leq \rho_j \left( \frac{\partial E(|g(W)|)}{\partial t} \right). \quad (80)$$

We have that  $\lim_{n \rightarrow \infty} {}_j A_n \left( \frac{\partial E(|g(W)|)}{\partial t} \right) = \frac{\partial E(|g(W)|)}{\partial t},$  pointwise and uniformly.

**Proof.** By Theorem 17.  $\square$

Next we give the following fractional calculus related result.

**Theorem 22.** Let  $0 < \alpha, \beta < 1, t \in [t_1, t_2], t_1 > 0, n \in \mathbb{N} : n^{1-\beta} > 2, j = 1, 2, 3, 4.$  Then

$$\begin{aligned} & \left| {}_j A_n (E(|g(W)|)(t)) - E(|g(W)|)(t) \right| \leq \frac{c_j}{\Gamma(\alpha + 1)} \\ & \left\{ \frac{\omega_1 \left( D_t^\alpha E(|g(W)|), \frac{1}{n^\beta} \right)_{[t_1, t]} + \omega_1 \left( D_{*t}^\alpha E(|g(W)|), \frac{1}{n^\beta} \right)_{[t, t_2]} +}{n^{\alpha\beta}} \right. \\ & \left. + \epsilon_j(n) \left( \|D_t^\alpha E(|g(W)|)\|_{\infty, [t_1, t]} (t - t_1)^\alpha + \|D_{*t}^\alpha E(|g(W)|)\|_{\infty, [t, t_2]} (t_2 - t)^\alpha \right) \right\}. \quad (81) \end{aligned}$$

We have

$${}_j A_n (E(|g(W)|)(t)) \rightarrow E(|g(W)|)(t), \text{ as } n \rightarrow \infty$$

**Proof.** By Theorem 18.  $\square$

## 5 Applications

For a function  $g : \mathbb{R} \rightarrow \mathbb{R},$  which is bounded on  $[0, \pi]$  and Lebesgue measurable on  $\mathbb{R}$  and  $W(t, \phi)$  the Brownian motion on  $S^2.$  We will use the following notations

$$E(|g(W)|) := E(|g(W)|)^{(0)} \quad (82)$$

and

$$\frac{\partial E(|g(W)|)}{\partial t} := E(|g(W)|)^{(1)}. \quad (83)$$

We can apply our main results to the function  $g(W) = W.$  Consider the function  $g : \mathbb{R} \rightarrow \mathbb{R},$  where  $g(x) = x$  for every  $x \in \mathbb{R}.$  Let also  $W(t, \phi)$  be the Brownian motion on  $S^2.$  Then the expectation

$$E(|W|)(t) = \int_0^\pi \phi p(t, \phi) d\phi$$

is continuous in  $t.$

Moreover



**Corollary 1.** Let  $0 < \alpha < 1, n \in \mathbb{N} : n^{1-\alpha} > 2, t \in [t_1, t_2]$ , where  $t_1 > 0, i = 0, 1$  and  $j = 1, 2, 3, 4$ . Then

1.

$$\left| {}_j A_n \left( E(|W|)^{(i)} \right) (t) - \left( E(|W|)^{(i)} \right) (t) \right| \leq c_j \left[ \omega_1 \left( E(|W|)^{(i)}, \frac{1}{n^\alpha} \right) + \gamma_j(n) \left\| E(|W|)^{(i)} \right\|_{\infty, [t_1, t_2]} \right] =: \rho_j(E(|W|)^{(i)}), \quad (84)$$

and

2.

$$\left\| {}_j A_n \left( E(|W|)^{(i)} \right) - E(|W|)^{(i)} \right\|_{\infty, [t_1, t_2]} \leq \rho_j(E(|W|)^{(i)}). \quad (85)$$

We have that  $\lim_{n \rightarrow \infty} {}_j A_n(E(|W|)^{(i)}) = E(|W|)^{(i)}$ , pointwise and uniformly.

**Proof.** By Theorems 20 and 21.  $\square$

Next we give the following fractional calculus related result.

**Corollary 2.** Let  $0 < \alpha, \beta < 1, t \in [t_1, t_2], t_1 > 0, n \in \mathbb{N} : n^{1-\beta} > 2, j = 1, 2, 3, 4$ . Then

$$\begin{aligned} |{}_j A_n(E(|W|)(t)) - E(|W|)(t)| &\leq \frac{c_j}{\Gamma(\alpha + 1)} \\ &\left\{ \frac{\omega_1(D_t^\alpha E(|W|), \frac{1}{n^\beta})_{[t_1, t]} + \omega_1(D_{*t}^\alpha E(|W|), \frac{1}{n^\beta})_{[t, t_2]} +}{n^{\alpha\beta}} \right. \\ &\left. + \epsilon_j(n) \left( \|D_t^\alpha E(|W|)\|_{\infty, [t_1, t]} (t - t_1)^\alpha + \|D_{*t}^\alpha E(|W|)\|_{\infty, [t, t_2]} (t_2 - t)^\alpha \right) \right\}. \end{aligned} \quad (86)$$

We have

$${}_j A_n(E(|W|))(t) \rightarrow E(|W|)(t), \text{ as } n \rightarrow \infty$$

**Proof.** By Theorem 22.  $\square$

For the next corollaries we consider the function  $g : \mathbb{R} \rightarrow \mathbb{R}$ , where  $g(x) = \cos x$  for every  $x \in \mathbb{R}$ . Let also  $W(t, \phi)$  be the Brownian motion on  $S^2$ . Then the expectation

$$E(|\cos W|)(t) = \int_0^\pi |\cos \phi| p(t, \phi) d\phi$$

is continuous in  $t$ .

Moreover,

**Corollary 3.** Let  $0 < \alpha < 1, n \in \mathbb{N} : n^{1-\alpha} > 2, t \in [t_1, t_2]$ , where  $t_1 > 0, i = 0, 1$  and  $j = 1, 2, 3, 4$ . Then

1.

$$\left| {}_j A_n \left( E(|\cos W|)^{(i)} \right) (t) - \left( E(|\cos W|)^{(i)} \right) (t) \right| \leq c_j \left[ \omega_1 \left( E(|\cos W|)^{(i)}, \frac{1}{n^\alpha} \right) + \gamma_j(n) \left\| E(|\cos W|)^{(i)} \right\|_{\infty, [t_1, t_2]} \right] =: \rho_j(E(|\cos W|)^{(i)}), \quad (87)$$

and

2.

$$\left\| {}_j A_n \left( E(|\cos W|)^{(i)} \right) - E(|\cos W|)^{(i)} \right\|_{\infty, [t_1, t_2]} \leq \rho_j(E(|\cos W|)^{(i)}). \quad (88)$$

We have that  $\lim_{n \rightarrow \infty} {}_j A_n(E(|\cos W|)^{(i)}) = E(|\cos W|)^{(i)}$ , pointwise and uniformly.

**Proof.** By Theorems 20 and 21.  $\square$

Next we give the following fractional calculus related result.

**Corollary 4.** Let  $0 < \alpha, \beta < 1, t \in [t_1, t_2], t_1 > 0, n \in \mathbb{N} : n^{1-\beta} > 2, j = 1, 2, 3, 4$ . Then

$$\begin{aligned} & \left| {}_j A_n(E(|\cos W|)(t)) - E(|\cos W|)(t) \right| \leq \frac{c_j}{\Gamma(\alpha + 1)} \\ & \left\{ \frac{\omega_1 \left( D_t^\alpha E(|\cos W|), \frac{1}{n^\beta} \right)_{[t_1, t]} + \omega_1 \left( D_{*t}^\alpha E(|\cos W|), \frac{1}{n^\beta} \right)_{[t, t_2]} +}{n^{\alpha\beta}} \right. \\ & \left. + \epsilon_j(n) \left( \|D_t^\alpha E(|\cos W|)\|_{\infty, [t_1, t]} (t - t_1)^\alpha + \|D_{*t}^\alpha E(|\cos W|)\|_{\infty, [t, t_2]} (t_2 - t)^\alpha \right) \right\}. \quad (89) \end{aligned}$$

We have

$${}_j A_n(E(|\cos W|))(t) \rightarrow E(|\cos W|)(t), \text{ as } n \rightarrow \infty$$

**Proof.** By Theorem 22.  $\square$

Let the function  $g : \mathbb{R} \rightarrow \mathbb{R}$ , where  $g(x) = \sin x$  for every  $x \in \mathbb{R}$ . Let also  $W(t, \phi)$  be the Brownian motion on  $S^2$ . Then the expectation

$$E(|\sin W|)(t) = \int_0^\pi \sin(\phi) p(t, \phi) d\phi$$

is continuous in  $t$ .

Moreover,

**Corollary 5.** Let  $0 < \alpha < 1, n \in \mathbb{N} : n^{1-\alpha} > 2, t \in [t_1, t_2]$ , where  $t_1 > 0, i = 0, 1$  and  $j = 1, 2, 3, 4$ . Then

1.

$$\begin{aligned} & \left| {}_j A_n \left( E(|\sin W|)^{(i)} \right) (t) - \left( E(|\sin W|)^{(i)} \right) (t) \right| \leq \\ & c_j \left[ \omega_1 \left( E(|\sin W|)^{(i)}, \frac{1}{n^\alpha} \right) + \gamma_j(n) \left\| E(|\sin W|)^{(i)} \right\|_{\infty, [t_1, t_2]} \right] =: \rho_j(E(|\sin W|)^{(i)}), \quad (90) \end{aligned}$$

and

2.

$$\left\| {}_j A_n \left( E(|\sin W|)^{(i)} \right) - E(|\sin W|)^{(i)} \right\|_{\infty, [t_1, t_2]} \leq \rho_j(E(|\sin W|)^{(i)}). \quad (91)$$

We have that  $\lim_{n \rightarrow \infty} {}_j A_n(E(|\sin W|)^{(i)}) = E(|\sin W|)^{(i)}$ , pointwise and uniformly.

**Proof.** By Theorems 20 and 21.  $\square$

Next we give the following fractional calculus related result.

**Corollary 6.** Let  $0 < \alpha, \beta < 1, t \in [t_1, t_2], t_1 > 0, n \in \mathbb{N} : n^{1-\beta} > 2, j = 1, 2, 3, 4$ . Then

$$\begin{aligned} & |{}_j A_n (E(|\sin W|)(t)) - E(|\sin W|)(t)| \leq \frac{c_j}{\Gamma(\alpha + 1)} \\ & \left\{ \frac{\omega_1 (D_{t^-}^\alpha E(|\sin W|), \frac{1}{n^\beta})_{[t_1, t]} + \omega_1 (D_{*t}^\alpha E(|\sin W|), \frac{1}{n^\beta})_{[t, t_2]} +}{n^{\alpha\beta}} \right. \\ & \left. + \epsilon_j(n) \left( \|D_{t^-}^\alpha E(|\sin W|)\|_{\infty, [t_1, t]} (t - t_1)^\alpha + \|D_{*t}^\alpha E(|\sin W|)\|_{\infty, [t, t_2]} (t_2 - t)^\alpha \right) \right\}. \end{aligned} \quad (92)$$

We have

$${}_j A_n (E(|\sin W|))(t) \rightarrow E(|\sin W|)(t), \text{ as } n \rightarrow \infty$$

**Proof.** By Theorem 22.  $\square$

Let the function  $g : \mathbb{R} \rightarrow \mathbb{R}$ , where  $g(x) = \tanh x$  for every  $x \in \mathbb{R}$ . Let also  $W(t, \phi)$  be the Brownian motion on  $S^2$ . Then the expectation

$$E(|\tanh W|)(t) = \int_0^\pi |\tanh(\phi)| p(t, \phi) d\phi$$

is continuous in  $t$ .

Moreover,

**Corollary 7.** Let  $0 < \alpha < 1, n \in \mathbb{N} : n^{1-\alpha} > 2, t \in [t_1, t_2]$ , where  $t_1 > 0, i = 0, 1$  and  $j = 1, 2, 3, 4$ . Then

1.

$$\begin{aligned} & \left| {}_j A_n \left( E(|\tanh W|)^{(i)} \right) (t) - \left( E(|\tanh W|)^{(i)} \right) (t) \right| \leq \\ & c_j \left[ \omega_1 \left( E(|\tanh W|)^{(i)}, \frac{1}{n^\alpha} \right) + \gamma_j(n) \left\| E(|\tanh W|)^{(i)} \right\|_{\infty, [t_1, t_2]} \right] =: \rho_j (E(|\tanh W|)^{(i)}), \end{aligned} \quad (93)$$

and

2.

$$\left\| {}_j A_n \left( E(|\tanh W|)^{(i)} \right) - E(|\tanh W|)^{(i)} \right\|_{\infty, [t_1, t_2]} \leq \rho_j (E(|\tanh W|)^{(i)}). \quad (94)$$

We have that  $\lim_{n \rightarrow \infty} {}_j A_n (E(|\tanh W|)^{(i)}) = E(|\tanh W|)^{(i)}$ , pointwise and uniformly.

**Proof.** By Theorems 20 and 21.  $\square$

Next we give the following fractional calculus related result.

**Corollary 8.** Let  $0 < \alpha, \beta < 1, t \in [t_1, t_2], t_1 > 0, n \in \mathbb{N} : n^{1-\beta} > 2, j = 1, 2, 3, 4$ . Then

$$\begin{aligned} & |{}_j A_n (E(|\tanh W|)(t)) - E(|\tanh W|)(t)| \leq \frac{c_j}{\Gamma(\alpha + 1)} \\ & \left\{ \frac{\omega_1 (D_{t^-}^\alpha E(|\tanh W|), \frac{1}{n^\beta})_{[t_1, t]} + \omega_1 (D_{*t}^\alpha E(|\tanh W|), \frac{1}{n^\beta})_{[t, t_2]} +}{n^{\alpha\beta}} \right. \\ & \left. + \epsilon_j(n) \left( \|D_{t^-}^\alpha E(|\tanh W|)\|_{\infty, [t_1, t]} (t - t_1)^\alpha + \|D_{*t}^\alpha E(|\tanh W|)\|_{\infty, [t, t_2]} (t_2 - t)^\alpha \right) \right\}. \end{aligned} \quad (95)$$

We have

$${}_j A_n (E(|\tanh W|))(t) \rightarrow E(|\tanh W|)(t), \text{ as } n \rightarrow \infty$$

**Proof.** By Theorem 22.  $\square$

In the following let us consider the function  $g : \mathbb{R} \rightarrow \mathbb{R}$ , where  $g(x) = e^{-\ell x}$ ,  $\ell > 0$  for every  $x \in \mathbb{R}$ . Let also  $W(t, \phi)$  be the Brownian motion on  $S^2$ . Then the expectation

$$E\left(e^{-\ell x}\right)(t) = \int_0^\pi e^{-\ell \phi} p(t, \phi) d\phi$$

is continuous in  $t$ .

Moreover,

**Corollary 9.** Let  $0 < \alpha < 1$ ,  $n \in \mathbb{N} : n^{1-\alpha} > 2$ ,  $\ell > 0$  and  $t \in [t_1, t_2]$ , where  $t_1 > 0$ ,  $i = 0, 1$  and  $j = 1, 2, 3, 4$ . Then

$$1. \quad \left| {}_j A_n \left( E \left( e^{-\ell w} \right)^{(i)} \right) (t) - \left( E \left( e^{-\ell w} \right)^{(i)} \right) (t) \right| \leq c_j \left[ \omega_1 \left( E \left( e^{-\ell w} \right)^{(i)}, \frac{1}{n^\alpha} \right) + \gamma_j(n) \left\| E \left( e^{-\ell w} \right)^{(i)} \right\|_{\infty, [t_1, t_2]} \right] =: \rho_j \left( E \left( e^{-\ell w} \right)^{(i)} \right), \quad (96)$$

and

$$2. \quad \left\| {}_j A_n \left( E \left( e^{-\ell w} \right)^{(i)} \right) - E \left( e^{-\ell w} \right)^{(i)} \right\|_{\infty, [t_1, t_2]} \leq \rho_j \left( E \left( e^{-\ell w} \right)^{(i)} \right). \quad (97)$$

We have that  $\lim_{n \rightarrow \infty} {}_j A_n \left( E \left( e^{-\ell w} \right)^{(i)} \right) = E \left( e^{-\ell w} \right)^{(i)}$ , pointwise and uniformly.

**Proof.** By Theorems 20 and 21.  $\square$

Next we give the following fractional calculus related result.

**Corollary 10.** Let  $0 < \alpha, \beta < 1$ ,  $\ell > 0$ ,  $t \in [t_1, t_2]$ ,  $t_1 > 0$ ,  $n \in \mathbb{N} : n^{1-\beta} > 2$ ,  $j = 1, 2, 3, 4$ . Then

$$\left| {}_j A_n \left( E \left( e^{-\ell w} \right) (t) \right) - E \left( e^{-\ell w} \right) (t) \right| \leq \frac{c_j}{\Gamma(\alpha + 1)} \left\{ \frac{\omega_1 \left( D_{t^-}^\alpha E \left( e^{-\ell w} \right), \frac{1}{n^\beta} \right)_{[t_1, t]} + \omega_1 \left( D_{*t}^\alpha E \left( e^{-\ell w} \right), \frac{1}{n^\beta} \right)_{[t, t_2]}}{n^{\alpha\beta}} + \epsilon_j(n) \left( \left\| D_{t^-}^\alpha E \left( e^{-\ell w} \right) \right\|_{\infty, [t_1, t]} (t - t_1)^\alpha + \left\| D_{*t}^\alpha E \left( e^{-\ell w} \right) \right\|_{\infty, [t, t_2]} (t_2 - t)^\alpha \right) \right\}. \quad (98)$$

We have

$${}_j A_n \left( E \left( e^{-\ell w} \right) \right) (t) \rightarrow E \left( e^{-\ell w} \right) (t), \text{ as } n \rightarrow \infty$$

**Proof.** By Theorem 22.  $\square$

In the following we consider the logistic sigmoid function  $g : \mathbb{R} \rightarrow \mathbb{R}$ , where  $g(x) = \frac{1}{1+e^{-x}}$  for every  $x \in \mathbb{R}$ . Let also  $W(t, \phi)$  be the Brownian motion on  $S^2$ . Then the expectation

$$E \left( \frac{1}{1+e^{-W}} \right) (t) = \int_0^\pi \frac{1}{1+e^{-\phi}} p(t, \phi) d\phi$$

is continuous in  $t$ .

Moreover,

**Corollary 11.** Let  $0 < \alpha < 1, n \in \mathbb{N} : n^{1-\alpha} > 2$  and  $t \in [t_1, t_2]$ , where  $t_1 > 0, i = 0, 1$  and  $j = 1, 2, 3, 4$ . Then

1.

$$\left| {}_j A_n \left( E \left( \frac{1}{1+e^{-W}} \right)^{(i)} \right) (t) - \left( E \left( \frac{1}{1+e^{-W}} \right)^{(i)} \right) (t) \right| \leq c_j \left[ \omega_1 \left( E \left( \frac{1}{1+e^{-W}} \right)^{(i)}, \frac{1}{n^\alpha} \right) + \gamma_j(n) \left\| E \left( \frac{1}{1+e^{-W}} \right)^{(i)} \right\|_{\infty, [t_1, t_2]} \right] =: \rho_j \left( E \left( \frac{1}{1+e^{-W}} \right)^{(i)} \right), \quad (99)$$

and

2.

$$\left\| {}_j A_n \left( E \left( \frac{1}{1+e^{-W}} \right)^{(i)} \right) - E \left( \frac{1}{1+e^{-W}} \right)^{(i)} \right\|_{\infty, [t_1, t_2]} \leq \rho_j \left( E \left( \frac{1}{1+e^{-W}} \right)^{(i)} \right). \quad (100)$$

We have that  $\lim_{n \rightarrow \infty} {}_j A_n \left( E \left( \frac{1}{1+e^{-W}} \right)^{(i)} \right) = E \left( \frac{1}{1+e^{-W}} \right)^{(i)}$ , pointwise and uniformly.

**Proof.** By Theorems 20 and 21.  $\square$

Next we give the following fractional calculus related result.

**Corollary 12.** Let  $0 < \alpha, \beta < 1, t \in [t_1, t_2], t_1 > 0, n \in \mathbb{N} : n^{1-\beta} > 2, j = 1, 2, 3, 4$ . Then

$$\left| {}_j A_n \left( E \left( \frac{1}{1+e^{-W}} \right) (t) \right) - E \left( \frac{1}{1+e^{-W}} \right) (t) \right| \leq \frac{c_j}{\Gamma(\alpha+1)} \left\{ \frac{\omega_1 \left( D_{t^-}^\alpha E \left( \frac{1}{1+e^{-W}} \right), \frac{1}{n^\beta} \right)_{[t_1, t]} + \omega_1 \left( D_{*t}^\alpha E \left( \frac{1}{1+e^{-W}} \right), \frac{1}{n^\beta} \right)_{[t, t_2]}}{n^{\alpha\beta}} + \epsilon_j(n) \left( \left\| D_{t^-}^\alpha E \left( \frac{1}{1+e^{-W}} \right) \right\|_{\infty, [t_1, t]} (t-t_1)^\alpha + \left\| D_{*t}^\alpha E \left( \frac{1}{1+e^{-W}} \right) \right\|_{\infty, [t, t_2]} (t_2-t)^\alpha \right) \right\}. \quad (101)$$

We have

$${}_j A_n \left( E \left( \frac{1}{1+e^{-W}} \right) \right) (t) \rightarrow E \left( \frac{1}{1+e^{-W}} \right) (t), \text{ as } n \rightarrow \infty$$

**Proof.** By Theorem 22.  $\square$

Let now as consider the generalised logistic sigmoid function  $g : \mathbb{R} \rightarrow \mathbb{R}$ , where  $g(x) = (1 + e^{-x})^{-\delta}$ , where  $\delta > 0$ , for every  $x \in \mathbb{R}$ . Let also  $W(t, \phi)$  be the Brownian motion on  $S^2$ . Then the expectation

$$E \left( (1 + e^{-W})^{-\delta} \right) (t) = \int_0^\pi \left( 1 + e^{-\phi} \right)^{-\delta} p(t, \phi) d\phi$$

is continuous in  $t$ .

Moreover,

**Corollary 13.** Let  $0 < \alpha < 1, \delta > 0, n \in \mathbb{N} : n^{1-\alpha} > 2$  and  $t \in [t_1, t_2]$ , where  $t_1 > 0, i = 0, 1$  and  $j = 1, 2, 3, 4$ . Then

$$\begin{aligned}
1. \quad & \left| {}_j A_n \left( E \left( (1 + e^{-W})^{-\delta} \right)^{(i)} \right) (t) - \left( E \left( (1 + e^{-W})^{-\delta} \right)^{(i)} \right) (t) \right| \leq \\
& c_j \left[ \omega_1 \left( E \left( (1 + e^{-W})^{-\delta} \right)^{(i)}, \frac{1}{n^\alpha} \right) + \gamma_j(n) \left\| E \left( (1 + e^{-W})^{-\delta} \right)^{(i)} \right\|_{\infty, [t_1, t_2]} \right] \\
& =: \rho_j \left( E \left( (1 + e^{-W})^{-\delta} \right)^{(i)} \right), \tag{102}
\end{aligned}$$

and

$$\begin{aligned}
2. \quad & \left\| {}_j A_n \left( E \left( (1 + e^{-W})^{-\delta} \right)^{(i)} \right) - E \left( (1 + e^{-W})^{-\delta} \right)^{(i)} \right\|_{\infty, [t_1, t_2]} \leq \rho_j \left( E \left( (1 + e^{-W})^{-\delta} \right)^{(i)} \right). \tag{103}
\end{aligned}$$

We have that  $\lim_{n \rightarrow \infty} {}_j A_n \left( E \left( (1 + e^{-W})^{-\delta} \right)^{(i)} \right) = E \left( (1 + e^{-W})^{-\delta} \right)^{(i)}$ , pointwise and uniformly.

**Proof.** By Theorems 20 and 21.  $\square$

Next we give the following fractional calculus related result.

**Corollary 14.** Let  $0 < \alpha, \beta < 1, \delta > 0, t \in [t_1, t_2], t_1 > 0, n \in \mathbb{N} : n^{1-\beta} > 2, j = 1, 2, 3, 4$ .

Then

$$\begin{aligned}
& \left| {}_j A_n \left( E \left( (1 + e^{-W})^{-\delta} \right) (t) \right) - E \left( (1 + e^{-W})^{-\delta} \right) (t) \right| \leq \frac{c_j}{\Gamma(\alpha + 1)} \\
& \left\{ \frac{\omega_1 \left( D_{t^-}^\alpha E \left( (1 + e^{-W})^{-\delta} \right), \frac{1}{n^\beta} \right)_{[t_1, t]} + \omega_1 \left( D_{*t}^\alpha E \left( (1 + e^{-W})^{-\delta} \right), \frac{1}{n^\beta} \right)_{[t, t_2]} +}{n^{\alpha\beta}} \right. \\
& \left. + \epsilon_j(n) \left( \left\| D_{t^-}^\alpha E \left( (1 + e^{-W})^{-\delta} \right) \right\|_{\infty, [t_1, t]} (t - t_1)^\alpha + \left\| D_{*t}^\alpha E \left( (1 + e^{-W})^{-\delta} \right) \right\|_{\infty, [t, t_2]} (t_2 - t)^\alpha \right) \right\}. \tag{104}
\end{aligned}$$

We have

$${}_j A_n \left( E \left( (1 + e^{-W})^{-\delta} \right) \right) (t) \rightarrow E \left( (1 + e^{-W})^{-\delta} \right) (t), \text{ as } n \rightarrow \infty$$

**Proof.** By Theorem 22.  $\square$

The Gompertz function  $g : \mathbb{R} \rightarrow \mathbb{R}$ , with  $g(x) = e^{\mu e^{-x}}, \mu < 0$  is a sigmoid function which describes growth as being slowest at the start and end of a given time period. Let  $W(t, \phi)$  be the Brownian motion on  $S^2$ . Then the expectation

$$E \left( e^{\mu e^{-W}} \right) (t) = \int_0^\pi e^{\mu e^{-\phi}} p(t, \phi) d\phi$$

is continuous in  $t$ .

Moreover,

**Corollary 15.** Let  $0 < \alpha < 1, \mu < 0, n \in \mathbb{N} : n^{1-\alpha} > 2$  and  $t \in [t_1, t_2]$ , where  $t_1 > 0, i = 0, 1$  and  $j = 1, 2, 3, 4$ . Then

$$1. \quad \left| {}_j A_n \left( E \left( e^{\mu e^{-W}} \right)^{(i)} \right) (t) - \left( E \left( e^{\mu e^{-W}} \right)^{(i)} \right) (t) \right| \leq c_j \left[ \omega_1 \left( E \left( \left( e^{\mu e^{-W}} \right)^{(i)} \right), \frac{1}{n^\alpha} \right) + \gamma_j(n) \left\| E \left( \left( e^{\mu e^{-W}} \right)^{(i)} \right) \right\|_{\infty, [t_1, t_2]} \right] =: \rho_j \left( E \left( e^{\mu e^{-W}} \right)^{(i)} \right), \quad (105)$$

and

$$2. \quad \left\| {}_j A_n \left( E \left( \left( e^{\mu e^{-W}} \right)^{(i)} \right) \right) - E \left( \left( e^{\mu e^{-W}} \right)^{(i)} \right) \right\|_{\infty, [t_1, t_2]} \leq \rho_j \left( E \left( e^{\mu e^{-W}} \right)^{(i)} \right). \quad (106)$$

We have that  $\lim_{n \rightarrow \infty} {}_j A_n \left( E \left( e^{\mu e^{-W}} \right)^{(i)} \right) = E \left( e^{\mu e^{-W}} \right)^{(i)}$ , pointwise and uniformly.

**Proof.** By Theorems 20 and 21.  $\square$

Next we give the following fractional calculus related result.

**Corollary 16.** Let  $0 < \alpha, \beta < 1, \mu < 0, t \in [t_1, t_2], t_1 > 0, n \in \mathbb{N} : n^{1-\beta} > 2, j = 1, 2, 3, 4$ . Then

$$\left| {}_j A_n \left( E \left( e^{\mu e^{-W}} \right) (t) \right) - E \left( e^{\mu e^{-W}} \right) (t) \right| \leq \frac{c_j}{\Gamma(\alpha + 1)} \left\{ \frac{\omega_1 \left( D_{t^-}^\alpha E \left( e^{\mu e^{-W}} \right), \frac{1}{n^\beta} \right)_{[t_1, t]} + \omega_1 \left( D_{*t}^\alpha E \left( e^{\mu e^{-W}} \right), \frac{1}{n^\beta} \right)_{[t, t_2]}}{n^{\alpha\beta}} + \epsilon_j(n) \left( \left\| D_{t^-}^\alpha E \left( e^{\mu e^{-W}} \right) \right\|_{\infty, [t_1, t]} (t - t_1)^\alpha + \left\| D_{*t}^\alpha E \left( e^{\mu e^{-W}} \right) \right\|_{\infty, [t, t_2]} (t_2 - t)^\alpha \right) \right\}. \quad (107)$$

We have

$${}_j A_n \left( E \left( e^{\mu e^{-W}} \right) \right) (t) \rightarrow E \left( e^{\mu e^{-W}} \right) (t), \text{ as } n \rightarrow \infty$$

**Proof.** By Theorem 22.  $\square$

An other special case of The Gompertz functions is  $g : \mathbb{R} \rightarrow \mathbb{R}$ , with  $g(x) = e^{-e^{\kappa x}}, \kappa < 0$ . Let  $W(t, \phi)$  be the Brownian motion on  $S^2$ . Then the expectation

$$E \left( e^{-e^{\kappa W}} \right) (t) = \int_0^\pi e^{-e^{-\kappa \phi}} p(t, \phi) d\phi$$

is continuous in  $t$ .

Moreover,

**Corollary 17.** Let  $0 < \alpha < 1, \kappa < 0, n \in \mathbb{N} : n^{1-\alpha} > 2$  and  $t \in [t_1, t_2]$ , where  $t_1 > 0, i = 0, 1$  and  $j = 1, 2, 3, 4$ . Then

1.

$$\begin{aligned} & \left| {}_j A_n \left( E \left( e^{-e^{\kappa W}} \right)^{(i)} \right) (t) - \left( E \left( e^{-e^{\kappa W}} \right)^{(i)} \right) (t) \right| \leq \\ & c_j \left[ \omega_1 \left( E \left( \left( e^{-e^{\kappa W}} \right)^{(i)} \right), \frac{1}{n^\alpha} \right) + \gamma_j(n) \left\| E \left( \left( e^{-e^{\kappa W}} \right)^{(i)} \right) \right\|_{\infty, [t_1, t_2]} \right] \\ & =: \rho_j \left( E \left( e^{-e^{\kappa W}} \right)^{(i)} \right), \end{aligned} \quad (108)$$

and

2.

$$\left\| {}_j A_n \left( E \left( e^{-e^{\kappa W}} \right)^{(i)} \right) - E \left( e^{-e^{\kappa W}} \right)^{(i)} \right\|_{\infty, [t_1, t_2]} \leq \rho_j \left( E \left( e^{-e^{\kappa W}} \right)^{(i)} \right). \quad (109)$$

We have that  $\lim_{n \rightarrow \infty} {}_j A_n \left( E \left( e^{-e^{\kappa W}} \right)^{(i)} \right) = E \left( e^{-e^{\kappa W}} \right)^{(i)}$ , pointwise and uniformly.

**Proof.** By Theorems 20 and 21.  $\square$

Next we give the following fractional calculus related result.

**Corollary 18.** Let  $0 < \alpha, \beta < 1, \kappa < 0, t \in [t_1, t_2], t_1 > 0, n \in \mathbb{N} : n^{1-\beta} > 2, j = 1, 2, 3, 4$ .

Then

$$\begin{aligned} & \left| {}_j A_n \left( E \left( e^{-e^{\kappa W}} \right) (t) \right) - E \left( e^{-e^{\kappa W}} \right) (t) \right| \leq \frac{c_j}{\Gamma(\alpha + 1)} \\ & \left\{ \frac{\omega_1 \left( D_{t^-}^\alpha E \left( e^{-e^{\kappa W}} \right), \frac{1}{n^\beta} \right)_{[t_1, t]} + \omega_1 \left( D_{*t}^\alpha E \left( e^{-e^{\kappa W}} \right), \frac{1}{n^\beta} \right)_{[t, t_2]} +}{n^{\alpha\beta}} \right. \\ & \left. + \epsilon_j(n) \left( \left\| D_{t^-}^\alpha E \left( e^{-e^{\kappa W}} \right) \right\|_{\infty, [t_1, t]} (t - t_1)^\alpha + \left\| D_{*t}^\alpha E \left( e^{-e^{\kappa W}} \right) \right\|_{\infty, [t, t_2]} (t_2 - t)^\alpha \right) \right\}. \end{aligned} \quad (110)$$

We have

$${}_j A_n \left( E \left( e^{-e^{\kappa W}} \right) \right) (t) \rightarrow E \left( e^{-e^{\kappa W}} \right) (t), \text{ as } n \rightarrow \infty$$

**Proof.** By Theorem 22.  $\square$

Finally, we consider the function  $g : \mathbb{R} \rightarrow \mathbb{R}$ , with  $g(x) = P_m^0(\cos x)$ , where  $P_m^0(x)$  is the Legendre Polynomial of degree  $m$ . Let  $W(t, \phi)$  be the Brownian motion on  $S^2$ . Then the expectation

$$E \left( |P_m^0(\cos W)| \right) (t) = \int_0^\pi |P_m^0(\cos \phi)| p(t, \phi) d\phi$$

is continuous in  $t$ .

Moreover,

**Corollary 19.** Let  $0 < \alpha < 1, n \in \mathbb{N} : n^{1-\alpha} > 2$  and  $t \in [t_1, t_2]$ , where  $t_1 > 0, i = 0, 1$  and  $j = 1, 2, 3, 4$ . Then

1.

$$\left| {}_j A_n \left( E \left( |P_m^0(\cos W)| \right)^{(i)} \right) (t) - \left( E \left( |P_m^0(\cos W)| \right)^{(i)} \right) (t) \right| \leq$$



$$\begin{aligned}
& c_j \left[ \omega_1 \left( E \left( (|P_m^0(\cos W)|)^{(i)} \right), \frac{1}{n^\alpha} \right) + \gamma_j(n) \left\| E \left( (|P_m^0(\cos W)|)^{(i)} \right) \right\|_{\infty, [t_1, t_2]} \right] \\
& \quad =: \rho_j \left( E \left( (|P_m^0(\cos W)|)^{(i)} \right) \right), \tag{111}
\end{aligned}$$

and

2.

$$\left\| {}_j A_n \left( E \left( (|P_m^0(\cos W)|)^{(i)} \right) \right) - E \left( (|P_m^0(\cos W)|)^{(i)} \right) \right\|_{\infty, [t_1, t_2]} \leq \rho_j \left( E \left( (|P_m^0(\cos W)|)^{(i)} \right) \right). \tag{112}$$

We have that  $\lim_{n \rightarrow \infty} {}_j A_n(E(|P_n^0(\cos W)|)^{(i)}) = E(|P_n^0(\cos W)|)^{(i)}$ , pointwise and uniformly.

**Proof.** By Theorems 20 and 21.  $\square$

Next we give the following fractional calculus related result.

**Corollary 20.** Let  $0 < \alpha, \beta < 1$ ,  $t \in [t_1, t_2]$ ,  $t_1 > 0$ ,  $n \in \mathbb{N} : n^{1-\beta} > 2$ ,  $j = 1, 2, 3, 4$ . Then

$$\begin{aligned}
& \left| {}_j A_n \left( E \left( (|P_m^0(\cos W)|) \right) (t) \right) - E \left( (|P_m^0(\cos W)|) \right) (t) \right| \leq \frac{c_j}{\Gamma(\alpha + 1)} \\
& \left\{ \frac{\omega_1 \left( D_{t^-}^\alpha E \left( (|P_m^0(\cos W)|) \right), \frac{1}{n^\beta} \right)_{[t_1, t]} + \omega_1 \left( D_{*t}^\alpha E \left( (|P_m^0(\cos W)|) \right), \frac{1}{n^\beta} \right)_{[t, t_2]} + \right. \\
& \left. + \epsilon_j(n) \left( \left\| D_{t^-}^\alpha E \left( (|P_m^0(\cos W)|) \right) \right\|_{\infty, [t_1, t]} (t - t_1)^\alpha + \left\| D_{*t}^\alpha E \left( (|P_m^0(\cos W)|) \right) \right\|_{\infty, [t, t_2]} (t_2 - t)^\alpha \right) \right\}. \tag{113}
\end{aligned}$$

We have

$${}_j A_n \left( E \left( (|P_m^0(\cos W)|) \right) \right) (t) \rightarrow E \left( (|P_m^0(\cos W)|) \right) (t), \text{ as } n \rightarrow \infty$$

**Proof.** By Theorem 22.  $\square$

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