



Article

Fractional Fokker-Planck Equation [†]

Gerd Baumann 1,2,* and Frank Stenger 3

- Mathematics Department, German University in Cairo, New Cairo City 11835, Egypt
- ² University of Ulm, D-89069 Ulm, Germany
- University of Utah, Salt Lake City, UT 84112, USA; sinc_f@msn.com
- * Correspondence: gerd.baumann@guc.edu.eg
- † In memoriam Th.F. Nonnenmacher

Academic Editor: Rui A. C. Ferreira

Received: 27 October 2016; Accepted: 3 February 2017; Published: 11 February 2017

Abstract: We shall discuss the numerical solution of the Cauchy problem for the fully fractional Fokker-Planck (fFP) equation in connection with Sinc convolution methods. The numerical approximation is based on *Caputo* and *Riesz-Feller* fractional derivatives. The use of the transfer function in Laplace and Fourier spaces in connection with Sinc convolutions allow to find exponentially converging computing schemes. Examples using different initial conditions demonstrate the effective computations with a small number of grid points on an infinite spatial domain.

Keywords: sinc methods; approximation; computation; integral equations; *Riesz-Feller* derivative; *Caputo* derivative; fractional Fokker Planck equation

MSC: 65D05, 65D30, 44A35, 81-04

1. Introduction

The Fokker-Planck Equation (FP) is used in models of standard diffusion problems involving external fields:

$$\partial_t u(x,t) = L_{\text{FP}} u(x,t) \text{ with } -\infty < x < \infty \text{ and } t \ge 0$$
 (1)

where the linear FP operator is defined by L_{FP} :

$$L_{\rm FP} = \frac{\partial}{\partial x} \frac{V'(x)}{m\eta} + \epsilon \frac{\partial^2}{\partial x^2} \tag{2}$$

with the external potential V(x) [1]. V' represents the negative external forces in the system. The parameters m, η and ϵ are the mass, the friction coefficient, and the diffusion constant, respectively. Risken in his book collected and developed a rich variety of solution methods for the FP equation [1]. The basis of the FP equation is a diffusion process re-distributing u(x,t) in space x and time t the quantity may be related e.g., to mass, charge, or probability. From (2) it becomes apparent that the FP equation is a generalization of the diffusion equation. In recent years the integer order diffusion equation was generalized to a fractional diffusion equation in which the differentiations with respect to t and x are replaced by differentiations of non-integer order [2–5]. The aim of the current paper is to use such fractional generalizations for the FP equation, too.

Let us first examine Risken's approach. To this end, we first separate the diffusion part from the rest of the equation, as follows:

$$\partial_t u(x,t) - \epsilon \frac{\partial^2 u(x,t)}{\partial x^2} = -\frac{\partial}{\partial x} \left(\frac{V'(x)}{m\eta} u(x,t) \right) \text{ with } -\infty < x < \infty \text{ and } t \ge 0$$
 (3)

Mathematics 2017, 5, 12 2 of 19

We now assume that the function u(x,t) satisfies natural Dirichlet boundary conditions; i.e., $u(\pm \infty,t)=0$. The initial condition for the problem is $u(x,t=0)=u^{(0)}(x)$. Without loss of generality, we rescale the temporal coordinate by using the diffusion constant $t\to t\varepsilon$ so that the equation becomes:

$$\partial_t u(x,t) - \frac{\partial^2 u(x,t)}{\partial x^2} = -\frac{\partial}{\partial x} \left(\frac{V'(x)}{\epsilon m \eta} u(x,t) \right) \text{ with } -\infty < x < \infty \text{ and } t \ge 0$$
 (4)

It is well known that for vanishing external forces, the fundamental solution of (4) is given by a Gaussian:

$$G(x,t) = (4\pi t)^{1/2} \exp\left(-\frac{x^2}{4t}\right)$$
 (5)

This fundamental solution enables us to transform Equation (4) to the integral equation:

$$u(x,t) = v(x,t) + \int_0^t \int_{\mathbb{R}} G(x-\xi,t-\tau) \frac{\partial}{\partial \xi} \left(\frac{V'(\xi)}{\epsilon m\eta} u(\xi,\tau) \right) d\xi d\tau$$

$$= v(x,t) - \int_0^t \int_{\mathbb{R}} \frac{\partial}{\partial \xi} (G(x-\xi,t-\tau)) w(\xi) u(\xi,\tau) d\xi d\tau$$

$$= v(x,t) - \int_0^t \int_{\mathbb{R}} G_x (x-\xi,t-\tau) w(\xi) u(\xi,\tau) d\xi d\tau$$
(6)

in which $w(x) = V'(x)/(\epsilon m\eta)$ is the scaled external force, and in which the first derivative G_x is given by:

$$G_x(x,t) = -\pi \frac{x}{(4\pi t)^{3/2}} \exp\left(-\frac{x^2}{4t}\right) \tag{7}$$

The function v(x, t) related to the initial condition is given by:

$$v(x,t) = \int_{\mathbb{D}} G(x-\xi,t)u^{(0)}(\xi)d\xi \tag{8}$$

The integral equations (IE) (6) and (8) can now be used to derive the solution of the FP equation. These straight forward steps enable us to transform the FP problem into an integral equation, which is suitable for approximate solution by our methods given below in this paper (see also [6,7]). We remark that there already exist a number of other methods for solving the FP equation. We refer to Risken [1] for an overview of such methods, and to [8–11] for a discussion of more recent developments. Our method of approximate solution has advantages over these other methods, for ease of solution and for exponential convergence.

Our aim in this paper is to generalize the integer order original FP equation to it's fractional form, the fractional Fokker-Planck (fFP) equation. There exists a large number of papers dealing with such generalizations see e.g., [12–14] and references therein. However, constraints were added in each of these papers by way of replacing integer order derivatives by fractional order derivatives. Only the temporal derivative is replaced in some of these publications [13]. Some other papers consider replacing only the second order spacial whereas the other derivatives are kept at integer order, and so on. Thus these other approaches enable solutions which are not the most general possible. We will introduce a fractional representation which uses the full conversion of all integer derivatives to fractional order derivatives in all independent coordinates. Since the FP equation is of order one and two and one in time and spatial coordinates, respectively, we have to consider three different fractional differentiation orders β , α , and μ as follows:

$${}^{C}\mathcal{D}_{t}^{\beta}u(x,t) - \mathcal{D}_{\theta_{\alpha};x}^{\alpha}u(x,t) = -\mathcal{D}_{\theta_{\mu};x}^{\mu}\left(\frac{V'(x)}{\epsilon m\eta}u(x,t)\right) \text{ with } -\infty < x < \infty \text{ and } t \ge 0$$
 (9)

Mathematics 2017, 5, 12 3 of 19

Here $0<\beta\leq 1$, $0<\alpha\leq 2$, and $0<\mu\leq 1$ are constraints for the fractional derivatives consistent with the FP equation. The two skewness parameters θ_{α} and θ_{μ} satisfy the conditions $|\theta_{\alpha}|=\min(\alpha,2-\alpha)$ and $|\theta_{\mu}|=\min(\mu,1-\mu)$, respectively. The fractional derivatives for temporal and spatial coordinates are defined in terms of *Caputo* and *Riesz-Feller* derivatives. For details of these definitions see Appendix A.

We begin by considering the Cauchy problem for the (spatially one-dimensional) space-time fractional FP Equation (9). Note that our approximation method is also applicable to spatially higher dimensional cases and can be generalized to systems of FP equations.

We shall first generate the fundamental solution of the IE via application of Laplace and Fourier transforms to the fractional operators. Assuming, $\Re(s) > \sigma_0$, $\kappa \in \mathbb{R}$, the transforms of Laplace and Fourier are defined as:

$$\mathcal{L}(f(t),s) = \tilde{f}(s) = \int_0^\infty f(t)e^{-st}dt \tag{10}$$

and:

$$\mathcal{F}(g(x),\kappa) = \hat{g}(\kappa) = \int_{-\infty}^{\infty} g(x)e^{i\kappa x}dx \tag{11}$$

Thus the corresponding transform of the fractional derivatives ${}^{C}\mathcal{D}_{t}^{\beta}f(t)$ and $\mathcal{D}_{\theta;x}^{\alpha}g(x)$ are:

$$\mathcal{L}\left({}^{C}\mathcal{D}_{t}^{\beta}f(t),s\right) = s^{\beta}\tilde{f}(s) - s^{\beta-1}f(0)$$
(12)

and:

$$\mathcal{F}\left(\mathcal{D}_{\theta,\gamma}^{\alpha}g(x),\kappa\right) = -|\kappa|^{\alpha}i^{\theta\,\mathrm{sign}(\kappa)}\hat{g}(\kappa) = -|\kappa|^{\alpha}e^{i\,\mathrm{sign}(\kappa)\theta\pi/2}\hat{g}(\kappa) \tag{13}$$

The fundamental solution of the fractional diffusion equation in the Fourier-Laplace domain with the initial condition $\hat{u}_0(\kappa) = 1$ thus becomes:

$$\mathcal{F}\left(\mathcal{L}\left({}^{C}\mathcal{D}_{t}^{\beta}u(x,t),s\right),\kappa\right) - \mathcal{F}\left(\mathcal{L}\left(\mathcal{D}_{\theta_{1};x}^{\alpha}u(x,t),s\right),\kappa\right) = 0 \tag{14}$$

which delivers the equations:

$$s^{\beta}\hat{u}(\kappa,s) - s^{\beta-1}\hat{u}(0) = -|\kappa|^{\alpha} e^{i\operatorname{sign}(\kappa)\theta\pi/2}\hat{u}(\kappa,s)$$
(15)

$$\hat{u}(\kappa, s) = \frac{s^{\beta - 1} \hat{u}(0)}{s^{\beta} + |\kappa|^{\alpha} e^{i \operatorname{sign}(\kappa)\theta\pi/2}}$$
(16)

If we use the initial condition $\hat{u}(0) = 1$ then we finally get:

$$\hat{u}(\kappa, s) = \frac{s^{\beta - 1}}{s^{\beta} + |\kappa|^{\alpha} e^{i \operatorname{sign}(\kappa)\theta\pi/2}} = \hat{G}^{\alpha, \beta}(\kappa, s)$$
(17)

representing the Laplace-Fourier solution of the Green's or transfer function for the fractional diffusion equation [15]. In fact the knowledge of the Laplace-Fourier representation of the fractional diffusion equation is sufficient to solve the fractional FP Equation (9) by Sinc convolution approximation.

Next, the Fourier–Laplace of the fFP Equation (9) takes the form:

$$\mathcal{F}\left(\mathcal{L}\left({}^{C}\mathcal{D}_{t}^{\beta}u(x,t),s\right),\kappa\right) - \mathcal{F}\left(\mathcal{L}\left(\mathcal{D}_{\theta_{\alpha};x}^{\alpha}u(x,t),s\right),\kappa\right) = -\mathcal{F}\left(\mathcal{L}\left(\mathcal{D}_{\theta_{\mu};x}^{\mu}(w(x)u(x,t)),s\right),\kappa\right)$$
(18)

Applying Leibniz's rule for fractional derivatives, we write:

$$\mathcal{F}\left(\mathcal{L}\left({}^{C}\mathcal{D}_{t}^{\beta}u(x,t),s\right),\kappa\right) - \mathcal{F}\left(\mathcal{L}\left(\mathcal{D}_{\theta_{\alpha};x}^{\alpha}u(x,t),s\right),\kappa\right)$$

$$= -\mathcal{F}\left(\mathcal{L}\left(w(x)\mathcal{D}_{\theta_{\mu};x}^{\mu}u(x,t) + \mu w^{(1)}(x)\mathcal{D}_{\theta_{\mu};x}^{\mu-1}u(x,t) \mp \dots,s\right),\kappa\right)$$
(19)

Mathematics 2017, 5, 12 4 of 19

The Green's function $G^{\alpha,\beta}(x,t)$ has to satisfy:

$${}^{C}\mathcal{D}_{t}^{\beta}G^{\alpha,\beta}(x,t) - \mathcal{D}_{\theta_{\alpha};x}^{\alpha}G^{\alpha,\beta}(x,t) + \mathcal{D}_{\theta_{\mu};x}^{\mu}\left(w(x)G^{\alpha,\beta}(x,t)\right) = \delta(t)\delta(x) \tag{20}$$

with $-\infty < x < \infty$ and $t \ge 0$, which applied to (19) yields:

$${}^{C}\mathcal{D}_{t}^{\beta}G^{\alpha,\beta}(x,t) - \mathcal{D}_{\theta_{x}:x}^{\alpha}G^{\alpha,\beta}(x,t) + w(x)\mathcal{D}_{\theta_{y}:x}^{\mu}\left(G^{\alpha,\beta}(x,t)\right) + \mu w^{(1)}(x)\mathcal{D}_{\theta_{y}:x}^{\mu-1}\left(G^{\alpha,\beta}(x,t)\right) \mp \ldots = \delta(t)\delta(x) \quad (21)$$

Assuming that $w^{(1)}$ is negligible, relative to w, and since the Green's function satisfies:

$${}^{C}\mathcal{D}_{t}^{\beta}G^{\alpha,\beta}(x,t) - \mathcal{D}_{\theta_{\alpha};x}^{\alpha}G^{\alpha,\beta}(x,t) + w(x)\mathcal{D}_{\theta_{\mu};x}^{\mu}\left(G^{\alpha,\beta}(x,t)\right) \approx \delta(t)\delta(x) \tag{22}$$

which can be further simplified if we use a Taylor expansion of w around x = 0:

$${}^{C}\mathcal{D}_{t}^{\beta}G^{\alpha,\beta}(x,t) - \mathcal{D}_{\theta_{\alpha};x}^{\alpha}G^{\alpha,\beta}(x,t) + \left(w(0) + w^{(1)}(0)x + \ldots\right)\mathcal{D}_{\theta_{\mu};x}^{\mu}\left(G^{\alpha,\beta}(x,t)\right) \approx \delta(t)\delta(x) \tag{23}$$

The major contribution now is assigned to the constant term w(0) so that:

$${}^{C}\mathcal{D}_{t}^{\beta}G^{\alpha,\beta}(x,t) - \mathcal{D}_{\theta_{\alpha};x}^{\alpha}G^{\alpha,\beta}(x,t) + w(0)\mathcal{D}_{\theta_{\mu};x}^{\mu}\left(G^{\alpha,\beta}(x,t)\right) \approx \delta(t)\delta(x) \tag{24}$$

Equation (24) delivers when Fourier-Laplace transforms are applied to the determining equation for $G^{\alpha,\beta}$:

$$s^{\beta}\hat{G}^{\alpha,\beta}(\kappa,s) - s^{\beta-1}\hat{G}^{\alpha,\beta}(0) = -|\kappa|^{\alpha} e^{i\operatorname{sign}(\kappa)\theta_{\alpha}\pi/2} \hat{G}^{\alpha,\beta}(\kappa,s) + w(0) |\kappa|^{\mu} e^{i\operatorname{sign}(\kappa)\theta_{\mu}\pi/2} \hat{G}^{\alpha,\beta}(\kappa,s)$$
(25)

Under the assumption that the major contribution is generated by u itself in the Fourier-Laplace space, we approximate w(0) = 1 which then yields:

$$s^{\beta} \hat{G}^{\alpha,\beta}(\kappa,s) - s^{\beta-1} \hat{G}^{\alpha,\beta}(0) \approx -|\kappa|^{\alpha} e^{i\operatorname{sign}(\kappa)\theta_{\alpha}\pi/2} \hat{G}^{\alpha,\beta}(\kappa,s) + |\kappa|^{\mu} e^{i\operatorname{sign}(\kappa)\theta_{\mu}\pi/2} \hat{G}^{\alpha,\beta}(\kappa,s)$$
(26)

Solving with respect to the Fourier-Laplace variable we get:

$$\hat{G}^{\alpha,\beta}(\kappa,s) = \frac{s^{\beta-1}\hat{G}^{\alpha,\beta}(0)}{s^{\beta} + |\kappa|^{\alpha}e^{i\operatorname{sign}(\kappa)\theta_{\alpha}\pi/2} - |\kappa|^{\mu}e^{i\operatorname{sign}(\kappa)\theta_{\mu}\pi/2}}$$
(27)

and setting $\hat{G}^{\alpha,\beta}(0) = 1$, we finally get:

$$\hat{G}^{\alpha,\beta}(\kappa,s) = \frac{s^{\beta-1}}{s^{\beta} + |\kappa|^{\alpha} e^{i\operatorname{sign}(\kappa)\theta_{\alpha}\pi/2} - |\kappa|^{\mu} e^{i\operatorname{sign}(\kappa)\theta_{\mu}\pi/2}}$$
(28)

Here, α as a superscript of \hat{G} denotes the collection of the parameters $\alpha = \{(\alpha, \theta_{\alpha}), (\mu, \theta_{\mu})\}$. This transfer/Green function collects the basic information about the fractional FP equation. It will be used in the discrete convolution representation to approximate the fractional IE:

$$u(x,t) = v(x,t) - \int_0^t \int_{\mathbb{R}} G_x^{\alpha,\beta}(x-\xi,t-\tau)w(\xi)u(\xi,\tau)d\xi d\tau$$
 (29)

incorporating the initial condition:

$$v(x,t) = \int_{\mathbb{R}} G(x-\xi,t)u^{(0)}(\xi)d\xi \tag{30}$$

with $w(x) = V'(x)/(\epsilon m\eta)$.

Mathematics 2017, 5, 12 5 of 19

Comparing the spectral representation of the fFP equation with the Montrol-Weiss equation used for continuous time random walks (CTRW) [15], we identify the transfer function by means of the expression:

$$\hat{u}(\kappa, s) \approx \frac{\Psi(s)}{1 - \Phi(\kappa, s)} \tag{31}$$

where $\Psi(s)=(1-\psi(s))/s$ and $\Phi(\kappa,s)=\hat{W}(\kappa)\psi(s)$ with $\psi(s)=1-s^{\beta}$ representing the asymptotic scaling behavior for $s\to 0$. The separation of Fourier and Laplace variables in $\hat{B}^{\alpha,\beta}(\kappa,s)=\hat{B}^{\alpha}(\kappa)\psi^{1/2}(s)$ and $\hat{B}^{\mu,\beta}(\kappa,s)=\hat{B}^{\mu}(\kappa)\psi^{1/2}(s)$ yields $\Phi(\kappa,s)=\hat{W}(\kappa)\psi(s)=\hat{B}^{\alpha}(\kappa)\hat{B}^{\mu}(\kappa)\psi(s)$:

$$\hat{u}(\kappa, s) \approx \frac{s^{\beta - 1}}{s^{\beta} - \hat{B}^{\alpha}(\kappa) + \hat{w}(\kappa)\hat{B}^{\mu}(\kappa)}$$
(32)

In (32), $\hat{B}^{\gamma}(\kappa) \approx -|\kappa|^{\gamma} e^{i\operatorname{sign}(\kappa)\theta_{\gamma}\pi/2}$ represents the asymptotic scaling law in Fourier space as $\kappa \to 0$, with $\gamma \in \{\alpha, \mu\}$. If in addition we also assume that $\hat{w}(k) = \hat{w}(0) + \hat{w}'(0)\kappa + \ldots$ as $\kappa \to 0$, we set $\hat{w}(k) = \hat{w}(0) = 1$ as a proper normalization. Thus the fundamental solution in Fourier-Laplace representation becomes:

$$\hat{u}(\kappa, s) \approx \frac{s^{\beta - 1}}{s^{\beta} + |\kappa|^{\alpha} e^{i \operatorname{sign}(\kappa)\theta_{\alpha}\pi/2} - |\kappa|^{\mu} e^{i \operatorname{sign}(\kappa)\theta_{\mu}\pi/2}}$$
(33)

corresponding to a CTRW equation of the type:

$$u(x,t) = \delta(x)\Psi(t) + \int_0^t \psi(t-\tau) \left[\int_{\mathbb{R}} W(x-\xi)u(\xi,\tau)d\xi \right] d\tau$$
 (34)

with the survival function $\Psi(t) = \int_t^\infty \psi(\tau) d\tau$ denoting the probability that at instant t the particle is still sitting in its starting position x = 0. The transfer function $\hat{u}(\kappa, s)$ in (33) will be the key expression in the solution of the fFP equation:

$${}^{C}\mathcal{D}_{t}^{\beta}u(x,t) - \mathcal{D}_{\theta_{\alpha};x}^{\alpha}u(x,t) = -\mathcal{D}_{\theta_{\mu};x}^{\mu}\left(\frac{V'(x)}{\epsilon m\eta}u(x,t)\right) \text{ with } -\infty < x < \infty \text{ and } t \ge 0$$
 (35)

The equivalent IE to (35) is numerically approximated by a successive Neumann iteration. Examples of fractional integral equations are solved for single valued functions in [16,17]. The work in [16] proves that for fractional integral equations the solution exists and converges exponentially under the condition that the functions are analytic. Using the property of separation of variables in convolution representations the results for single valued functions are also valid for multivalued functions [18]. Thus the existence of solutions and convergence of the numerical approximation to the solution is guaranteed (for details see [16,18]). Based on these facts, we shall next introduce Sinc methods to approximate fractional partial differential equations specifically the fFP equation.

2. Methods of Approximation

This section introduced the basic ideas of Sinc methods [19]. We will discuss only the main ideas as a collection of recipes to set up a Sinc approximation. We omit most of the proofs of the different important theorems and lemmas because these proofs are available in literature [6,18,20–22]. The following subsections collect information on the basic mathematical functions used in Sinc approximation. We introduce Sinc methods to represent indefinite integrations and convolution integrals. These types of integrals are essential for representing the fractional operators of differentiation and integration [7,18].

Mathematics 2017, 5, 12 6 of 19

2.1. Sinc Basis

To start with we first introduce some definitions and theorems allowing us to specify the space of functions, domains, and arcs for a Sinc approximation.

Definition 1. Domain and Conditions

Let \mathcal{D} be a simply connected domain in the complex plane and $z \in \mathbb{C}$ having a boundary $\partial \mathcal{D}$. Let a and b denote two distinct points of $\partial \mathcal{D}$ and ϕ denote a conformal map of \mathcal{D} onto \mathcal{D}_d , where $\mathcal{D}_d = \{z \in \mathbb{C} : |\mathcal{I}m(z)| < d\}$, such that $\phi(a) = -\infty$ and $\phi(b) = \infty$. Let $\psi = \phi^{-1}$ denote the inverse conformal map, and let Γ be an arc defined by $\Gamma = \{z \in \mathbb{C} : z = \psi(x), x \in \mathbb{R}\}$. Given ϕ , ψ , and a positive number h, let us set $z_k = \psi(kh)$, $k \in \mathbb{Z}$ to be the Sinc points, let us also define $\rho(z) = e^{\phi(z)}$.

Note the Sinc points are an optimal choice of approximation points in the sense of Lebesgue measures for Sinc approximations [23].

Definition 2. Function Space

Let $d \in (0, \pi)$, and let the domains \mathcal{D} and \mathcal{D}_d be given as in Definition 1. If d' is a number such that d' > d, and if the function ϕ provides a conformal map of \mathcal{D}' onto $\mathcal{D}_{d'}$, then $\mathcal{D} \subset \mathcal{D}'$. Let α and β denote positive numbers, and let $\mathbf{L}_{\alpha,\beta}(\mathcal{D})$ denote the family of functions $u \in \mathbf{Hol}(\mathcal{D})$, for which there exists a positive constant c_1 such that, for all $z \in \mathcal{D}$:

$$|u(z)| \le c_1 \frac{|\rho(z)|^{\alpha}}{(1+|\rho(z)|)^{\alpha+\beta}}$$
 (36)

Now let the positive numbers α and β belong to (0,1], and let $\mathbf{M}_{\alpha,\beta}(\mathcal{D})$ denote the family of all functions $g \in \mathbf{Hol}(\mathcal{D})$, such that g(a) and g(b) are finite numbers, where $g(a) = \lim_{z \to a} g(z)$ and $g(b) = \lim_{z \to b} g(z)$, and such that $u \in \mathbf{L}_{\alpha,\beta}(\mathcal{D})$ where:

$$u(z) = g(z) - \frac{g(a) + \rho(z)g(b)}{1 + \rho(z)}$$
(37)

The two definitions allow us to formulate the following theorem for Sinc approximations.

Theorem 1. *Sinc Approximation* [20]

Let $u \in \mathbf{L}_{\alpha,\beta}(\mathcal{D})$ for $\alpha > 0$ and $\beta > 0$, take $M = [\beta \ N/\alpha]$, where [x] denotes the greatest integer in x, and then set m = M + N + 1. If $u \in \mathbf{M}_{\alpha,\beta}(\mathcal{D})$, and if $h = (\pi d/(\beta N))^{1/2}$ then there exists a positive constant c_2 independent of N, such that:

$$\left| u(z) - \sum_{k=-M}^{N} u(z_k) w_k \right| \le c_2 N^{1/2} e^{-(\pi d\beta N)^{1/2}}$$
(38)

with w_k the base function (see (41)).

The proof of this Theorem 1 is given in [20]. Note the choice $h = (\pi d/(\beta N))^{1/2}$ is close to optimal for an approximation in the space $\mathbf{M}_{\alpha,\beta}(\mathcal{D})$ in the sense that the error bound in Theorem 1 cannot be appreciably improved regardless of the basis [20]. It is also optimal in the sense of the Lebesgue measure achieving an optimal value less than Chebyshev approximations [23]. We also note that this behavior is found also in bi-variate approximations based on Poly-Sinc methods [24–26].

These definitions and the theorem directly allow the formulation of an algorithm for a Sinc approximation. Let $\mathbb Z$ denote the set of all integers. Select positive integers N and $M = [\beta N/\alpha]$ and set m = M + N + 1. The step length is determined by $h = (\pi d/(\beta N))^{1/2}$ where α , β , and d are real parameters. In addition assume there is a conformal map ϕ and its inverse ψ such that we can define Sinc points $z_j = \psi(jh)$, $j \in \mathbb Z$ [20]. The following relations define the basis of a Sinc approximation:

Mathematics 2017, 5, 12 7 of 19

$$\operatorname{Sinc}(z) = \frac{\sin(\pi z)}{\pi z} \tag{39}$$

The shifted Sinc is derived from relation (39) by translating the argument by integer steps of length h and applying the conformal map to the independent variable:

$$S(j,h) \circ \phi(z) = \operatorname{Sinc}([\phi(z) - jh]/h), \ j = -M, \dots, N$$
(40)

The discrete shifting allows us to cover the approximation interval (a, b) in a dense way while the conformal map is used to map the interval of approximation from an infinite range of values to a finite one. Using the Sinc basis we are able to represent the basis functions as a piecewise defined function $w_i(z)$ by:

$$w_{j} = \begin{cases} \frac{1}{1+\rho(z)} - \sum_{k=-M+1}^{N} \frac{1}{1+e^{kh}} S(k,h) \circ \phi(z) & j = -M \\ S(j,h) \circ \phi(z) & j = -M+1, \dots, N-1 \\ \frac{\rho(z)}{1+\rho(z)} - \sum_{k=-M}^{N-1} \frac{e^{kh}}{1+e^{kh}} S(k,h) \circ \phi(z) & j = N \end{cases}$$
(41)

where $\rho(z) = \exp(\phi(z))$. This form of the Sinc basis is chosen as to satisfy the interpolation at the boundaries. The basis functions defined in (41) suffice for purposes of uniform—norm approximation over (a,b).

This notation allows us to define a row vector $V_m(S)$ of basis functions:

$$V_m(S) = (w_{-M}, \dots, w_N) \tag{42}$$

with w_j defined as in (41). For a given vector $V_m(u) = (u_{-M}, \dots, u_N)^T$ we now introduce the product as an approximation of the function u(z) by:

$$u(z) \approx V_m(S)V_m(u) = \sum_{k=-M}^{N} u_k w_k$$
(43)

Based on this notation, we will introduce in the next few subsections the different integrals we need [18].

2.2. Indefinite Integral

We need for the representation of fractional derivatives indefinite integrals. This subsection describes how indefinite integrals can be defined [20] and how these definitions are related to the definition of definite integrals. For collocating an indefinite integral and for obtaining explicit approximations of the functions $(\mathcal{J}u)(x)$ and $(\mathcal{J}'u)(x)$ defined by:

$$(\mathcal{J}u)(x) = \int_{a}^{x} u(t)dt \text{ with } x \in (a,b)$$
(44)

$$(\mathcal{J}'u)(x) = \int_{x}^{b} u(t)dt \text{ with } x \in (a,b)$$
(45)

We use the following basic relations [20]. Let \mathbb{Z} denote the set of all integers, and let \mathbb{C} denote the complex plane. Let $\operatorname{Sinc}(x)$ be given by (40) and e_k be defined as:

$$\sigma_k = \int_0^k \operatorname{Sinc}(x) \, dx = \frac{1}{\pi} \operatorname{Si}(\pi k) \tag{46}$$

with Si(x) the sine integral. This put us into position to write:

$$e_k = \frac{1}{2} + \sigma_k, \ k \in \mathbb{Z} \tag{47}$$

Mathematics 2017, 5, 12 8 of 19

If now ϕ denotes a one—to—one transformation of the interval (a,b) onto the real line \mathbb{R} , let h denote a fixed positive number, and let the Sinc points be defined on (a,b) by $z_k = \phi^{-1}(kh)$, $k \in \mathbb{Z}$, where ϕ^{-1} denotes the inverse function of the conformal map ϕ . Let M and N be positive integers, set m = M + N + 1, and for a given function u defined on (a,b), define a diagonal matrix D(u) by $D(u) = \operatorname{diag}\left[u\left(z_{-M}\right), \ldots, u\left(z_{N}\right)\right]$. Let $I^{(-1)}$ be a square Töplitz matrix of order m having e_{i-j} , as its $(i,j)^{\text{th}}$ entry i.e., $i,j = -M, \ldots, N$:

$$(I^{(-1)})_{i,j} = e_{i-j} \text{ with } i, j = -M, \dots, N$$
 (48)

Define square matrices A_m and B_m by:

$$A_m = hI^{(-1)}D(1/\phi') \tag{49}$$

$$B_m = h \left\{ I^{(-1)} \right\}^T D(1/\phi') \tag{50}$$

where the superscript "T" denotes the transpose. The collocated representation of the indefinite integrals are thus given by:

$$\mathcal{J}_m u = V_m(S) A_m V_m(u) = h V_m(S) I^{(-1)} D(1/\phi') V_m(u)$$
(51)

$$\mathcal{J}'_{m}u = V_{m}(S)B_{m}V_{m}(u) = hV_{m}(S)\left\{I^{(-1)}\right\}^{T}D(1/\phi')V_{m}(u)$$
(52)

These are collocated representations of the indefinite integrals defined in (44) and (45), respectively [18]. If we compare our results with the target equations for example (35) we observe that we need one additional step to extend the indefinite integrals to convolution integrals.

2.3. Convolution Integrals

Indefinite convolution integrals can also be effectively collocated via Sinc methods [20]. This section discusses the core procedure of this paper, for collocating the convolution integrals and for obtaining explicit approximations of the functions p and q defined by:

$$p(x) = \int_{a}^{x} f(x - t)g(t)dt \text{ with } x \in (a, b)$$
(53)

$$q(x) = \int_{x}^{b} f(t - x)g(t)dt \text{ with } x \in (a, b)$$
(54)

where $x \in \Gamma$. In presenting these convolution results, we shall assume that $\Gamma = (a,b) \subseteq \mathbb{R}$, unless otherwise indicated. Note also, that being able to collocate p and q enables us to collocate definite convolutions like $\int_a^b f(x-t)g(t)dt$.

Before we start to present the collocation of the Equations (53) and (54) we mention that there is a special approach to evaluate the convolution integrals by using a Laplace transform. In fact Lubich [27,28] introduced this way of calculation by the following idea:

$$p(x) = \int_0^x f(x - t)g(t)dt = \frac{1}{2\pi i} \int_{\mathcal{C}} F_+(s) \int_0^x e^{st} g(x - t)dtds$$
 (55)

for which the inner integral solves the initial value problem y' = sy + g with g(0) = 0. We assume that the Laplace transform (Stenger-Laplace transform):

$$F_{+}(s) = \int_{E} f(t)e^{-t/s}dt$$
 (56)

Mathematics 2017, 5, 12 9 of 19

with E any subset of \mathbb{R} such that $E \supseteq (0, b - a)$, exists for all $s \in \Omega^+ = \{s \in \mathbb{C} | \Re(s) > 0\}$. The eigenvalues of A_m and B_m are all positive which was a 20 year old conjecture by FS. This conjecture was recently solved by Han and Xu [29].

In the notation introduced above we get:

$$p = F_{+}(\mathcal{J})g \approx F_{+}(\mathcal{J}_{m})g \tag{57}$$

and:

$$q = F_{+}(\mathcal{J}')g \approx F_{+}(\mathcal{J}'_{m})g \tag{58}$$

are accurate approximations, provided that the inverse Laplace transform of $F_+(1/s)g(1/s)$ again belongs to a space $M_{\alpha',d'}$ for some positive numbers α' and d' [18]. The procedure to calculate the convolution integrals is now as follows. The collocated integral $\mathcal{J}_m = V_m(S)A_mV_m$ and $\mathcal{J}'_m = V_m(S)B_mV_m$, upon diagonalization of A_m and B_m in the form:

$$A_m = X_m \text{diag}[s_{m,-M}, \dots, s_{m,N}] X_m^{-1}$$
(59)

$$B_m = Y_m \text{diag}[s_{m,-M}, \dots, s_{m,N}] Y_m^{-1}$$
 (60)

with $\Sigma = \text{diag}[s_{-M}, \dots, s_N]$ as the eigenvalues arranged in a diagonal matrix for each of the matrices A_m and B_m . Note that A_m and B_m have the same vector of eigenvalues. Then the Stenger-Laplace transform (56) delivers the square matrices F_+ (A_m) and F_+ (B_m) defined via the equations:

$$F_{+}(A_{m}) = X_{m} \operatorname{diag}\left[F_{+}(s_{m,-M}), \dots, F_{+}(s_{m,N})\right] X_{m}^{-1} = X_{m} F_{+}(\Sigma) X_{m}^{-1}$$
(61)

$$F_{+}(B_{m}) = Y_{m} \operatorname{diag}\left[F_{+}(s_{m,-M}), \dots, F_{+}(s_{m,N})\right] Y_{m}^{-1} = Y_{m} F_{+}(\Sigma) Y_{m}^{-1}$$
(62)

We can get the approximation of (57) and (58) by:

$$F_{+}(\mathcal{J})g \approx F_{+}(\mathcal{J}_{m})g = V_{m}(S)F_{+}(A_{m})V_{m}(g) = V_{m}(S)X_{m}F_{+}(\Sigma)X_{m}^{-1}V_{m}(g)$$
(63)

$$F_{+}(\mathcal{J}^{*})g \approx F_{+}(\mathcal{J}_{m}^{*})g = V_{m}(S)F_{+}(B_{m})V_{m}(g) = V_{m}(S)Y_{m}F_{+}(\Sigma)Y_{m}^{-1}V_{m}(g)$$
(64)

These two formulas deliver a finite approximation of the convolution integrals p and q. The convergence of the method is exponential under the above Laplace transform assumptions—see [20] for a proof of a special case.

2.4. Convolution Integrals in Two Variables

Examining the IE (35) in more detail we realize that we have to extend the one dimensional convolution to a two dimensional one in fact we have to deal with integrals of the type:

$$p(x,t) = \int_0^t \int_{-\infty}^{\infty} f(x-\xi,t-\tau)g(\xi,\tau)d\xi d\tau$$

$$= \int_0^t \int_{-\infty}^x f(x-\xi,t-\tau)g(\xi,\tau)d\xi d\tau + \int_0^t \int_x^{\infty} f(x-\xi,t-\tau)g(\xi,\tau)d\xi d\tau$$
 (65)

As discussed in the previous section we will use a Laplace representation of the convolution integrals but now in two dimensions. Where the approximation is sought over the region $(a_1,b_1)\times(a_2,b_2)$, and with $(a_i,b_i)\in\mathbb{R}$, i=1,2. Note that the method of separation of variables made possible by the one-dimensional convolution approximation, as well as to how easily the final algorithm can be adapted to parallel computation. In order to guarantee some accuracy in the final approximation, we shall simply assume, without going into detail, that the function p belongs to the class $\mathcal{M}_{\alpha_i,\beta_i}$ with respect to each variable x_j , for all fixed values of the other variables, each in its respective interval of definition, j=1,2.

Mathematics 2017, 5, 12 10 of 19

We shall assume that the mappings ϕ_j have been selected properly according to the limits of the intervals. We furthermore assume that positive integers N_j and M_j as well as positive numbers h_j (j=1,2) have been selected such that $h_j=\pi/\sqrt{N_j}$. These definitions ensure that we get the same order of accuracy of approximation $\epsilon=\mathcal{O}\left(N_1^{1/2}\exp\left(-cN_1^{1/2}\right)\right)$ in each variable [16,18]. We set $m_j=M_j+N_j+1$, and we define the Sinc points by $z_l^{(j)}=\phi_j^{-1}\left(lh_j\right)$ for $l=-M_j,\ldots,N_j; j=1,2$. Next, we determine matrices A_j,X_j,S_j , and X_j^{-1} such that:

$$A_j = h_j I_{m_j}^{(-1)} D\left(1/\phi_j'\right) = X_j S_j X_j^{-1}, \ j = 1, 2$$
 (66)

In (66) $I_{m_i}^{(-1)}$ is defined as in (48), and S_j are diagonal matrices:

$$S_{j} = \operatorname{diag}\left[s_{-M_{j}}^{(j)}, \dots, s_{N_{j}}^{(j)}\right] \tag{67}$$

Arbitrarily taking $c_i \in [2(b_i - a_i), \infty)$, we set:

$$F\left(s^{(1)}, x\right) = \int_0^{c_1} f(t, x) e^{-t/s^{(1)}} dt \tag{68}$$

$$G\left(s^{(1)}, s^{(2)}\right) = \int_0^{c_2} F\left(s^{(1)}, x\right) e^{-x/s^{(2)}} dx \tag{69}$$

We now illustrate the method of separation of variables. To this end, we first rewrite (65) in the notationally more convenient form:

$$p_1(x,t) = \int_{a_1}^t \int_{a_2}^x f(x-\xi,t-\tau)g(\xi,\tau)d\xi d\tau$$
 (70)

representing the first integral in (65). Discretization with respect to t. We set:

$$g(\xi,\tau) = \left(g\left(\xi, z_{-M_1}^{(1)}\right), \dots, g\left(\xi, z_{N_1}^{(1)}\right)\right)^T \tag{71}$$

and we then define a vector p(x) by:

$$\boldsymbol{p}(x) = \int_{a_2}^{x} F(x - \xi, A_1) \, \boldsymbol{g}(\xi) d\xi \tag{72}$$

where A_1 and F are defined in (66) and (68) respectively. Using the diagonalization identity $A_1 = X_1 S_1 X_1^{-1}$ given in (66), it now follows from (72) that:

$$\mathbf{p}(x) = X_1 \int_{a_2}^{x} \begin{pmatrix} F\left(x - \xi, s_{-M_1}^{(1)}\right) & & \\ & \ddots & \\ & & F\left(x - \xi, s_{N_1}^{(1)}\right) \end{pmatrix} X_1^{-1} \mathbf{g}(\xi) d\xi$$
 (73)

The expression (73) motivates the transformations:

$$h(\xi) = X_1^{-1} g(\xi), q(x) = X_1^{-1} p(x)$$
 (74)

Thus, denoting the components of \boldsymbol{h} and \boldsymbol{q} by h_i and q_1 respectively, with $i = -M_1, \dots, N_1$, Equation (73) reduces to the decoupled set of scalar equations:

$$q_i(x) = \int_{a_2}^x F\left(x - \xi, s_i^{(1)}\right) h_i(\xi) d\xi \tag{75}$$

Mathematics 2017, 5, 12 11 of 19

Discretization with respect to x. We now set:

$$\boldsymbol{h}_{i} = \left(h_{i}\left(z_{-M_{2}}^{(2)}\right), \dots, h_{i}\left(z_{N_{2}}^{(2)}\right)\right)^{T} \tag{76}$$

and we then define a vector q_i by:

$$\mathbf{q}_{i}(x) = \int_{a_{2}}^{x} G\left(A_{2}, s_{i}^{(1)}\right) \mathbf{h}_{i}(\xi) d\xi \tag{77}$$

where A_2 and G are defined in (66) and (69) respectively. Using the diagonalization identity $A_2 = X_2 S_2 X_2^{-1}$ given in (66), it now follows from (77) that:

$$\mathbf{q}_{i}(x) = X_{2} \int_{a_{2}}^{x} \begin{pmatrix} G\left(s_{-M_{2}}^{(2)}, s_{i}^{(1)}\right) & & \\ & \ddots & \\ & & G\left(s_{N_{2}}^{(2)}, s_{i}^{(1)}\right) \end{pmatrix} X_{2}^{-1} \mathbf{h}_{i}(\xi) d\xi$$
(78)

This last expression motivates the transformations:

$$\mathbf{k}_i = X_2^{-1} \mathbf{h}_i(\xi), \ \mathbf{r}_i = X_2^{-1} \mathbf{q}_i(\xi)$$
 (79)

Denoting the components of \mathbf{r}_i and \mathbf{k}_i by $r_{i,j}$ and $k_{i,j}$ respectively, with $i = -M_1, \dots, N_1$ and $j = -M_2, \dots, N_2$, Eq. (78) reduces to the decoupled set of scalar equations:

$$r_{i,j} = G\left(s_j^{(2)}, s_i^{(1)}\right) k_{i,j}$$
 (80)

By assumption, the $k_{i,j}$ are known at this point, and (80) then determines the $r_{i,j}$. The second equation in (79) is then used to determine the vectors \mathbf{q}_i at the Sinc points $z^{(2)}$. The second transformation of (74) is next used to determine the vector $\mathbf{p}(x)$ at the set of Sinc points. We can thus recover the complete array of values p(x,t) at the set of Sinc points $\left(z_i^{(1)}, z_j^{(2)}\right)$. The algorithmic procedure is described in more detail in [21].

2.5. Sinc Collocation of fFP

For fractional differential equations there exists a solution in terms of a second kind Volterra integral equation incorporating the initial conditions of the Cauchy problem as given in (29) and (30) in the representation of (34) connected to (33). These relations are equivalent to the fFP Equation (35). The discrete version of Equation (34) follow as a collocation approximation as:

$$u - \lambda F_{+}(\mathcal{J})u = \mathcal{N} \approx V_{m}(u) - \lambda_{1}F_{+}\left(A_{m}^{t}, A_{m}^{x}\right)V_{m}(u) - \lambda_{2}F_{+}\left(A_{m}^{t}, B_{m}^{x}\right)V_{m}(u) = V_{m}(\mathcal{N}) \tag{81}$$

Here, \mathcal{N} collects the given initial conditions of the problem and $\lambda_1 = \sin((-\alpha + \theta)\pi/2)/(\sin(-\alpha\pi)\Gamma(-\alpha)\Gamma(\beta))$, $\lambda_2 = \sin((-\alpha - \theta)\pi/2)/(\sin(-\alpha\pi)\Gamma(-\alpha)\Gamma(\beta))$ are factors independent of t or x related to the *Caputo* and *Riesz-Feller* operators (see Appendix A). The discrete version follows straight forward from (81) to be:

$$\left\{I - \lambda_1 F_+ \left(A_m^t, A_m^x\right) - \lambda_2 F_+ \left(A_m^t, B_m^x\right)\right\} V_m(u) = V_m(\mathcal{N})$$
(82)

The matrices F_+ are set up according to the discussion in Section 2.4 using the transfer function (33) in the Laplace transform representation. Solving this linear system with respect to $V_m(u)$ allows us to approximate the solution by:

$$u(t,x) \approx V_m(S)V_m(u) \tag{83}$$

Mathematics 2017, 5, 12 12 of 19

Note the solution of (82) is generated using a Neumann iteration applying a trash hold level for two consecutive approximations in view of generalizing the linear equation to a nonlinear one [16].

3. Experimental Section

This section collects four examples for the fFP equation using different types of external forces and different values for fractional derivatives. We also demonstrate the influence of the skewness parameters to exemplify the effects on the approximations such as the counter gradient transport [30]. The first three examples use a normalized Gaussian distribution with different amplitudes. Since the Sinc approximation allows an arbitrary initial distribution, we examine in the fourth example the influence of the initial distribution on the solution.

Example 1. The first example uses a normalized Gaussian as initial distribution. The distribution follows from $u^{(0)}(x) = \sqrt{b}e^{-bx^2}/\sqrt{\pi}$ with b=4. The initial state is shown in Figure 1. The approximation on the time domain [0,T]=[0,2.0] was iteratively generated using a Neumann iteration scheme. The external force w(x) was set to a constant value w(x)=-1. Figure 1 collects the initial distribution, a sequence of temporal steps represented on the spatial domain, and a(x,t) graph of u(x,t). The level lines of this graphs are also shown in a contour plot.

The graphs in Figure 1 demonstrate that the approximations represent the expected dispersive behavior. The initially localized distribution starts to spread with reduced amplitudes for times larger than zero. The dispersion is visible by the broadening of the distribution if we step along the time axis the amplitude u(x,t) is reduced. We also observe that the function is positive; i.e., $u(x,t) \geq 0$.

Example 2. The second example uses the same fractional parameters β , α , and μ except $\theta_{\alpha}=0.21$ and $\theta_{\mu}=0$. Figure 2 also includes a graph showing the convergence of the Neumann iteration of the IE (top left). The specified error level ϵ is reached within four iterations. The graphs in Figure 2 show again dispersive behavior of the approximation (broadening of the distribution and reduction of the amplitude). It becomes obvious that the decay of the amplitude is slower than that in Example 1 (see Figure 1). In addition we observe that the decaying maximum in time is shifted to the left in the graphs and the symmetry of the solution is distorted. The finite value of θ_{α} is responsible for this effect. Again the distribution is positive on the whole domain. Figure 2 also shows how the time evolution of the distribution u(x,t) as 3d and contour plot. We observe that due to the finite value of θ_{α} the symmetry of the solution is broken and that a steep ramp on the right of the approximation exists. The left side of the solution decays in the spatial direction much more slower than that on the steep ramp. Over all, the dispersion of the solution is present but occurs mainly on the left side of the maximum of the distribution. The maximum location of the distribution follows an asymptotic relation like $\sim t^{\alpha}$.

Example 3. The third example uses the same initial condition and fractional parameters as that in Example 2 (see Figure 2) except for $\theta_{\mu} = -0.1$. The second skewness parameter is set to a negative value which affects the decay rate and the shifting of the maximum of the distribution. The dispersion rate is also affected by this second skewness parameter (see Figure 3).

In Figure 4 we collect some characteristics of the approximations generated in Example 1,2, and 3. The data shown in Figure 4 represent the location of the maximum x_0 of u(x,t), the decay of the maximum value in time, and the change of the width Δx of the maximum on the level $u_{max}(1-e)/e$ corresponding to the 63% value of the maximum. The three graphs summarize the scaling properties of the approximations. The first graph (top row) shows that the maximum location is shifted in time if the skewness parameters are different from zero. The shift of the maximum follows a relation $x_0 \sim t^{0.28}$ for $\theta_\alpha = 0.21$ and $\theta_\mu = -0.1$. As a reference for zero skewness parameters there is no shift of the maximum (top line in Figure 4 top row). The second characteristic of the approximation is the decay of the maximum in time. The decay follows the relation $u_{max} \sim exp\left(-\delta t^{0.1}\right)$. Note the exponent of t is the same for all approximations. The width Δx of the maximum also changes with time according to the dispersion. For vanishing skewness we have a relation $\Delta x \sim t^{0.67}$ while for the skewness $\theta_\alpha = 0.21$ and $\theta_\mu = -0.1$ we found the relation $\Delta x \sim t^{0.21}$. This indicates that the dispersion process is influenced by the skewness parameters.

Mathematics 2017, 5, 12

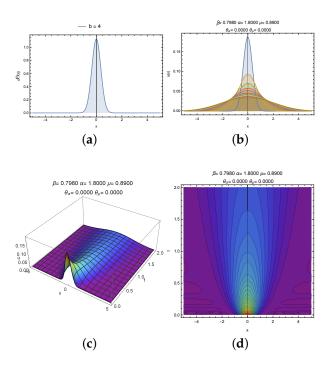


Figure 1. (a) Initial distribution $u^{(0)}(x) = 2/\sqrt{\pi} e^{-4x^2}$ used in the approximation of the fFP equation (35); (b) approximation of fFP for different times $(t \in \{0.1, 0.3, 0.5, 0.7, 0.9, 1.1, 1.3, 1.5, 1.7, 1.9\}$, top down) using a normalized Gaussian with amplitude b = 4; (c) 3D graph of the approximation; and (d) contour plot of the approximation. The fractional exponent are given on top of the graphs (b), (c), and (d). Skewness parameters are zero resulting to a symmetric representation.

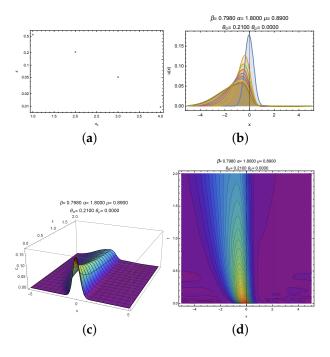


Figure 2. (a) Top left panel shows the convergence of the Neumann iteration toward a fixed error level ϵ ; (b) approximation of fFP for different times ($t \in \{0.1, 0.3, 0.5, 0.7, 0.9, 1.1, 1.3, 1.5, 1.7, 1.9\}$, top down) using a normalized Gaussian with amplitude b = 4; (c) 3D graph of the approximation; and (d) contour plot of the approximation. The fractional exponent are given on top of the graphs (b), (c), and (d).

Mathematics 2017, 5, 12 14 of 19

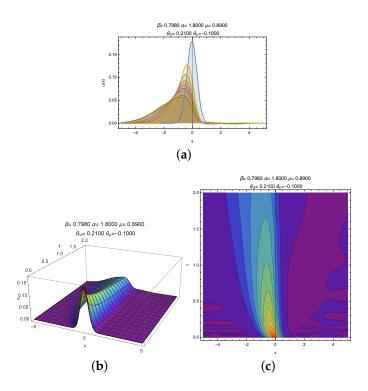


Figure 3. (a) Approximation of fFP for different times ($t \in \{0.1, 0.3, 0.5, 0.7, 0.9, 1.1, 1.3, 1.5, 1.7, 1.9\}$, top down) using a normalized Gaussian with amplitude b = 4; (b) 3D graph of the approximation; and (c) contour plot of the approximation. The fractional exponent are given on top of the graphs.

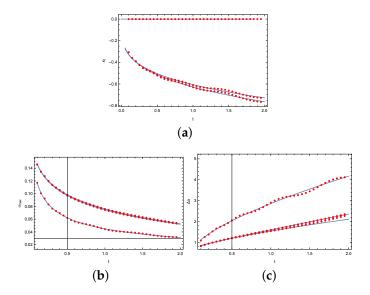


Figure 4. (a) Location of the maximum x_0 as a function of time. The top curve shows the location of the maximum where the skewness parameters are zero. The lower two curves represent the skewness for $\theta_{\alpha}=0.21$ and $\theta_{\mu}=-0.1$. The bending of the curve follows a relation $x_0\sim t^{\sigma}$ with $\sigma\approx 0.28$; (b) decay of the maximum amplitude satisfies the relation $u_{\rm max}\sim \exp\left(-\delta t^{\gamma}\right)$ with $\gamma\approx 0.1$; and (c) change of the width Δx d) at a level $u_{\rm max}(e-1)/e$. The width changes according to the relation $\Delta x\sim t^{\eta}$ with $\eta\approx 0.67$ for zero skewness (top curve) and $\eta\approx 0.9$ with $\theta_{\alpha}=0.21$ and $\theta_{\mu}=-0.1$ (lower curve).

Example 4. This example uses an external force $w(x) = -2\sin^2(2\pi x)$ with an initial condition $u^{(0)}(x) = 0.797885e^{-\frac{x^2}{2}}\sin^2(\pi x)$ where the prefactor is chosen in such a way that the integral on $\mathbb R$ is

Mathematics 2017, 5, 12 15 of 19

normalized to one. Figure 5 collects the information about the fractional exponents used and displays the behavior of the approximation.

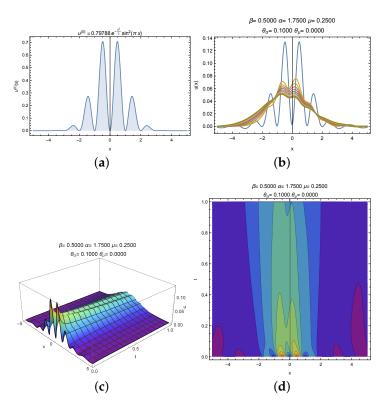


Figure 5. (a) Initial condition to approximate the fFP equation $u^{(0)}(x) = 0.797885e^{-\frac{x^2}{2}}\sin^2(\pi x)$. The external force is set to $w(x) = -2\sin^2(2\pi x)$; (b) the graph shows the approximation for different times $t \in \{0.001, 0.101, 0.201, 0.301, 0.401, 0.501, 0.601, 0.701, 0.801, 0.901\}$ (top down); (c) 3D graph of the approximation; and (d) contour plot of the approximation. The fractional exponent are given on top of the graphs (b), (c), and (d). The number of Sinc points is $N_x = N_t = 64$.

4. Conclusions

Starting from the integer order Fokker-Planck equation and applying the fractional operators to the temporal and spatial coordinates we derived a fully fractional Fokker-Planck equation in temporal and spatial coordinates. The equivalent representation of the Montrol-Weiss equation using the adapted transfer function with three fractional exponents and two skewness parameters provides us with a suitable Volterra integral equation of second type for the numerical approximation. The approximation of the integral equation is using exponentially converging Sinc convolution integrals extended to higher dimensions. The Volterra equation of the second kind is converted in Sinc convolution methods to a linear system of equations which are solved by a Neumann iteration. The convergence of this iteration is absolute and reaches a specified error margin in a reliable fast way. The examples presented demonstrate for a choice of different initial conditions $u^{(0)}(x)$ dispersing approximations. The dispersion is affected by the fractional exponents and the two skewness parameters. Numerical examination of the shift of the maximum and the width of the distribution for different times show that these quantities follow a scaling relation in time. We also observe in these examples so called counter-gradient transport for the function u. Furthermore, a drift of the maximum concentration is observed.

Acknowledgments: One of us, Gerd Baumann, acknowledges financial support by the DAAD/BMBF Project No. 57128284.

Mathematics 2017, 5, 12 16 of 19

Author Contributions: Gerd Baumann and Frank Stenger performed the calculations and computations. Both of us were writing and editing the manuscript.

Conflicts of Interest: The authors declare no conflict of interest.

Appendix A

This appendix collects additional relations and definitions used in the formulation and computation of fractional derivatives.

Let us here discuss a simplified version of the fFP equation reduced to a fractional diffusion equation. The aim here is to introduce the different operators used in the representation of the fFP equation. The space-time fractional diffusion equation in 1+1 dimensions is written as:

$$\mathcal{D}_t^{\beta} u(x,t) = \mathcal{D}_{x:\theta}^{\alpha} u(x,t) \text{ with } -\infty < x < \infty \text{ and } t \ge 0$$
(A1)

where α , β , and θ are real parameters restricted to $0 < \alpha \le 2$, $|\theta| \le \min\{\alpha, 2 - \alpha\}$, and $0 < \beta \le 1$. The fractional derivatives $\mathcal{D}_{x;\theta}^{\alpha}$ is the *Riesz-Feller* fractional derivative of order α and skewness θ , and \mathcal{D}_{t}^{β} is the *Caputo* fractional derivative of order β .

The time fractional derivative on the LHS of (1) can then be removed by a suitable fractional integration leading to:

$$u(x,t) = u(x,0) + \frac{1}{\Gamma(\beta)} \int_0^t \mathcal{D}_{x;\theta}^{\alpha} u(x,\tau) (t-\tau)^{\beta-1} d\tau$$
 (A2)

here the initial condition is used as $u(x,0) = f(x) t^{-\beta} / \Gamma(\beta)$ reflecting the fact of fractional differentiating a constant. The Riesz–Feller fractional derivative is defined as follows:

$$\mathcal{D}_{x:\theta}^{\alpha}f(x) = -\mathcal{I}_{x:\theta}^{-\alpha}f(x) = c_{-}(\alpha,\theta)\mathcal{I}_{+:x:\theta}^{-\alpha}f(x) + c_{+}(\alpha,\theta)\mathcal{I}_{-:x:\theta}^{-\alpha}f(x)$$
(A3)

where the following relations hold:

$$c_{+}(\alpha,\theta) = \frac{\sin((\alpha-\theta)\pi/2)}{\sin(\alpha\pi)} \text{ and } c_{-}(\alpha,\theta) = \frac{\sin((\alpha+\theta)\pi/2)}{\sin(\alpha\pi)}$$
(A4)

and the corresponding fractional integral operators $\mathcal{I}_{x;\theta}^{\alpha}$ (Weyl integrals see e.g., [31]) are given as:

$$\mathcal{I}^{\alpha}_{+;x;\theta}f(x) = \frac{1}{\Gamma(\alpha)} \int_{-\infty}^{x} (x - \xi)^{\alpha - 1} f(\xi) d\xi \tag{A5}$$

and:

$$\mathcal{I}^{\alpha}_{-;x;\theta}f(x) = \frac{1}{\Gamma(\alpha)} \int_{x}^{\infty} (\xi - x)^{\alpha - 1} f(\xi) d\xi \tag{A6}$$

Note the negative sign in the *Riesz-Feller* operator means that we are dealing with fractional derivatives which are defined in terms of integral operators as follows:

$$\mathcal{I}_{+;x;\theta}^{-\alpha}f(x) = \frac{1}{\Gamma(n-\alpha)} \int_{-\infty}^{x} (x-\xi)^{n-\alpha-1} f(\xi) d\xi \tag{A7}$$

and:

$$\mathcal{I}_{-;x;\theta}^{-\alpha}f(x) = \frac{1}{\Gamma(n-\alpha)} \int_{x}^{\infty} (\xi - x)^{n-\alpha-1} f(\xi) d\xi \tag{A8}$$

with $n = [Re(\alpha)] + 1$ and $Re(\alpha) > 0$; here $[\alpha]$ represents the integer part of α .

In terms of these integrals the solution is then:

Mathematics 2017, 5, 12 17 of 19

$$u(x,t) = u(x,0) + \frac{1}{\Gamma(\beta)} \int_{0}^{t} \mathcal{D}_{x;\theta}^{\alpha} u(x,\tau) (t-\tau)^{\beta-1} d\tau$$

$$= u(x,0) + \frac{1}{\Gamma(\beta)} \int_{0}^{t} \left(c_{-}(\alpha,\theta) \mathcal{I}_{+;x;\theta}^{-\alpha} u(x,\tau) + c_{+}(\alpha,\theta) \mathcal{I}_{-;x;\theta}^{-\alpha} u(x,\tau) \right) (t-\tau)^{\beta-1} d\tau$$

$$\Leftrightarrow u(x,t) - \frac{1}{\Gamma(-\alpha)} \frac{1}{\Gamma(\beta)} \left(\frac{\sin((\alpha+\theta)\pi/2)}{\sin(\alpha\pi)} \int_{0}^{t} \int_{-\infty}^{x} (x-\xi)^{-\alpha-1} (t-\tau)^{\beta-1} u(\xi,\tau) d\xi d\tau + \frac{\sin((\alpha-\theta)\pi/2)}{\sin(\alpha\pi)} \int_{0}^{t} \int_{x}^{\infty} (\xi-x)^{-\alpha-1} (t-\tau)^{\beta-1} u(\xi,\tau) d\xi d\tau \right) = u^{0}(x)$$
(A9)

The above equation is a linear Volterra integral equation in two variables t and x for $0 < \alpha \le 2$, $|\theta| \le \min\{\alpha, 2 - \alpha\}$, and $0 < \beta \le 1$.

Here ${}^{C}\mathcal{D}_{t}^{\beta}$ denotes the *Caputo* fractional derivative of order β , acting on a sufficiently well-behaved function f(t) of the variable t:

$${}^{C}\mathcal{D}_{t}^{\beta}f(t) = \frac{1}{\Gamma(1-\beta)} \int_{0}^{t} \frac{f^{(1)}(\tau)}{(t-\tau)^{\beta}} d\tau, 0 < \beta < 1$$
(A10)

The generalized higher order derivative uses the following definition:

$${}^{C}\mathcal{D}_{t}^{\beta}f(t) = \begin{cases} \frac{1}{\Gamma(m-\beta)} \int_{0}^{t} \frac{f^{(m)}(\tau)}{(t-\tau)^{m-\beta-1}} d\tau, & m-1 < \beta < m, m \in \mathbb{N} \\ \frac{d^{m}f(t)}{dt^{m}} & \beta = m, m \in \mathbb{N} \end{cases}$$
(A11)

where we have the two properties for integration and differentiation as follows:

$$\mathcal{D}_{t}^{-\beta}\mathcal{D}_{t}^{\beta}f(t) = f(t) - \sum_{k=0}^{n-1} \frac{t^{k}}{k!} f^{(k)}(0), \ t > 0$$
(A12)

Caputo fractional derivative:

$${}^{C}\mathcal{D}_{t}^{\beta}f(t) = \mathcal{D}_{t}^{\beta}f(t) - \sum_{k=0}^{n-1} \frac{t^{k-\beta}}{\Gamma(k-\beta+1)} f^{(k)}(0), \ t > 0$$
(A13)

In particular for $0 < \beta < 1$ (i.e., n = 1) we have:

$${}^{C}\mathcal{D}_{t}^{\beta}f(t) = \mathcal{D}_{t}^{\beta}f(t) - f(0)\frac{t^{-\beta}}{\Gamma(1-\beta)} = \mathcal{D}_{t}^{\beta}[f(t) - f(0)], \ t > 0$$
(A14)

From Equation (A14) we recognize that the *Caputo* fractional derivative represents a sort of regularization in the time origin for the Riemann-Liouville fractional derivative.

References

- 1. Risken, H. *The Fokker-Planck Equation. Methods of Solution and Applications*; Springer: Berlin, Germany; New York, NY, USA, 1989.
- 2. Gorenflo, R.; Mainardi, F.; Moretti, D.; Pagnini, G.; Paradisi, P. Discrete random walk models for space-time fractional diffusion. *Strange Kinetics* **2002**, *284*, 521–541.
- 3. Scales, E.; Gorenflo, R.; Mainardi, F.; Raberto, M. Revising the Derivation of the Fractional Diffusion Equation. In *Scaling and Disordered Systems*; Family, F., Daoud, M., Hermann, H., Stanley, H.E., Eds.; World Scientific Publishing Co., Pte. Ltd.: Singapore, 2002; pp. 281–289.
- 4. Scalas, E.; Gorenflo, R.; Mainardi, F. Fractional calculus and continuous-time finance. *Physica A* **2000**, *284*, 376–384.

Mathematics 2017, 5, 12 18 of 19

5. Mainardi, F.; Rabertob, M.; Gorenflo, R.; Scalas, E. Fractional calculus and continuous—Time finance II: The waiting-time distribution. *Physica A* **2000**, *287*, 468–481.

- 6. Stenger, F.; Tucker, D.; Baumann, G. *Navier-Stokes Equation on* $\mathbb{R}^3 \times [0,T]$; Springer International Publishing: Cham, Switzerland, 2016.
- 7. Stenger, F.; Baumann, G.; Koures, V.G. Computational Methods for Chemistry and Physics, and Schrödinger in 3 + 1. In *Concepts of Mathematical Physics in Chemistry: A Tribute to Frank E. Harris—Part A*; Elsevier: Amsterdam, The Netherlands, 2015; Volume 71, pp. 265–298.
- 8. Dolgov, S.V.; Khoromskij, B.N.; Oseledets, I. Fast solution of multi-dimensional parabolic problems in the TT/QTT formats with initial application to the Fokker-Planck equation. *SIAM J. Sci. Comput.* **2012**, *34*, A3016–A3038.
- 9. Dolgov, S.; Khoromskij, B.N. Simultaneous state-time approximation of the chemical master equation using tensor product formats. *Numer. Linear Algebra Appl.* **2015**, 22, 197–219.
- 10. Chen, W.; Wang, F. Singular boundary method using time-dependent fundamental solution for transient diffusion problems. *Eng. Anal. Bound. Elem.* **2016**, *68*, 115–123.
- 11. Wang, F.; Chen, W. Accurate empirical formulas for the evaluation of origin intensity factor in singular boundary method using time-dependent diffusion fundamental solution. *Int. J. Heat Mass Transf.* **2016**, 103, 360–369.
- 12. Jeon, J.-H.; Metzler, R. Fractional Brownian motion and motion governed by the fractional Langevin equation in confined geometries. *Phys. Rev. E* **2010**, *81*, 021103.
- 13. Metzler, R.; Klafter, J. The Random Walks Guide to Anomalous Diffusion: A Fractional Dynamics Approach. *Phys. Rep.* **2000**, 339, 1–77.
- 14. Chechkin, A. V.; Zaid, I. M.; Lomholt, M.A.; Sokolov, I.M.; Metzler, R. Bulk-mediated surface diffusion along a cylinder: Propagators and crossovers. *Phys. Rev. E* **2009**, *79*, 040105.
- 15. Gorenflo, R.; Mainardi, F. Some recent advances in theory and simulation of fractional diffusion processes. *J. Comput. Appl. Math.* **2009**, 229, 400–415.
- 16. Baumann, G.; Stenger, F. Sinc-Approximations of Fractional Operators: A Computing Approach. *Mathematics* **2015**, *3*, 444–480.
- 17. Baumann, G.; Stenger, F. Fractional calculus and Sinc methods. Fract. Calc. Appl. Anal. 2011, 14, 568–622.
- 18. Stenger, F. Handbook of Sinc Numerical Methods; CRC Press: Boca Raton, FL, USA, 2011.
- 19. Stenger, F. Summary of Sinc numerical methods. J. Comp. Appl. Math. 2000, 121, 379–420.
- 20. Stenger, F. Collocating convolutions. *Math. Comp.* **1995**, *64*, 211–235.
- 21. Stenger, F. Numerical Methods Based on Sinc and Analytic Functions; Springer: New York, NY, USA, 1993.
- 22. Kowalski, M.A.; Sikorski, K.A.; Stenger, F. *Selected Topics in Approximation and Computation*; Oxford University Press: New York, NY, USA, 1995.
- 23. Stenger, F.; El-Sharkawy, H.A.M.; Baumann, G. The Lebesgue Constant for Sinc Approximations. In *New Perspectives on Approximation and Sampling Theory—Festschrift in the Honor of Paul Butzer's 85th Birthday*; Zayed, A., Schmeisser, G., Eds.; Birkhaeuser: Basel, Switzerland, 2014; pp. 319–335.
- 24. Youssef, M.; El-Sharkawy, H.A.; Baumann, G. Lebesgue Constant Using Sinc Points. *Adv. Numer. Anal.* **2016**, doi:10.1155/2016/6758283.
- 25. Youssef, M.; El-Sharkawy, H.A.; Baumann, G. Multivariate Lagrange Interpolation at Sinc Points Error Estimation and Lebesgue Constant. *J. Math. Res.* **2016**, *8*, 118.
- 26. Youssef, M.R.; Baumann, G. Collocation Method to Solve Elliptic Equations, Bivariate Poly-Sinc Approximation. *J. Progress. Res. Math.* **2016**, *7*, 1079–1091.
- 27. Lubich, C. Convolution Quadrature and Discretized Operational Calculus. I. Numer. Math. 1988, 52, 129–145.
- 28. Lubich, C. Convolution Quadrature and Discretized Operational Calculus. II. *Numer. Math.* **1988**, 52, 413–425.
- 29. Han, L.; Xu, J. Proof of Stenger's conjecture on matrix $I^{(-1)}$ of Sinc methods. *J. Comput. Appl. Math.* **2014**, 255, 805–811.

Mathematics 2017, 5, 12 19 of 19

30. Paradisi, P.; Cesari, R.; Mainardi, F.; Maurizi, A.; Tampieri, F. A generalized Fick's law to describe non-local transport effects. *Phys. Chem. Earth Part B Hydrol. Oceans Atmos.* **2001**, *26*, 275–279.

31. Kilbas, A.A.; Srivastava, H.M.; Trujillo, J.J. *Theory and Applications of Fractional Differential Equations*; Elsevier: Amsterdam, The Netherlands, 2006.



© 2017 by the authors; licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (http://creativecommons.org/licenses/by/4.0/).