A Theory of Everything: When Information Geometry Meets the Generalized Brownian Motion and the Einsteinian Relativity

Ismail A Mageed*

Member IAENG, IEEE, School of Computing, AI, and Electronics, University of Bradford, United Kingdom

*Corresponding Author
Ismail A Mageed, Member IAENG, IEEE, School of Computing, AI, and Electronics, University of Bradford, United Kingdom.

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Abstract
Information geometry (IG) is a fascinating combination of differential geometry and statistics where a Riemannian manifold's structure is applied to a statistical model. It may find numerous fascinating uses in the domains of theoretical neurology, machine learning, complexity, and (quantum) information theory, among others. (IG) aims to provide a differential-geometric perspective on statistical geodesic models' structure. In this case, IG, KD, and J-divergence (JD) are used to define the manifold of the Generalised Brownian Motion (GBM). Consequently, the geodesic equations (GEs) are devised, and GB information matrix exponential (IME) is presented. Moreover, for first time ever, the necessary and sufficient mathematical requirement that characterizes the developability of Generalized Brownian Motion (GBM) manifold is devised. Also, a novel sufficient and necessary conditions which characterizes the regions where the surface describing GBM is minimal is determined. Also, it is shown that GBM has a non-zero 0-Gaussian and Ricci curvatures. Consequently, this advances the establishment of a unified GBM- Relativistic Info-Geometric based analysis.

Keywords: Generalized Brownian Motion (GBM), Probability Density Function (PDF), Matrix Exponential of A (e^A), (Geodesic Equations (GEs)

1. Introduction

1.1 Information Geometry
Stochastic control, neural networks, statistical inference, and other study domains have all made extensive use of information geometry (IG) [1]. To put it another way, IG seeks to integrate statistics with differential geometry (DG) methods. Applying non-Euclidean geometry approaches and techniques to probability theories and stochastic processes is, thus, the central premise of IG. A topological finite dimensional Cartesian space, R^n, where one possesses an infinite-dimensional manifold, is called a manifold [2-4]. The challenge of making decisions can be applied to the parameter inference of a model based on data in figure 1.
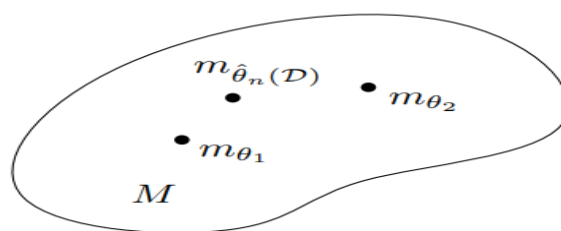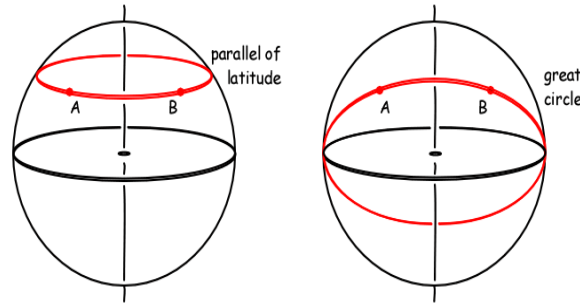

Figure 1: SM's Prametrization [1]

Matrix exponential plays a significant role in Lie groups theory [5,6]. From a visual perspective, figure 2 portrays the geometric representation of geodesics on curved surfaces.

Figure 2: Geodesics on Curved Surfaces Geometric Representation (c.f., [7])

Einstein's theory of classical general relativity (GR) holds that objects travel along a geodesic in a curved space-time that extremizes the proper time between two locations. A geodesic, which reduces the distance between two places, is also known as a "straight line on a curved surface."

1.2 Generalized Brownian Motion (GBM)

Simple physical systems' internal diffusion was firstly defined by Einstein [8-39].
The Tsallisian entropic measure in discrete domain reads:

$$H_q = \frac{1}{(q-1)}\left(1 - \sum_{n=1}^{\infty}(p(n))^q\right) \tag{1.1}$$

where $p(n)$ is the probability of being in the nth state. In the limit $q\to1$, the Tsallis entropy reduces to the Boltzmann-Gibbs entropy.

The Tsallisian entropic measure in continuous domain reads: [40-42]

$$H_q(X) = \frac{1}{q-1}\left(1 - \int_{-\infty}^{\infty}[f_X(x)]^q \, dx\right) \tag{1.2}$$

where $f_X(x)$ is PDF(X)
PDF$_{q-Gaussian}$ reads

$$f(x) = \frac{\sqrt{\beta}}{C_q}e_q(-\beta x^2) \tag{1.3}$$

with

$$e_q(x) = [1 + (1-q)x]^{1/(1-q)} \tag{1.4}$$

is called the q-exponential, C_q is a normalization constant, and $\beta > 0$ is a scale parameter, and $q, \ 1 < q < 3$
finetunes extensivity, for a distributional similarity scaling $\beta = \frac{v+1}{2v} = \frac{1}{3-q}$.

1.3 Random Diffusivity [41]

Looking at

$$dX(t) = vdt + \sqrt{D}dB(t) \tag{1.5}$$

where $B(t)$ is a Brownian motion, and D is a random variable that is independent of $B(t)$. If the probability density function, $f_D(x)$, of D is given by:

$$f_D(x) = \delta(x - D_0) \tag{1.6}$$

D is a constant. Suppose that [43]:

$$D \sim D_0\left(\frac{v}{V}\right)^2 \equiv g(V) \tag{1.7}$$

$$X(t) - X(0) - vt \sim \sqrt{Dt}Z \tag{1.8}$$

$$\sim \sqrt{D_0 t}\frac{Z}{V/v} \tag{1.9}$$

Most importantly, (1.7) can be re-written as:

$$f_D(x) = f_V(g^{-1}(x)) \left|\frac{d}{dx}g^{-1}(x)\right|$$

$$= \frac{D_0^{\left(\frac{v}{4}\right)}v^{\left(\frac{v}{2}\right)}x^{-\left(\frac{v}{4}+1\right)}}{2^{\left(\frac{v}{2}+1\right)}\Gamma\left(\frac{v}{2}\right)}e^{\left(-\frac{v\sqrt{D_0}}{2\sqrt{x}}\right)} \tag{1.10}$$
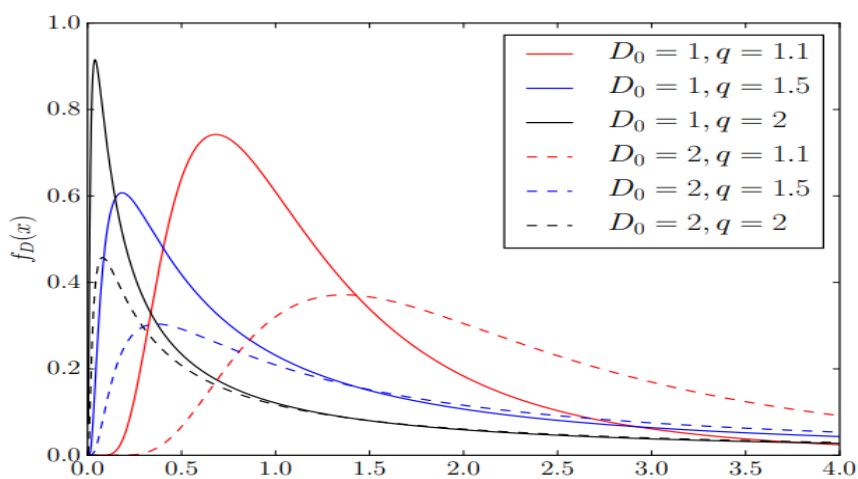

Figure 3: Several plots of $f_D(x)$ for several combinations of q, $v= \frac{3-q}{q-1}$ and D_0(c.f. [41])

An idealised approximation to real random dynamics, known as Brownian motion, has been the subject of intense research over an extended temporal period [44–56]. A numerical simulation of a particle's route is shown in figure 4.
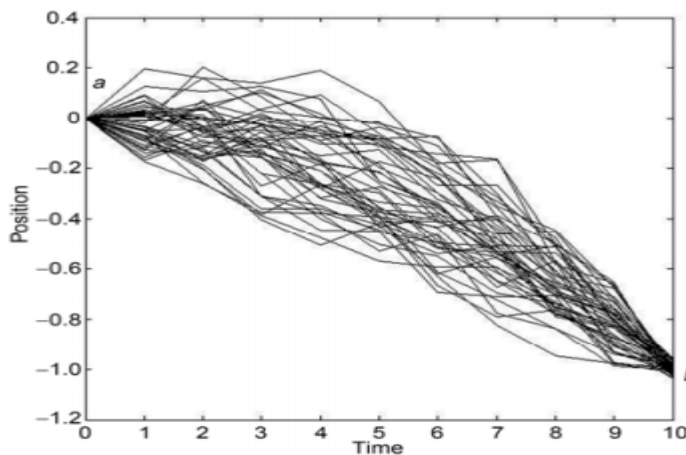

Figure 4: (c.f., [57])

1.4 Polygamma Functions

The classical Euler gamma function may be defined for by $\Re(z) > 0$ by

$$\Gamma(z) = \int_0^\infty y^{z-1} e^{-y} dy \tag{1.11}$$

The digamma function, $\psi(z)$ reads as:

$$\psi(z) = \frac{d}{dz}\left(ln\Gamma(z)\right) = \frac{\Gamma'(z)}{\Gamma(z)}, \ \psi(1) = -e \tag{1.12}$$

The derivatives $\psi'(z)$ and $\psi''(z)$ ($\psi_1(z)$ and $\psi_2(z)$ respectively) are called the trigamma and tetragamma functions respectively.

The Trigamma function, $\psi_1(z)$ is determined by [58]

$$\psi_1(z) = \sum_{n=0}^{\infty} \frac{1}{(n+z)^2} \tag{1.13}$$

where $\psi_1(z)$ satisfies $\psi_1(1) = \dfrac{\pi^2}{6}$ $(c.f., [58])$

Clearly, $\psi_2(z)$ solves (1.13) [59]

$$\psi_2(z) = -2\sum_{n=0}^{\infty} \frac{1}{(n+z)^3} \tag{1.14}$$

This paper contributes to:
- The above discussion clearly demonstrates the significance of our novel approach as using the powerful tool of IG will open new grounds by looking for the first time (GB) as a manifold, by which we will directly determine visualization of (GBM) and searching in more depth for the temporal paths of the parameters of (GBM). Furthermore, IG technique is used to analyse (GBM) as well as the derivation of the corresponding curvature tensors, which physically describe the geometric shape of (GBM). More importantly, this clarifies the fact that Ig provides such an analysis of (GBM) which is impossible to be done by any other mathematical approach.
- The new discovery of the informational geometric equations of motion (IGEMs) for (GBM) as well as obtaining the solutions of these equations.
- Further advancements in the current paper are made by providing the information theoretic impact on the devised path of motion characterizing IGEMs.
- Included are the physical applications and explanations of Gaussian and Ricci curvatures.
- Providing a cutting-edge method for using computational information geometry to visualise queueing systems that has been suggested.
- The creation of a novel quantitative approach to ascertain GBM's temporal dynamics for the first time ever.

The rest of this paper is organised as follows: Section 2 presents preliminary definitions associated with (IG). The FIM and its inverse as well as the Fisher information metric for the Generalized Brownian Motion Manifold (GBM) are introduced in Section 3. The α(or $\nabla^{(\alpha)}$)-*connection* of the Generalized Brownian Motion Manifold (GBM) is obtained in Section 4. The GEs, KD and JD (c.f., [60]) of the Generalized Brownian Motion Manifold (GBM) are obtained in Section 5, numerical solutions to the devised GEs are obtained by using Rung-Kutta method of order 4 and numerical experiments are determined to illustrate the physical interpretations of our solutions. The structured proofs of calculating the 0-Gaussian curvature as well as the investigation of the developability of Generalized Brownian Motion Manifold (GBM) and showing that GBM has a non-zero Ricci curvature (RC) tensor are devised in Section 6. The exponential matrix analysis of the Generalized Brownian Motion Manifold (GBM) is obtained in Section 7. Conclusions and future research directions are included in Section 8.

2. Main Definitions in IG
2.1 Preliminary Definitions

Definition 2.1 Statistical Manifold (SM)
An SM is referred to as $M = \{p(x,\theta) \,|\, \theta \in \Theta\}$ [60], with a probability density function $p(x,\theta)$, and $\theta = (\theta_1, \theta_2, .., \theta_n) \in \Theta$ as the manifold M's coordinates.

Definition 2.2 Potential Function
The potential function $\Phi(\theta)$ (c.f., [60]) is the coordinates only part of $(-ln\,((p(x,\theta)))$ (c.f., (2.1)).

Definition 2.3 FIM, namely $[g_{ij}]$
$[g_{ij}]$(c.f., [61]) reads as:

$$[g_{ij}] = \left[\frac{\partial^2}{\partial\theta_i \,\partial\theta_j}(\Phi(\theta))\right], i, j = 1, 2, \ldots, n \tag{2.1}$$

Definition 2.4 FIM's inverse

$[g_{ij}]$ inverse (c.f.,[60]) reads as:

$$[g^{ij}] = ([g_{ij}])^{-1} = \frac{adj[g_{ij}]}{\Delta}, \Delta = \det[g_{ij}] \tag{2.2}$$

The arc length function reads:

$$(ds)^2 = \sum_{i,j=1}^{n} g_{ij}(d\theta^i)(d\theta^j) \tag{2.3}$$

Definition 2.5 α-Connection, $\nabla^{(\alpha)}$, $\alpha \in R$ [61]

$\nabla^{(\alpha)}$ reads

$$\Gamma_{ij,k}^{(\alpha)} = (\frac{1-\alpha}{2})(\partial_i\partial_j\partial_k(\Phi(\theta))) \tag{2.4}$$

Definition 2.6 Kullback's Divergence (KD), $K(p,q)$

$$K(p,r) = \int p(x;\theta_p)ln\left(\frac{p(x;\theta_p)}{r(x;\theta_r)}\right)dx \tag{2.5}$$

The J-divergence reads as:

$$J(p,r) = \int (p(x;\theta_p) - r(x;\theta_r))\, ln\left(\frac{p(x;\theta_p)}{r(x;\theta_r)}\right)dx \tag{2.6}$$

Definitions 2.7 Geodesic Equations (GEs)

1. The GEs of manifold M with coordinate system $\theta=(\theta_1,\theta_2,..,\theta_n)$ c.f., [60]) are defined by

$$\frac{d^2\theta^k}{dt^2} + \Gamma_{ij}^{k(0)}\left(\frac{d\theta^i}{dt}\right)(\frac{d\theta^j}{dt}) =, i, j = 1, 2, \ldots, n, \Gamma_{ij}^{k(\alpha)} = \Gamma_{ij,s}^{(\alpha)}g^{sk}, i, j, k, s = 1, 2, \ldots, n \tag{2.7}$$

2.[60] A path $\theta = \theta(t)$ minimising the total energy $E=E(\theta)$ must satisfy:

$$\frac{d}{dt}\left(\frac{\partial^2 L}{\partial(\frac{d\theta_j}{dt})}\right) - \frac{\partial}{\partial\theta_j} = 0 \tag{2.8}$$

FIM's inverse defines g_{ij} (IM) so that:

$$\sum_{k=1}^{n} g^{ik}\, g_{kj} = \delta_{ij} \tag{2.9}$$

where $\delta_{ij} = \begin{cases} 1, & i = j \\ 0, & otherwise \end{cases}$ (c.f., [60])

5. Lemma [60] GEs are the equations of motion

$$L = L\left(\theta, \frac{d\theta}{dt}\right)$$

Such that

$$\frac{d^2\theta_i}{dt^2} + \sum_{j,k=1}^{n}\Gamma_{j,k}^i\left(\frac{d\theta_j}{dt}\right)\left(\frac{d\theta_k}{dt}\right) = 0 \tag{2.10}$$

Hence, GEs are interpreted physically as the information geometric equations of motion (IGEMs).

Definition 2.8

1. The $\alpha - $ curvature RiemannianTensors, $R_{ijkl}^{(\alpha)}$(c.f., [58]) are defined by

$$R_{ijkl}^{(\alpha)} = [(\partial_j \Gamma_{ik}^{s(\alpha)} - \partial_i \Gamma_{jk}^{s(\alpha)})g_{sl} + (\Gamma_{jt,l}^{(\alpha)} \Gamma_{ik}^{t(\alpha)} - \Gamma_{it,l}^{(\alpha)} \Gamma_{jk}^{t(\alpha)})] \tag{2.11}$$

where $\Gamma_{ij}^{k(\alpha)} = \Gamma_{ij,s}^{(\alpha)} g^{sk}$

2. $R_{ik}^{(\alpha)}$ define the Ricci tensors (c.f.,[58]),

$$R_{ik}^{(\alpha)} = R_{ijkl}^{(\alpha)} g^{jl} \tag{2.12}$$

3. $K_{ijij}^{(\alpha)}$ reads as (c.f.,[60])

$$K_{ijij}^{(\alpha)} = \frac{R_{ijij}^{(\alpha)}}{(g_{ii})(g_{jj}) - (g_{ij})^2} \tag{2.13}$$

$K_{1212}^{(\alpha)} = K^{(\alpha)}$ defines the α Gaussian curvature (c.f., [60])

$$K^{(\alpha)} = \frac{R_{1212}^{(\alpha)}}{det(g_{ij})} \tag{2.14}$$

4. The Ricci Tensor (c.f.,[60]) defines the contraction of $R_{ijkl}^{(\alpha)}$ (c.f.,[60]).

5. The difference between a geodesic ball's volume on the surface and its volume in Euclidean space is known as the oriented Riemannian Manifold M ' s Ricci curvature Tensor (c.f., [62]).

6. The volumes' evolution is contracted by the Ricci curvature RC) (c.f., [63]). The Bonnet Myers theorem (c.f., [64]) states that when Ricci curvature is positive, the Riemannian manifold has a smaller diameter and is more positively curved than a sphere.
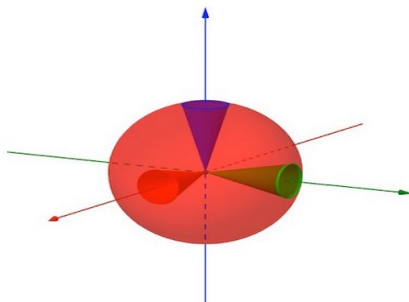

Figure 5: Visualizing Ricci Curvature (c.f., [65])

Definition 2.9[66]
1. Given a n-dimensional vector x is and A is a nxn matrix, such that

$$\frac{dx}{dt} = Ax \tag{2.15}$$

e^A reads:

$$e^A = \sum_{i=0}^{\infty} \frac{A^i}{i!} = I + A + \frac{A^2}{2!} + \cdots + \frac{A^k}{k!} + \dots \tag{2.16}$$

solves (2.15).

2. Let $\tag{2.17}$

$$\Phi(\delta) = det(A - \delta I)$$

$$\Phi(\delta) = 0 \tag{2.18}$$

The solution of (2.18) are referred to as A's eigen values such that:

$$Ax = \delta x \tag{2.19}$$

Notably, e^A may be re-written as:

$$e^A = T\ e^D\ T^{-1} \tag{2.20}$$

where D is the diagonal matrix of eigen values of A, and T is matrix having of the corresponding eigen vectors of A as its columns.

Definition 2.10

A unique type of ruled surface known as a developable surface has a Gaussian curvature of 0 and can be mapped onto a plane surface without causing any curve distortion; a curve drawn from such a surface onto a flat plane stays the same (c.f., [67]).
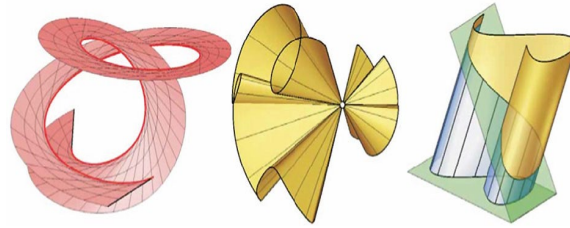
Figure 6: Three Kinds of Developable Surfaces: Tangential on the left, Conical on the centre and on the right, Cylindrical (c.f., [67])

In the following, a brief overview of Gaussian and Mean Curvatures (c.f.,[68])

2.2 Gaussian and Mean Curvatures [68]

Definition 2.11 (Mong Patch Technique)

1. Let K_1 and K_2 be the principal curvatures of a surface patch $\delta(u,v)$. The *Gaussian Curvature K_G* of δ is

$$K_G = K_1 K_2 \qquad (2.21)$$

and its *Mean Curvature is*

$$H = \frac{1}{2}(K_1 + K_2) \qquad (2.22)$$

2. For a Mong patch $z=f(x,y), K_G$ and H are given by

$$K_G = \frac{LN - M^2}{EG - F^2} \qquad (2.23)$$

and its *Mean Curvature* reads

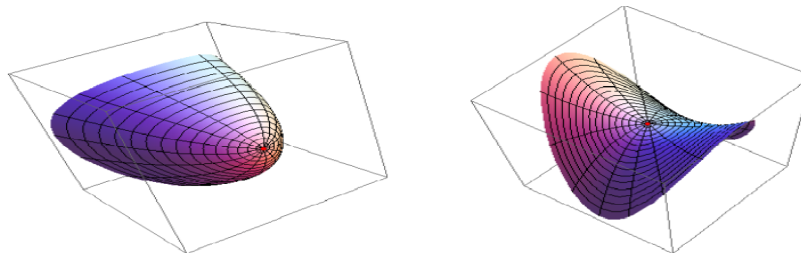$$H = \frac{1}{2}\left(\frac{LG - 2MF + NE}{EG - F^2}\right) \qquad (2.24)$$

with $E = (\frac{\partial f}{\partial x})^2, F = \frac{\partial f}{\partial x \partial y}, G = (\frac{\partial f}{\partial y})^2, L = \frac{\partial^2 f}{\partial x^2}, M = \frac{\partial f}{\partial xy}, N = \frac{\partial^2 f}{\partial y^2}$

2.2.1 Classification of Surface Points

The physical interpretation of the sign of K_G at a point r on a surface S is:
1. $\boldsymbol{K_G} > 0$ The principal curvatures K_1 and K_2 have the same sign at r. An illustrative example is given by figure 7.

Figure 7: The Elliptic Paraboloids $z = x^2 + 2y^2$ (*to the left*) *and* $z = x^2 - 2y^2$ (*to the right*) (c.f., [68])

2. $\boldsymbol{K_G} > 0$ The principal curvatures K_1 and K_2 have opposite signs at r.
3. $\boldsymbol{K_G} > 0$, point in the plane and the origin of a monkey saddle $z = x^3 - 3xy^2$ as in figure 8, are both planar points, but they have quite different shapes.
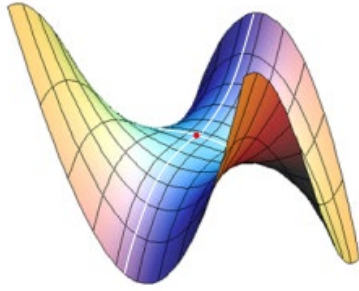
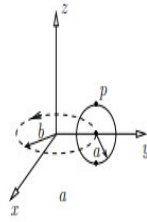**Figure 8: (c.f., [68])**   **Figure 9: (c.f., [68])**

A torus as shown in figure 9. Minimal surfaces have Gaussian curvature $K_G \leq 0$. The catenoid in figure 10 is a minimal surface.
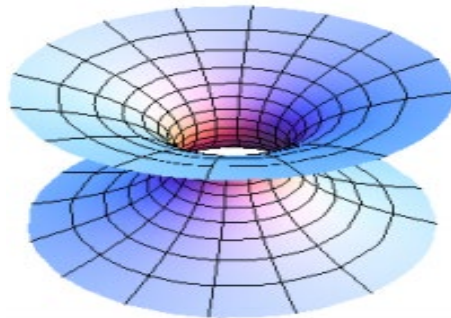

Figure 10: (c.f., [68])

2.3 Different Approach to Gaussian and Mean Curvatures (Angular Technique) [69]

A new formulation (c.f., [69]) is introduced for *Gaussian Curvature* K_G and the *Mean Curvature* is H are defined as

$$K_G = K_1 K_2 \tag{2.25}$$

and

$$H = \frac{1}{2}(K_1 + K_2) \tag{2.26}$$

with K_1 and K_2 as the principal curvatures are determined by:

$$K_1 = \frac{B_{11}}{(1+(\frac{\partial f}{\partial x_1'})^2)^{\frac{3}{2}}}, K_2 = \frac{B_{22}}{(1+(\frac{\partial f}{\partial x_2'})^2)^{\frac{3}{2}}} \tag{2.27}$$

where $x_3 = f(x_1, x_2)$ defines the shape of the surface. x_1' and x_2' are parallel to the directions of the principal curvature, which are rotated through an angle Ͻ with respect to x_1 and x_2, and

$$\frac{\partial f}{\partial x_1'} = cos Ͻ \frac{\partial f}{\partial x_1} - sin Ͻ \frac{\partial f}{\partial x_2} \tag{2.28}$$

$$\frac{\partial f}{\partial x_2'} = sin Ͻ \frac{\partial f}{\partial x_1} + cos Ͻ \frac{\partial f}{\partial x_2} \tag{2.29}$$

$$B_{11} = \frac{\partial^2 f}{\partial x_1^2} cos^2 Ͻ - 2 \frac{\partial^2 f}{\partial x_1 \partial x_2} sin Ͻ cos Ͻ + \frac{\partial^2 f}{\partial x_2^2} sin^2 Ͻ = \frac{\partial^2 f}{\partial x_1'^2} \tag{2.30}$$

$$B_{22} = \frac{\partial^2 f}{\partial x_1^2} sin^2 Ͻ + 2 \frac{\partial^2 f}{\partial x_1 \partial x_2} sin Ͻ cos Ͻ + \frac{\partial^2 f}{\partial x_2^2} cos^2 Ͻ = \frac{\partial^2 f}{\partial x_2'^2} \tag{2.31}$$

$$tan2\mathcal{O} = \frac{-2\left(\frac{\partial^2 f}{\partial x_1 \partial x_2}\right)}{\left(\frac{\partial^2 f}{\partial x_1^2} - \frac{\partial^2 f}{\partial x_2^2}\right)} \tag{2.32}$$

Where structures appear on a surface is indicated by a Gaussian curvature contour map. A surface is considered to have twofold curvature if both of its primary curvatures are non-zero [70].
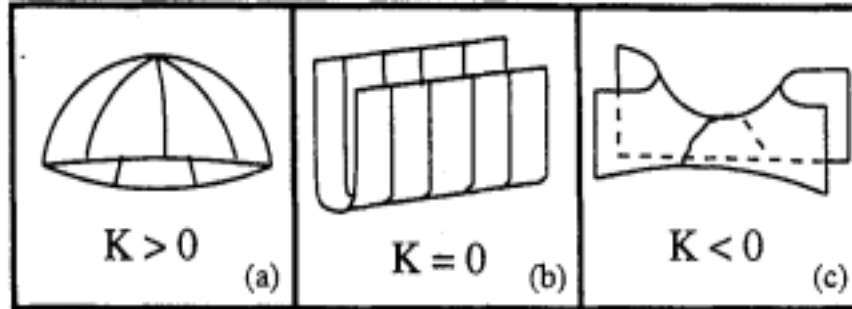

Figure 11: (c.f., [69]). (a) $K_G > 0$, **(b)** $K_G = 0$ **and (c)** $K_G < 0$

This paper contributes to:
i) The FIM and its inverse as well as the FIM for (GBM) are introduced.
ii) A novel GBM's $\nabla^{(\alpha)}$ [61] are devised. These novel derivations are used to determine the GEs for the coordinates of (GBM). These equations are considered as analogues to the equations of motion in dynamics. In other words, the information geometric equations of motion are devised for the coordinates of the (GBM) manifold. iii) The GEs, the KD, and the JD of (GBM) are determined. iv) (GBM) could be compressible or non-solenoidal (for a more detailed survey, consult [71]). Moreover, for first time ever, the necessary and sufficient mathematical requirement that characterizes the developability of GBM manifold is devised. Also, a novel sufficient and necessary conditions which characterizes the regions where the surface describing GBM is minimal is determined. v) The exponential of the FIM for GBM solves $\frac{dx}{dt} = Ax$.

3. The Fim and its Inverse For (GBM)

Theorem 3.1 For GBM manifold, we have
(i)The FIM is given by:

$$[g_{ij}] = \begin{pmatrix} -\frac{1}{v} + \frac{1}{4}\psi_1\left(\frac{v}{2}\right) & -\frac{1}{4D_0} \\ -\frac{1}{4D_0} & \frac{v}{4D_0^2} \end{pmatrix} \tag{3.1}$$

(ii) $(ds)^2 = \left(-\frac{1}{v} + \frac{1}{4}\psi_1\left(\frac{v}{2}\right)\right)(dv)^2 - \frac{1}{4D_0}(dv)(dD_0) + \frac{v}{4D_0^2}(dD_0)^2 \tag{3.2}$

(iii) IFIM reads as:

$$[g^{ij}] = ([g_{ij}])^{-1} = \frac{adj[g_{ij}]}{\Delta}$$

$$= \frac{1}{\Delta}\begin{pmatrix} \frac{v}{4D_0^2} & \frac{1}{4D_0} \\ \frac{1}{4D_0} & -\frac{1}{v} + \frac{1}{4}\psi_1\left(\frac{v}{2}\right) \end{pmatrix} \tag{3.3}$$

with $\Delta = \frac{1}{16D_0^2}\left(v\psi_1\left(\frac{v}{2}\right) - 5\right)$

Proof

(i)We have $f_D(x) = \dfrac{D_0^{\left(\frac{v}{4}\right)}v^{\left(\frac{v}{2}\right)}x^{-\left(\frac{v}{4}+1\right)}}{2^{\left(\frac{v}{2}+1\right)}\Gamma\left(\frac{v}{2}\right)}e^{\left(-\frac{v\sqrt{D_0}}{2\sqrt{x}}\right)}$ (c.f., (1.10))

$$\mathcal{L}(x;\theta) = ln\left(f_D(x;\theta)\right)$$

$$= \left(-l\,n\,\Gamma\left(\frac{v}{2}\right) - \left(1 + \frac{v}{2}\right)ln2 - \left(-\frac{v\sqrt{D_0}}{2\sqrt{x}}\right) + \frac{v}{4}lnD_0 + \frac{v}{2}lnv - \left(1 + \frac{v}{4}\right)lnx\right),$$

$$\theta = (\theta_1, \theta_2) = (v, D_0)$$

$$(3.4)$$

$$\Phi(\theta) = \left(l\,n\,\Gamma\left(\frac{v}{2}\right) + \left(1 + \frac{v}{2}\right)ln2 - \frac{v}{4}lnD_0 - \frac{v}{2}lnv\right), \tag{3.5}$$

Thus,

$$\partial_1 = \frac{\partial\Phi}{\partial v} = \left(\frac{1}{2}ln2 + \frac{1}{2}\psi\left(\frac{v}{2}\right) - \frac{1}{2} - lnv\right) \tag{3.6}$$

$$\partial_2 = \frac{\partial\Phi}{\partial D_0} = -\frac{v}{4D_0} \tag{3.7}$$

$$\partial_1\partial_1 = \frac{\partial^2\Phi}{\partial v^2} = -\frac{1}{v} + \frac{1}{4}\psi_1\left(\frac{v}{2}\right) \tag{3.8}$$

$$\partial_1\partial_2 = \partial_2\partial_1 = -\frac{1}{4D_0} \tag{3.9}$$

$$\partial_2\partial_2 = \frac{v}{4D_0^2} \tag{3.10}$$

Therefore, the FIM is given by

$$[g_{ij}] = \begin{pmatrix} -\frac{1}{v} + \frac{1}{4}\psi_1\left(\frac{v}{2}\right) & -\frac{1}{4D_0} \\ -\frac{1}{4D_0} & \frac{v}{4D_0^2} \end{pmatrix} \qquad (\text{c. f.,}(3.1)), \text{ which proves (i).}$$

(ii) is immediate.

(iii) We must show that:

$$\Delta = \det([g_{ij}]) = \frac{1}{16D_0^2}\left(v\psi_1\left(\frac{v}{2}\right) - 5\right) \neq 0.$$

If $q(1 < q < 3)$, $v = \frac{3-q}{q-1}$, then the extensivity of v follows, and $v > 1$. Assuming that $\Delta = \det([g_{ij}]) = \frac{1}{16D_0^2}\left(v\psi_1\left(\frac{v}{2}\right) - 5\right) = 0$.

Then, without loss of generality, let $n = 1$ in (1.13). Hence, it follows by the assumption that:

$$\left(v\psi_1\left(\frac{v}{2}\right) - 5\right) = \left(v\left(\frac{1}{\left(\frac{v}{2}\right)^2} + \frac{1}{\left(1 + \frac{v}{2}\right)^2}\right) - 5\right) = 0 \tag{3.11}$$

Therefor

$$5v^3 + 12v^2 + 4v - 16 = 0 \tag{3.12}$$

Let's call $5v^3 + 12v^2 + 4v - 16$, the non-vanishing prover (NVP). More importantly, NVP$\neq 0$ (see figure 12), debunking (3.12), that is:

$$\Delta = \det([g_{ij}]) = \frac{1}{16D_0^2}\left(v\psi_1\left(\frac{v}{2}\right) - 5\right) \neq 0.$$
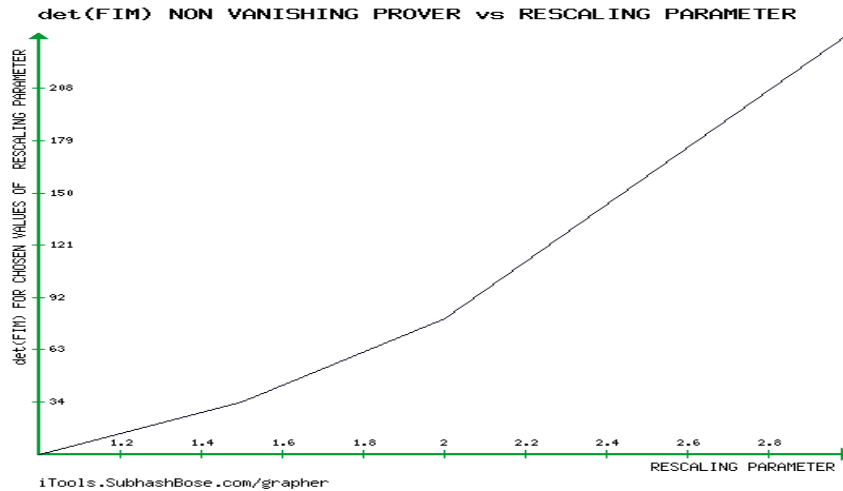
Figure 12: NVP is Non-Zero

Figure 12 shows that det (FIM) is non-zero. For higher level values of $n \geq 2$, the case is always true as it is obvious that the values of $v\psi_1\left(\frac{v}{2}\right)$ will be more increasing. This shows that the inverse of Fisher Information Matrix of GBM exists. The remaining proof of (iii) follows.

It is notable to mention that all the novel findings in this paper involve digamma, Trigamma and Tetragamma functions which indicates another physical importance of our breakthrough as it unifies IG with (GBM) together with many other physical as well as mathematical sciences.

4. The α(OR $\nabla^{(\alpha)}$)-Connection of (GBM)

By definition (2.4), we have

$$\Gamma^{(\alpha)}_{11,1} = \frac{(1-\alpha)}{2}\left(\frac{1}{v^2} + \frac{1}{8}\psi_2\left(\frac{v}{2}\right)\right) \tag{4.1}$$

$$\Gamma^{(\alpha)}_{11,2} = \Gamma^{(\alpha)}_{12,1} = \Gamma^{(\alpha)}_{21,1} = 0 \tag{4.2}$$

$$\Gamma^{(\alpha)}_{12,2} = \frac{(1-\alpha)}{8D_0^2} = \Gamma^{(\alpha)}_{22,1} = \Gamma^{(\alpha)}_{21,2} \tag{4.3}$$

$$\Gamma^{(\alpha)}_{22,2} = \frac{(\alpha-1)v}{4D_0^3} \tag{4.4}$$

Furthermore, the following expressions follow:

$$\Gamma^{1(\alpha)}_{11} = \Gamma^{(\alpha)}_{11,1}g^{11} + \Gamma^{(\alpha)}_{11,2}g^{21},$$

which is after some lengthy calculations:

$$\Gamma^{1(\alpha)}_{11} = \frac{(1-\alpha)v}{8\Delta D_0^2}\left(\frac{1}{v^2} + \frac{1}{8}\psi_2\left(\frac{v}{2}\right)\right), \Gamma^{1(0)}_{11} = \frac{v}{8\Delta D_0^2}\left(\frac{1}{v^2} + \frac{1}{8}\psi_2\left(\frac{v}{2}\right)\right) \tag{4.5}$$

$$\Gamma^{1(\alpha)}_{12} = \Gamma^{1(\alpha)}_{21} = \frac{(1-\alpha)}{32\Delta D_0^3}, \Gamma^{1(0)}_{12} = \Gamma^{1(0)}_{21} = \frac{1}{32\Delta D_0^3} \tag{4.6}$$

$$\Gamma^{1(\alpha)}_{22} = \frac{(\alpha-1)v}{32\Delta D_0^4}, \Gamma^{1(0)}_{22} = -\frac{v}{32\Delta D_0^4} \tag{4.7}$$

$$\Gamma^{2(\alpha)}_{11} = \frac{(1-\alpha)}{8\Delta D_0^2}\left(\frac{1}{v^2} + \frac{1}{8}\psi_2\left(\frac{v}{2}\right)\right), \Gamma^{2(0)}_{11} = \frac{1}{8\Delta D_0^2}\left(\frac{1}{v^2} + \frac{1}{8}\psi_2\left(\frac{v}{2}\right)\right) \tag{4.8}$$

$$\Gamma^{2(\alpha)}_{12} = \Gamma^{2(\alpha)}_{21} = \frac{(1-\alpha)}{8D_0^2\Delta}\left(-\frac{1}{v} + \frac{1}{4}\psi_1\left(\frac{v}{2}\right)\right), \Gamma^{2(0)}_{12} = \Gamma^{2(0)}_{21} = \frac{1}{8\Delta D_0^2}\left(-\frac{1}{v} + \frac{1}{4}\psi_1\left(\frac{v}{2}\right)\right) \tag{4.9}$$

$$\Gamma_{22}^{2(\alpha)} = \frac{(1-\alpha)}{32\Delta D_0^3}\left(9 - 2v\psi_1\left(\frac{v}{2}\right)\right), \Gamma_{22}^{2(0)} = \frac{1}{32\Delta D_0^3}\left(9 - 2v\psi_1\left(\frac{v}{2}\right)\right) \tag{4.10}$$

Using the above derivations, the GEs of the (GBM) can be devised, which are considered to be the information geometric equations of motion for the (GBM).

5. The GEs, the KD, and the JD of the (GBM). *KD* $_{GBM}$ and JD$_{GBM}$

By the definition, the GEs should take the form:

$$\frac{d^2\theta^k}{dt^2} + \Gamma_{ij}^{k(0)}\left(\frac{d\theta^i}{dt}\right)\left(\frac{d\theta^j}{dt}\right) = 0, i,j = 1,2,\dots,n \tag{5.1}$$

GES$_{k=1}$ reads as

$$\frac{d^2v}{dt^2} + \Gamma_{11}^{1(0)}\left(\frac{dv}{dt}\right)^2 + \Gamma_{22}^{1(0)}\left(\frac{dD_0}{dt}\right)^2 + \left(\Gamma_{12}^{1(0)} + \Gamma_{21}^{1(0)}\right)\left(\frac{dv}{dt}\right)\left(\frac{dD_0}{dt}\right) = 0 \tag{5.2}$$

Equation (5.2) presents a family of families of equations of motion of the rescaling parameter, RP = v which are dependent on the values of both the parameters v and D_0. One element of this family could be obtained by assuming that D_0=1.Thus,

$$\frac{d^2v}{dt^2} + \frac{v}{8\Delta D_0^2}\left(\frac{1}{v^2} + \frac{1}{8}\psi_2\left(\frac{v}{2}\right)\right)\left(\frac{dv}{dt}\right)^2 = 0 \tag{5.3}$$

with $\Delta = \det([g_{ij}]) = \frac{1}{16D_0^2}\left(v\psi_1\left(\frac{v}{2}\right) - 5\right)$

This transform (5.3) to be

$$\frac{d^2v}{dt^2} + \frac{2v}{\left(v\psi_1\left(\frac{v}{2}\right)-5\right)}\left(\frac{1}{v^2} + \frac{1}{8}\psi_2\left(\frac{v}{2}\right)\right)\left(\frac{dv}{dt}\right)^2 = 0 \tag{5.4}$$

At k = 2, we have a family of families of equations of motion of the initial maximiser parameter, IMP = D_0 which are dependent on the values both the parameters v and D_0. One element of this family could be obtained by assuming that $v = 2$ (*or equivalently, q =* $\frac{5}{3}$) there is a second set of GEs corresponding to the 2nd coordinate D_0, namely:

$$\frac{d^2D_0}{dt^2} + \frac{1}{\left(2\psi_1(1)-5\right)D_0^3}\left(9 - 4\psi_1(1)\right)\left(\frac{dD_0}{dt}\right)^2 = 0 \tag{5.4}$$

By $\psi_1(1) = \frac{\pi^2}{6}$, (5.4)will take the form:

$$\frac{d^2D_0}{dt^2} + \frac{1}{(\pi^2-15)D_0^3}\left(27 - 2\pi^2\right)\left(\frac{dD_0}{dt}\right)^2 = 0 \tag{5.5}$$

Using Rung-Kutta method of order 4 (c.f.,[72-74]), the numerical solution of (5.4) can be visualized for $n = 10$, as in figure 13 below.
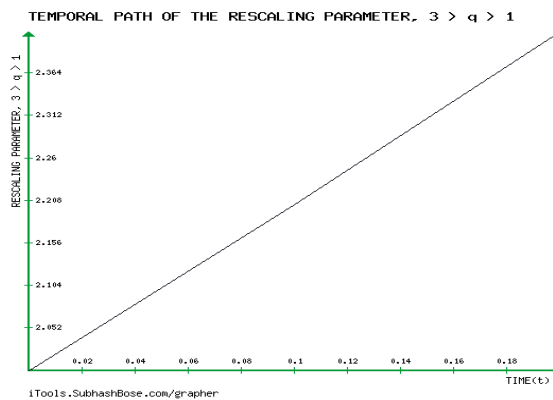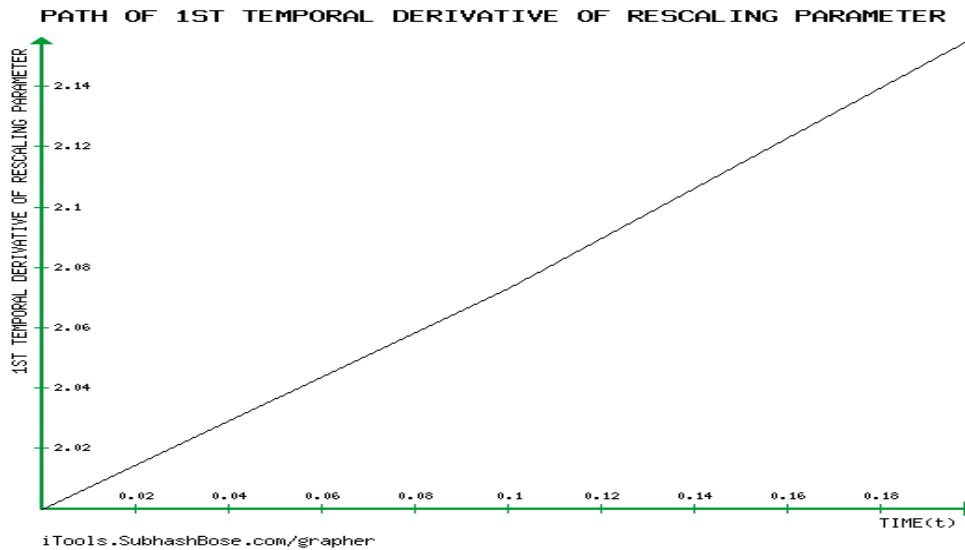

Figure 13: The Temporal Path of the Rescaling Parameter

Following figures 13 and 14, the analysis of the dynamics of both the position of information theoretic rescaling parameter and its velocity shows an increase against time, which describes a form of unbounded motion. This also justifies the physical concept of (GBM). This also allows us for first time ever to capture the hidden dynamics of GBM from an IG viewpoint.

Figure 14: The Temporal Path of the Velocity of the Rescaling Parameter

Analysis of the dynamics of both the position of information theoretic initial maximize (IMP) parameter and its velocity shows an increase against time in figures 15 and 16, a description of a form for an unbounded motion which justifies the physical concept of (GBM). Furthermore, the velocity of (IMP) decreases as time increase, which directly implies that the motion of the (IM) parameter will reach rest point at a certain time elapse. This also allows us for first time ever to capture the hidden dynamics of GBM from an IG viewpoint.
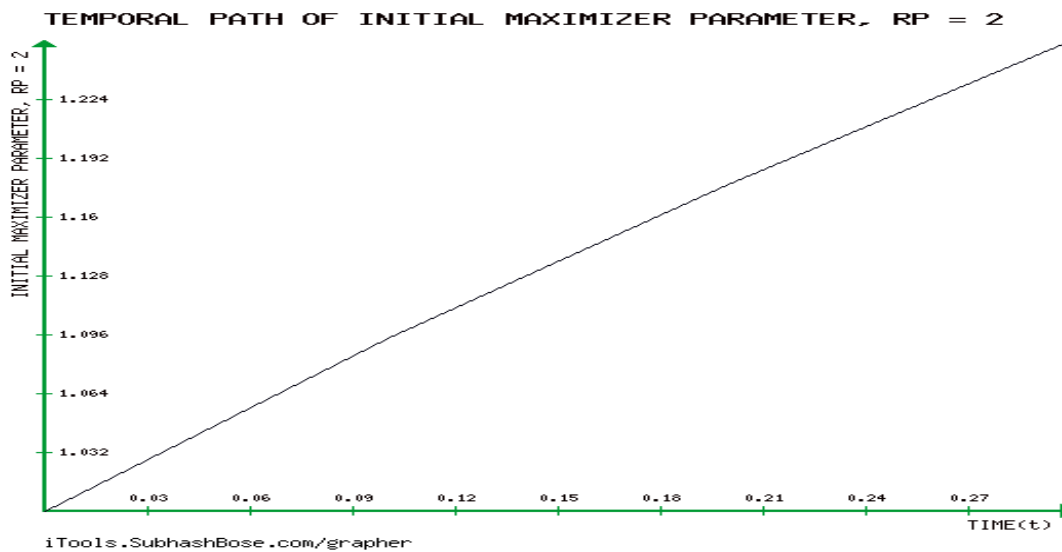
Figure 15: The Temporal Path of the Velocity of the Initial Maximizes Parameter
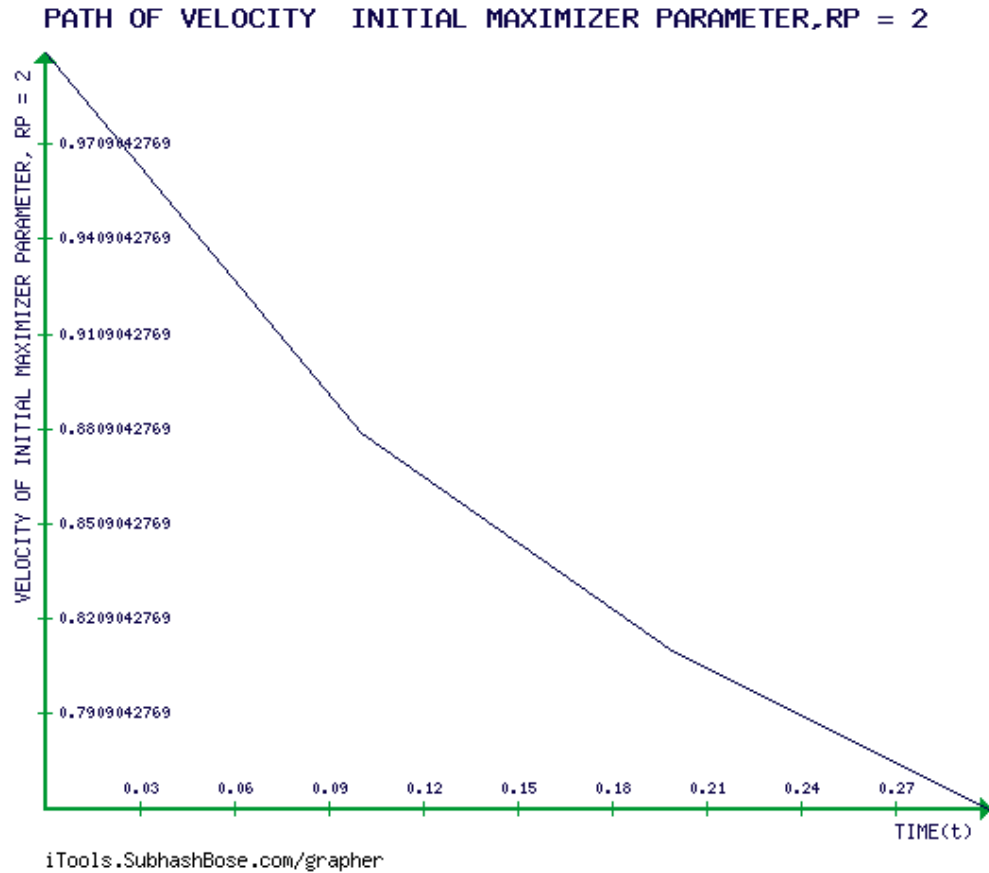
Figure 16: The Temporal Path of the Velocity of the Initial Maximizer Parameter (IMP), namely D_0

Now, by (2.6), after some few algebraic calculations, the $KD_{GBM}(p,r)$ reads as:

$$K_{GBM}\left(v_p, v_r, D_{0p}, D_{0r}\right) = \frac{\sqrt{(2^{(v_p, - v_r)})}\Gamma(\frac{v_r}{2})\sqrt{(\frac{v_p{}^{v_p}}{v_r{}^{v_r}})}(\frac{D_{0p}{}^{v_p}}{D_{0r}{}^{v_r}})^{\frac{1}{4}}}{\Gamma(\frac{v_p}{2})} \tag{5.6}$$

Moreover, in a similar fashion, the $JD_{GBM}(p,r)$ reads as

$$JD_{GBM}(p,r) = KD_{GBM}(p,r) + KD_{GBM}(r,p) = ln\left(\frac{p(x;\theta_p)}{r(x;\theta_r)}\right)^{(p(x;\theta_p) - r(x;\theta_r))} \neq 0 \tag{5.7}$$

(5.7) presents contributes to revealing that GBM manifold is compressible or non-solenoidal.
Getting more deeper with our mathematical calculations,

$$K_{GBM}\left(v_p, v_r, D_{0p}, D_{0r}\right)$$

$$= \frac{\sqrt{(2^{(v_p - v_r)})}\Gamma(\frac{v_r}{2})\sqrt{(\frac{v_p{}^{v_p}}{v_r{}^{v_r}})}(\frac{D_{0p}{}^{v_p}}{D_{0r}{}^{v_r}})^{\frac{1}{4}}}{\Gamma(\frac{v_p}{2})} + \frac{\sqrt{(2^{(v_r - v_p)})}\Gamma(\frac{v_p}{2})\sqrt{(\frac{v_r{}^{v_r}}{v_p{}^{v_p}})}(\frac{D_{0r}{}^{v_r}}{D_{0p}{}^{v_p}})^{\frac{1}{4}}}{\Gamma(\frac{v_r}{2})} \tag{5.8}$$

In what follows, we present some numerical experiments to illustrate the physical interpretations of our new findings.

By figures 17 and 18, the analysis shows that the underlying KD and ID of GBM are influenced by the information theoretic impact of both rescaling and initial maximize parameters, v and D_0 This is shown by the path continuity of the above graphs of coordinates as well as their temporal velocities.
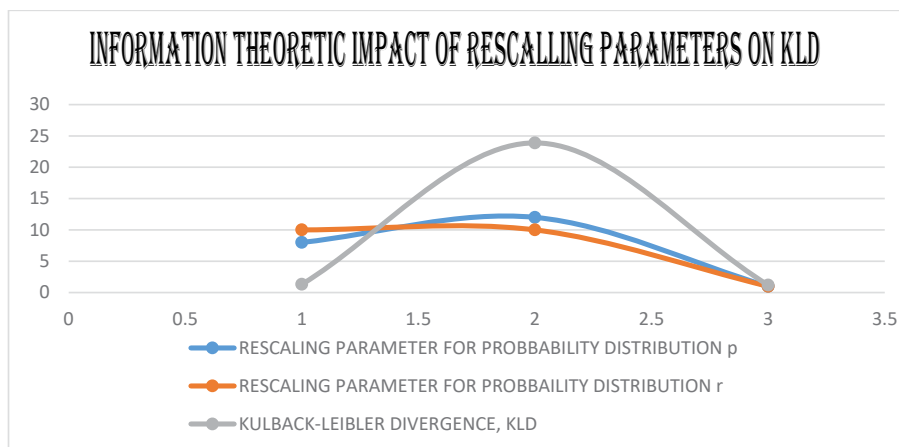
Figure 17

6. Novel Investigation of the Mathematical Requirements of the Developability of (GBM), Calculating 0-Gaussian Curvature of GBM, and Showing that RICCI CURVATURE (RC) Tensor of GBM is Non-Zero

In this section, new results providing under which mathematical requirements, GBM would be developable and has a non-zero 0-Gaussian and Ricci curvatures. Also, a novel sufficient and necessary conditions which characterizes the regions where the surface describing GBM is minimal are determined.

Theorem 6.1 GBM manifold has:

i) A non-zero 0-Gaussian curvature. ii) A non-zero Ricci tensor. (iii) GBM is developable under the Mong Patch Technique if the condition

$$v\psi_1\left(\frac{v}{2}\right) = 4 \tag{6.1}$$

is satisfied.

(iv) The surface describing GBM is minimal if and only if it holds that

$$v = 0 \text{ or } \frac{v}{8D_0^2} = \left(\frac{1}{2}\ln 2 + \frac{1}{2}\psi\left(\frac{v}{2}\right) - \frac{1}{2} - \ln v\right)^2 \tag{6.2}$$

(v) GBM is developable under the Angular Technique for the first principal curvature,K_1, if the condition

Either $v = 0$ or $D_0 = 0$ or

$$D_0^2\left(1 \mp \sqrt{\left(5 - v\psi_1\left(\frac{v}{2}\right)\right)}\right)^2 = \left(D_0^2\psi_1\left(\frac{v}{2}\right) - 4D_0^2 - v^2\right)\left(\mp\sqrt{\left(5 - v\psi_1\left(\frac{v}{2}\right)\right)}\right) + D_0^2\psi_1\left(\frac{v}{2}\right) - 4D_0^2 \tag{6.3}$$

are satisfied

Proof
Case i), by definition (2.10), part i), it is enough to show that the α-Gaussian curvature

$$K^{(\alpha)} = \frac{R_{1212}^{(\alpha)}}{det\left(g_{ij}\right)} \neq 0 \tag{6.4}$$

To this end, the $R_{ijkl}^{(\alpha)}$ can be expressed by $\tag{6.5}$

$$R_{ijkl}^{(\alpha)} = \left[\left(\partial_j \Gamma_{ik}^{s(\alpha)} - \partial_i \Gamma_{jk}^{s(\alpha)}\right)g_{sl} + \left(\Gamma_{jt,l}^{(\alpha)}\Gamma_{ik}^{t(\alpha)} - \Gamma_{it,l}^{(\alpha)}\Gamma_{jk}^{t(\alpha)}\right)\right], i, j, k, l, s, t = 1,2,3, \dots., n$$

where $\Gamma_{ij}^{k(\alpha)} = \Gamma_{ij,s}^{(\alpha)}g^{sk}$, $i, j, k, s = 1, 2, ..., n$.

In particular,

$$R_{1212}^{(\alpha)} = \left[\left(\partial_2\left(\Gamma_{11}^{1(\alpha)} + \Gamma_{11}^{2(\alpha)}\right) - \partial_1(\Gamma_{21}^{1(\alpha)} + \Gamma_{21}^{2(\alpha)})\right)(g_{12} + g_{22}) + \left(\Gamma_{21,2}^{(\alpha)}\Gamma_{11}^{1(\alpha)} + \Gamma_{22,2}^{(\alpha)}\Gamma_{11}^{2(\alpha)}\right) - \left(\Gamma_{11,2}^{(\alpha)}\Gamma_{21}^{1(\alpha)} + \Gamma_{12,2}^{(\alpha)}\Gamma_{21}^{2(\alpha)}\right)\right] \tag{6.6}$$

Following some lengthy calculations, it could be verified that we obtain a family of families of α-Gaussian curvatures determined by

$$
R_{1212}^{(\alpha)} = \left[\left(\frac{(D_0-\nu)(1-\alpha)}{4D_0^2(\nu\psi_1\left(\frac{\nu}{2}\right)-5)^2}\right)\left(\left(-\frac{1}{2D_0}\right)\left(\psi_1\left(\frac{\nu}{2}\right)+\frac{\nu}{2}\psi_2\left(\frac{\nu}{2}\right)\right)+\left(\nu\psi_1\left(\frac{\nu}{2}\right)-5\right)\left(\frac{1}{\nu^2}-\frac{\nu}{2}\psi_2\left(\frac{\nu}{2}\right)\right)\right.\right.
$$
$$
\left.+\left(\frac{1}{\nu}-\frac{1}{4}\psi_1\left(\frac{\nu}{2}\right)\right)\left(\psi_1\left(\frac{\nu}{2}\right)+\frac{\nu}{2}\psi_2\left(\frac{\nu}{2}\right)\right)\right)+
$$

$\frac{(1-\alpha)^2}{64\Delta D_0^4})((\nu-\frac{2}{D_0})(\frac{1}{\nu^2}+\frac{1}{8}\psi_2\left(\frac{\nu}{2}\right))+(\frac{1}{\nu}-\frac{1}{4}\psi_1\left(\frac{\nu}{2}\right))]]$

(6.7)

Here, a numerical experiment will indicate the required result without any loss of generality due to the higher complexity of calculations.

We have $K^{(0)} = \frac{R_{1212}^{(0)}}{det\left(g_{ij}\right)} = \frac{1}{\Delta}([\left(\frac{(D_0-\nu)}{4D_0^2(\nu\psi_1\left(\frac{\nu}{2}\right)-5)^2}\right)\left(\left(-\frac{1}{2D_0}\right)\left(\psi_1\left(\frac{\nu}{2}\right)+\frac{\nu}{2}\psi_2\left(\frac{\nu}{2}\right)\right)+\left(\nu\psi_1\left(\frac{\nu}{2}\right)-5\right)\left(\frac{1}{\nu^2}-\right.\right.$

$\frac{\nu}{2}\psi_2\left(\frac{\nu}{2}\right))+\left(\frac{1}{\nu}-\frac{1}{4}\psi_1\left(\frac{\nu}{2}\right)\right)\left(\psi_1\left(\frac{\nu}{2}\right)+\frac{\nu}{2}\psi_2\left(\frac{\nu}{2}\right)\right))+$

$(\frac{1}{64\Delta D_0^4})((\nu-\frac{2}{D_0})(\frac{1}{\nu^2}+\frac{1}{8}\psi_2\left(\frac{\nu}{2}\right))+(\frac{1}{\nu}-\frac{1}{4}\psi_1\left(\frac{\nu}{2}\right))]])$

(6.8)

For $D_0 =1$, we obtain the following family of zero-Gaussian curvatures devised by

$$
K_{(D_0=1)}^{(0)} = \frac{1}{(\Delta_{(D_0=1)})^2}([\left(\frac{(1-\nu)}{4(\nu\psi_1\left(\frac{\nu}{2}\right)-5)^2}\right)\left(\left(-\frac{1}{2}\right)\left(\psi_1\left(\frac{\nu}{2}\right)+\frac{\nu}{2}\psi_2\left(\frac{\nu}{2}\right)\right)+\left(\nu\psi_1\left(\frac{\nu}{2}\right)-5\right)\left(\frac{1}{\nu^2}-\frac{\nu}{2}\psi_2\left(\frac{\nu}{2}\right)\right)+\right.
$$

(6.9)

$\left(\frac{1}{\nu}-\frac{1}{4}\psi_1\left(\frac{\nu}{2}\right)\right)\left(\psi_1\left(\frac{\nu}{2}\right)+\frac{\nu}{2}\psi_2\left(\frac{\nu}{2}\right)\right))+$

$(\frac{1}{64})((\nu-2)(\frac{1}{\nu^2}+\frac{1}{8}\psi_2\left(\frac{\nu}{2}\right))+(\frac{1}{\nu}-\frac{1}{4}\psi_1\left(\frac{\nu}{2}\right))]])$

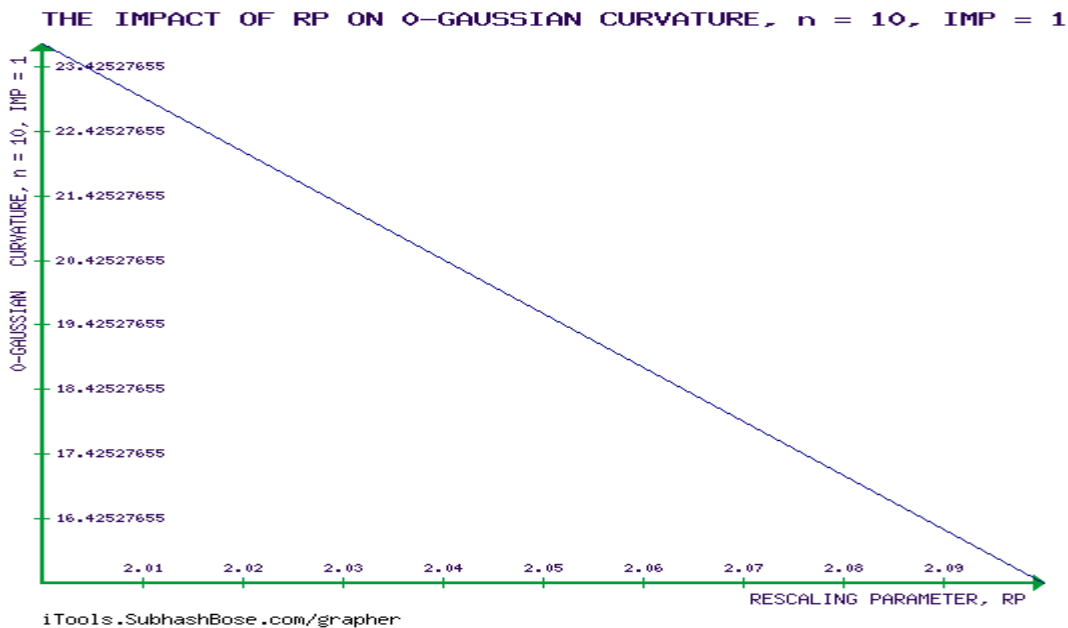**Figure 19: How (RP), namely ν impacts the 0-Gausian Curvature for IMP = D_0=1,n=10**

Hence, $K^{(\alpha)} = \dfrac{R_{1212}^{(\alpha)}}{det\left(g_{ij}\right)}$ is non zero.

Case ii)To prove that $R_{ik}^{(\alpha)}$ is non-zero of Eqn(2.12)

We need to show that at least one component is a non-zero. Following (6.1), it can be verified that $R_{11}^{(0)}$ is expressed by:

$$R_{11}^{(0)} = \frac{4}{D_0\left(v\psi_1\left(\frac{v}{2}\right)-5\right)}\left[\left(\frac{1}{4}\psi_1\left(\frac{v}{2}\right)-\frac{1}{4D_0}-\frac{1}{v}\right)\left(\frac{\left(\psi_1\left(\frac{v}{2}\right)+\frac{v}{2}\psi_2\left(\frac{v}{2}\right)\right)}{D_0\left(v\psi_1\left(\frac{v}{2}\right)-5\right)^2}+\frac{\left(\frac{1}{v^2}+\frac{1}{8}\psi_2\left(\frac{v}{2}\right)\right)\left(\psi_2\left(\frac{v}{2}\right)+\frac{8}{v^2}-1\right)}{D_0^2\left(v\psi_1\left(\frac{v}{2}\right)-5\right)}\right.+$$

$$\frac{4v}{\left(v\psi_1\left(\frac{v}{2}\right)-5\right)}\left(\frac{(D_0-v)}{4D_0^2\left(v\psi_1\left(\frac{v}{2}\right)-5\right)^2}\left(-\frac{1}{2D_0}\right)\left(\psi_1\left(\frac{v}{2}\right)+\frac{v}{2}\psi_2\left(\frac{v}{2}\right)\right)+\left(\left(v\psi_1\left(\frac{v}{2}\right)-5\right)\right)\left(\frac{1}{v^2}-\frac{v}{2}\psi_2\left(\frac{v}{2}\right)\right)+\left(\frac{1}{v}-\right.\right.$$

$$\left.\left.\frac{1}{4}\psi_1\left(\frac{v}{2}\right)\right)\left(\psi_1\left(\frac{v}{2}\right)+\frac{v}{2}\psi_2\left(\frac{v}{2}\right)\right)\right)+\frac{4\left(\left(v-\frac{2}{D_0}\right)\left(\frac{1}{v^2}+\frac{1}{8}\psi_2\left(\frac{v}{2}\right)+\frac{1}{v}-\frac{1}{4}\psi_1\left(\frac{v}{2}\right)\right)\right)}{D_0^2\left(v\psi_1\left(\frac{v}{2}\right)-5\right)}\right] \qquad (6.10)$$

The devised value of $R_{11}^{(0)}$ For IMP =1, we have,

$$R_{11}^{(0)} = \frac{4}{\left(v\psi_1\left(\frac{v}{2}\right)-5\right)}\left[\left(\frac{1}{4}\psi_1\left(\frac{v}{2}\right)-\frac{1}{4}-\frac{1}{v}\right)\left(\frac{\left(\psi_1\left(\frac{v}{2}\right)+\frac{v}{2}\psi_2\left(\frac{v}{2}\right)\right)}{\left(v\psi_1\left(\frac{v}{2}\right)-5\right)^2}+\frac{\left(\frac{1}{v^2}+\frac{1}{8}\psi_2\left(\frac{v}{2}\right)\right)\left(\psi_2\left(\frac{v}{2}\right)+\frac{8}{v^2}-1\right)}{\left(v\psi_1\left(\frac{v}{2}\right)-5\right)}\right.+$$

$$\frac{4v}{\left(v\psi_1\left(\frac{v}{2}\right)-5\right)}\left(\frac{(1-v)}{4\left(v\psi_1\left(\frac{v}{2}\right)-5\right)^2}\left(-\frac{1}{2}\right)\left(\psi_1\left(\frac{v}{2}\right)+\frac{v}{2}\psi_2\left(\frac{v}{2}\right)\right)+\left(\left(v\psi_1\left(\frac{v}{2}\right)-5\right)\right)\left(\frac{1}{v^2}-\frac{v}{2}\psi_2\left(\frac{v}{2}\right)\right)+\left(\frac{1}{v}-\right.\right.$$

$$\left.\left.\frac{1}{4}\psi_1\left(\frac{v}{2}\right)\right)\left(\psi_1\left(\frac{v}{2}\right)+\frac{v}{2}\psi_2\left(\frac{v}{2}\right)\right)\right)+\frac{4\left((v-2)\left(\frac{1}{v^2}+\frac{1}{8}\psi_2\left(\frac{v}{2}\right)+\frac{1}{v}-\frac{1}{4}\psi_1\left(\frac{v}{2}\right)\right)\right)}{\left(v\psi_1\left(\frac{v}{2}\right)-5\right)}\right] \qquad (6.11)$$
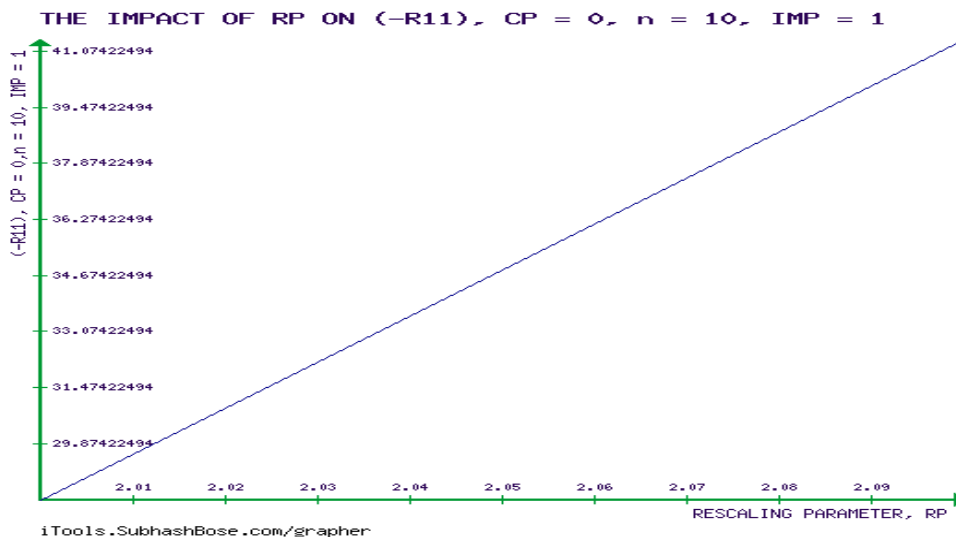

Figure 20: How v impacts $R_{11}^{(0)}$ for $D_0 = 1$, $n = 10$

It is observed that $R_{11}^{(0)}$ is non-zero according to the curve. It would be zero. Whenever

$$\left(v\psi_1\left(\tfrac{v}{2}\right) - 5\right) \to \infty \tag{6.12}$$

Following (1.13), $\psi_1\left(\tfrac{v}{2}\right) \to \infty$ if and only if $n \to \infty$. So, (6.12) holds if and only if n is infinite. Therefore, $R_{11}^{(0)}$ is everywhere non-zero unless n is infinite. This completes the proof.

The remaining components of the Ricci Curvature Tensor could be computed in a similar fashion.

(iii) Following (2.21), $K_G = \frac{LM - M^2}{EG - F^2}$, with $E = (\frac{\partial f}{\partial x})^2, F = \frac{\partial f}{\partial x \partial y}, G = (\frac{\partial f}{\partial y})^2, L = \frac{\partial^2 f}{\partial x^2}, M = \frac{\partial f}{\partial xy}, N = \frac{\partial^2 f}{\partial y^2}$.

$$K_G = \frac{LN - M^2}{EG - F^2} \tag{2.23}$$

and its Mean *Curvature* is

$$H = \frac{1}{2}\left(\frac{LG - 2MF + NE}{EG - F^2}\right) \tag{2.24}$$

with $E = (\frac{\partial f}{\partial x})^2, F = \frac{\partial f}{\partial x \partial y}, G = (\frac{\partial f}{\partial y})^2, L = \frac{\partial^2 f}{\partial x^2}, M = \frac{\partial f}{\partial xy}, N = \frac{\partial^2 f}{\partial y^2}$

Setting, $x = v, y = D_0$ and f to be the potential function $\Phi(\theta)$ (c.f.,(3.5)). It is immediate that,

$$E = (\tfrac{1}{2}\ln 2 + \tfrac{1}{2}\psi(\tfrac{v}{2}) - \tfrac{1}{2} - \ln v)^2, F = -\tfrac{1}{v} + \tfrac{1}{4}\psi_1(\tfrac{v}{2}), G = (\tfrac{v}{4D_0})^2, L = -\tfrac{1}{4D_0}, M = 0, N = \tfrac{v}{4D_0^2} \tag{6.13}$$

By (2.23) and (6.13), clearly it follows that:

$$K_G = \frac{2(v\psi_1\left(\tfrac{v}{2}\right) - 4)}{v^2((\ln 2 - 3) + \psi(\tfrac{v}{2}))} \tag{6.14}$$

It can be seen from (6.13) that $K_G = 0$ if and only if $v\psi_1\left(\tfrac{v}{2}\right) = 4$, which shows (iii).

To prove (iv), it suffices to show that the Mean Curvature H of (2.24) is zero, if and only if

$$v = 0 \ \text{ or } \ \frac{v}{8D_0^2} = (\tfrac{1}{2}\ln 2 + \tfrac{1}{2}\psi(\tfrac{v}{2}) - \tfrac{1}{2} - \ln v)^2 \qquad\qquad \text{(c.f., (6.2))}$$

After some lengthy calculations, it could be verified that

$$H = \frac{\frac{v}{4D_0}(\frac{v}{8D_0^2} - (\frac{1}{2}\ln 2 + \frac{1}{2}\psi(\frac{v}{2}) - \frac{1}{2} - \ln v)^2)}{((\frac{1}{2}\ln 2 + \frac{1}{2}\psi(\frac{v}{2}) - \frac{1}{2} - \ln v)^2\left(\frac{v}{4D_0}\right)^2 - \left(-\frac{1}{v} + \frac{1}{4}\psi_1\left(\frac{v}{2}\right)\right)^2)} \tag{6.15}$$

Clearly, (iv) follows from (6.15).

(v) we have $K_1 = \frac{B_{11}}{(1 + (\frac{\partial f}{\partial v'})^2)^{\frac{3}{2}}}$, $K_2 = \frac{B_{22}}{(1 + (\frac{\partial f}{\partial D_0'})^2)^{\frac{3}{2}}}$(c.f.,(2.27)), f to be the potential function$\Phi(\theta)\left(c.f., (3.5)\right)$.

After some manipulation, it could be verified that

$$\frac{\partial f}{\partial v'} = \left(\tfrac{1}{2}\ln 2 + \tfrac{1}{2}\psi\left(\tfrac{v}{2}\right) - \tfrac{1}{2} - \ln v\right)\cos\Game + (\tfrac{v}{4D_0})\sin\Game \tag{6.16}$$

$$\frac{\partial f}{\partial D_0'} = \left(\tfrac{1}{2}\ln 2 + \tfrac{1}{2}\psi\left(\tfrac{v}{2}\right) - \tfrac{1}{2} - \ln v\right)\sin\Game - (\tfrac{v}{4D_0})\cos\Game \tag{6.17}$$

$$B_{11} = \left(-\tfrac{1}{v} + \tfrac{1}{4}\psi_1\left(\tfrac{v}{2}\right)\right)\cos^2\Game + \frac{\sin\Game\cos\Game}{2D_0} + \frac{\sin^2\Game}{4D_0^2} \tag{6.18}$$

$$B_{22} = \left(-\frac{1}{v} + \frac{1}{4}\psi_1\left(\frac{v}{2}\right)\right)\sin^2 \omega - \frac{\sin\omega\cos\omega}{2D_0} + \frac{\cos^2\omega}{4D_0^2} \tag{6.19}$$

We also have, $\tan 2\omega = \dfrac{-2\left(\frac{\partial^2 f}{\partial v \partial D_0}\right)}{\left(\frac{\partial^2 f}{\partial v^2} - \frac{\partial^2 f}{\partial D_0^2}\right)}$(c.f., (2.32)). This directly implies that,

$$\frac{2\tan\omega}{1 - \tan^2\omega} = \tan 2\omega = \frac{2vD_0}{\left(D_0^2\psi_1\left(\frac{v}{2}\right) - 4D_0^2 - v^2\right)} \tag{6.20}$$

The developability of GBM is satisfied if and only if $K_G = 0$ or if either $K_1 = 0$ or $K_2 = 0$. Choosing $K_1 = 0$. Consequently, $B_{11} = 0$. Clearly, it follows that:

$$\tan^2\omega + \frac{2D_0}{v}\tan\omega = \frac{4D_0^2}{v^2}\left(1 - \frac{v}{4}\psi_1\left(\frac{v}{2}\right)\right) \tag{6.21}$$

Hence, $\tan\omega = -\frac{D_0}{v}(1 \mp \sqrt{(5 - v\psi_1\left(\frac{v}{2}\right))})$ (6.22). thus, it is obtained by linking (6.20) and (6.21), that

$$\frac{-\frac{2D_0}{v}(1\mp\sqrt{(5-v\psi_1(\frac{v}{2}))})}{1-(\frac{D_0}{v}(1\mp\sqrt{(5-v\psi_1(\frac{v}{2}))}))^2} = \frac{2vD_0}{(D_0^2\psi_1(\frac{v}{2})-4D_0^2-v^2)} \tag{6.23}$$

(6.23) implies $= 0$, or $D_0 = 0$, or $\dfrac{1}{(D_0^2\psi_1(\frac{v}{2})-4D_0^2-v^2)} = \dfrac{(1\mp\sqrt{(5-v\psi_1(\frac{v}{2}))})}{D_0^2(1\mp\sqrt{(5-v\psi_1(\frac{v}{2}))})^2-v^2}$ \tag{6.24}

$$\frac{1}{(D_0^2\psi_1(\frac{v}{2})-4D_0^2-v^2)} = \frac{(1\mp\sqrt{(5-v\psi_1(\frac{v}{2}))})}{D_0^2(1\mp\sqrt{(5-v\psi_1(\frac{v}{2}))})^2-v^2} \text{ implies}$$

$$D_0^2(1 \mp \sqrt{(5 - v\psi_1\left(\frac{v}{2}\right))})^2 - v^2 = (D_0^2\psi_1\left(\frac{v}{2}\right) - 4D_0^2 - v^2)(1 \mp \sqrt{(5 - v\psi_1\left(\frac{v}{2}\right))}) \tag{6.25}$$

Thus,

$$D_0^2(1 \mp \sqrt{(5 - v\psi_1\left(\frac{v}{2}\right))})^2 = (D_0^2\psi_1\left(\frac{v}{2}\right) - 4D_0^2 - v^2)(\mp\sqrt{(5 - v\psi_1\left(\frac{v}{2}\right))}) + D_0^2\psi_1\left(\frac{v}{2}\right) - 4D_0^2 \tag{6.26}$$

This completes the proof.

7. The Exponential Matrix of the FIM of the GBM ($e^{FIMGBM\,QM}$)

Theorem 7.1 $e^{FIM_{GBM}\,QM}$ solves $\frac{dx}{dt} = Ax$.

Proof
Recalling that GBM's FIM ,$\left[g_{ij}\right]$ reads

$$[g_{ij}] = \begin{pmatrix} -\frac{1}{v} + \frac{1}{4}\psi_1\left(\frac{v}{2}\right) & -\frac{1}{4D_0} \\ -\frac{1}{4D_0} & \frac{v}{4D_0^2} \end{pmatrix} \tag{7.1}$$

Rewriting $[g_{ij}]$ in a simpler form,

$$[g_{ij}] = \begin{pmatrix} a & b \\ b & c \end{pmatrix}, a = -\frac{1}{v} + \frac{1}{4}\psi_1\left(\frac{v}{2}\right), b = -\frac{1}{4D_0}, c = \frac{v}{4D_0^2} \tag{7.2}$$

If $\Phi(\delta) = (\delta) = \det([g_{ij}] - \delta I) = \det\begin{pmatrix} a-\delta & b \\ b & c-\delta \end{pmatrix} = 0$, then $\delta^2 - (a+c)\delta + (ac - b^2) = 0$.
Therefore,

$$\delta_{1,2} = \frac{(a+c) \pm \sqrt{(a^2+c^2+4b^2-2ac)}}{2}$$

Hence

$$D = \begin{pmatrix} \delta_1 & 0 \\ 0 & \delta_2 \end{pmatrix} , \quad e^D = \begin{pmatrix} e^{\delta_1} & 0 \\ 0 & e^{\delta_2} \end{pmatrix} \tag{7.3}$$

For $\delta_{1,2} = \frac{(a+c) \pm \sqrt{(a^2+c^2+4b^2-2ac)}}{2}$, the corresponding eigen vectors are $\begin{pmatrix} 1 \\ \frac{(\delta_1-a)}{b} \end{pmatrix}$, $\begin{pmatrix} 1 \\ \frac{(\delta_2-a)}{b} \end{pmatrix}$. Clearly, follows that the matrix

$$T = \begin{pmatrix} 1 & 1 \\ \frac{(\delta_1-a)}{b} & \frac{(\delta_2-a)}{b} \end{pmatrix} \tag{7.4}$$

Thus, $e^{FIM_{GBM\,QM}}$ reads:

$$e^{FIM_{GBM\,QM}} = Te^D T^{-1} = \frac{1}{\det(T)} \begin{pmatrix} 1 & 1 \\ \frac{(\delta_1-a)}{b} & \frac{(\delta_2-a)}{b} \end{pmatrix} \begin{pmatrix} e^{\delta_1} & 0 \\ 0 & e^{\delta_2} \end{pmatrix} \begin{pmatrix} \frac{(\delta_1-a)}{b} & -1 \\ \frac{(a-\delta_2)}{b} & 1 \end{pmatrix} \tag{7.5}$$

$$\det(T), = (a-c)^2 + 4b^2 \neq 0$$

By (7.5), the required statement is obtained.

8. Conclusions and Future Work

From the perspective of IG, GBM is described, and KD and J-D, the GEs, were established. Additionally, the GBM's matrix exponential of information is developed. This work creates new avenues for research by establishing connections between queueing theory and a wide range of other mathematical fields, including matrix theory, differential geometry, and information theory. In particular, the addition of geometric ties between queueing theory and information allows one to revealing a ground-breaking unified relativistic and of Riemannian geometric analysis of queue manifolds [3,4,7]. This paper's contributions suggest several potential study directions for the new applications of information geometry, such as using information geometrics on different statistical manifolds to further advance many existing physical phenomena.

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