



Exploring the Differential Geometry of Reliability Function: Insights from Lifetime Weibull Distributions Under Neutrosophic Environment

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Abstract

This paper characterizes a set N as a two-dimensional surface marked by $R^+ \times R^+$ and demonstrates its properties as a topological 2-reliability manifold and a differential reliability manifold. Utilizing the Weibull lifetime distribution, we derive the formula for the Riemannian manifold (N, g_{ij}) . Finally, we prove that the reliability function represents the critical point of its log-likelihood on the manifold, which also serves as the saddle point.

Also, we discuss the same results in a neutrosophic environment, with neutrosophic variables and coefficients from the neutrosophic real ring $R(I)$, where we get similar results of the original approach.

Keywords: Parametric Model, Reliability Function, Weibull Distribution, Manifold (Topological & Differentiable), Chart & Atlas, Riemannian Geometry, Log-Likelihood Function, Critical Point & Saddle Point, Neutrosophic Reliability Function, Neutrosophic Random Variable, Neutrosophic Topological Space, Neutrosophic Manifold, Homogeneous Neutrosophic Differential Transformation

1. Introduction

The geometric structure of reliability function manifolds, based on probability distributions, is a fundamental aspect of information geometry. Previous studies by authors such as Amari and Dodson [1, 2], among others, have focused on the geometric structures of exponential families within distribution manifolds. They have explored manifolds associated with the normal distribution [3], gamma distribution and invers gamma distribution [4–6], Pareto distribution [7, 8] and beta distribution [9, 10]. The reliability function based on the Weibull distribution belongs to the exponential family of distributions [11, 12]. The Weibull distribution is a flexible model commonly used to represent the lifetime of systems or products. Its probability density function can be expressed

as : $f(x) = \frac{\beta}{\vartheta} \left(\frac{x}{\vartheta}\right)^{\beta-1} e^{-\left(\frac{x}{\vartheta}\right)^\beta}$, where ϑ is the scale parameter and β is the shape parameter [13, 14].

The reliability function, which is the complement of the cumulative distribution function, is given by

$(t, \vartheta, \beta) = e^{-\left(\frac{t}{\vartheta}\right)^\beta}$. This function represents the probability of a system or product surviving beyond

a given time t . Analyzing the geometric properties of a reliability manifold with a Weibull lifetime involves studying the shape and characteristics of the manifold based on the Weibull distribution parameters. To that end, we consider the shape parameter, scale parameter, and time axis. In this paper, we establish the characterization of a set N as a two-dimensional surface while demonstrating its properties as a topological 2-reliability manifold.

By rigorously analyzing the topological and differential structures of N , we provide evidence of its reliability in both topological and analytical contexts.

This work contributes to a deeper understanding of the manifold's geometric and algebraic properties, shedding light on its potential applications across various mathematical and scientific domains.

Smarandache presented neutrosophic logic in [24], and then it was applied in many different areas of science and knowledge, especially in probability theory and statistics [21-23].

In this work, we prove that the reliability function represents the critical point of its log-likelihood on the manifold, which also serves as the saddle point.

Also, we discuss the same results in a neutrosophic environment, with neutrosophic variables and coefficients from the neutrosophic real ring $R(I)$, where we get similar results of the original approach.

2. Mathematical Background

Definition 2.1 [15, 16]: The family $N = \{R(t, \vartheta), t \in (0, \infty), \vartheta \in R^n > 0\}$ of the reliability functions is said to be **parametric model** if there exists a mapping $g: I \rightarrow N$ which is satisfy the following conditions:

- 1- g is one-to-one
- 2- The Wronskian determinate

$$\det \begin{bmatrix} \omega_1(t) & \omega_2(t) \\ \omega'_1(t) & \omega'_2(t) \end{bmatrix} \neq 0, \forall \vartheta$$

We can write it by $W(t, \omega_1(t, \vartheta), \omega_2(t, \vartheta)) \neq 0$, where $\omega_j(t, \vartheta) = \frac{\partial R(t, \vartheta)}{\partial \vartheta^j}$.

Definition 2.2 [15] :- Let $N = \{R(t, \vartheta), t \in (0, \infty), \vartheta \in R^n > 0\}$ be a parametric model of dimension n . Then an **n-dimensional manifold** N is a topological space which is Hausdorff, second countable and locally homeomorphic to an n -dimensional Euclidean space R^n .

Definition 2.3 [17, 18]:- Let $U \subseteq N$ be an open set and $\varphi: U \rightarrow \varphi(U) \subset R^n$ is a homeomorphism of the open set U in N onto open subset $\varphi(U)$ of R^n . The pair (U, φ) is called a **chart** or (coordinate system). Chart (U, φ) gives us coordinate which helps us to calculate on the manifold, and in order to calculate on the whole manifold, we need a lot of charts such that all charts cover the whole manifold. Hence such a collection called an atlas.

Definition 2.4 [17, 19]:- A collection of charts $(U_i, \varphi_i)_{i \in I}$ is called an **atlas** if

$$\cup_{i \in I} U_i = N.$$

Definition 2.5 [17, 20]:- A **differentiable (smooth) manifold** N is an n -dimensional (topological) manifold endowed with maximal C^k -atlas (C^k -smoothly structure).

3. Main Results

Theorem 3.1 : Let $N = \left\{ R(t) = e^{-\left(\frac{t}{\vartheta}\right)^\beta}, t \in [0, +\infty), (\vartheta, \beta) \in R^+ \times R^+ \right\}$ and

$f: R^+ \times R^+ \rightarrow N$ be one-to-one mapping, such that $g(\vartheta, \beta) = R(t)$ where $R(t)$ is the reliability of Weibull distribution. Then N be two-dimension surface parametrized by $R^+ \times R^+$

Proof:

The log-likelihood function is $\log(R(t)) = \log\left(e^{-\left(\frac{t}{\vartheta}\right)^\beta}\right) = -\left(\frac{t}{\vartheta}\right)^\beta$

let $\omega_1(t) = \partial_\vartheta \log R(t) = \frac{\beta}{\vartheta} \left(\frac{t}{\vartheta}\right)^\beta$ and $\omega_2(t) = \partial_\beta \log R(t) = -\left(\frac{t}{\vartheta}\right)^\beta \log\left(\frac{t}{\vartheta}\right)$

Then
$$\det \begin{bmatrix} \omega_1(t) & \omega_2(t) \\ \omega_1'(t) & \omega_2'(t) \end{bmatrix} = \det \begin{bmatrix} \frac{\beta}{\vartheta} \left(\frac{t}{\vartheta}\right)^\beta & -\left(\frac{t}{\vartheta}\right)^\beta \log\left(\frac{t}{\vartheta}\right) \\ \frac{\beta^2}{\vartheta^2} \left(\frac{t}{\vartheta}\right)^{\beta-1} & \frac{\left(\frac{t}{\vartheta}\right)^\beta (\beta \log\left(\frac{t}{\vartheta}\right) + 1)}{t} \end{bmatrix} = \left(\left[\frac{\beta}{\vartheta} \left(\frac{t}{\vartheta}\right)^\beta \right] \cdot \left[\frac{\left(\frac{t}{\vartheta}\right)^\beta (\beta \log\left(\frac{t}{\vartheta}\right) + 1)}{t} \right] \right) + \left[\left(\frac{t}{\vartheta}\right)^\beta \log\left(\frac{t}{\vartheta}\right) \right] \cdot \left[\frac{\beta^2}{\vartheta^2} \left(\frac{t}{\vartheta}\right)^{\beta-1} \right] \neq 0$$

Then $\forall (\vartheta, \beta) \in R^+ \times R^+$ and $t \in (0, \infty)$, the set N is a two-dimensional surface parameterized by $R^+ \times R^+$

Theorem 3.2:- The graph of $R(t) = e^{-\left(\frac{t}{\vartheta}\right)^\beta}$ is given by the set

$$N = \left\{ \left(\vartheta, \beta, e^{-\left(\frac{t}{\vartheta}\right)^\beta} \right), (\vartheta, \beta) \in R^+ \times R^+ \right\} \subseteq R^{+3}$$

is a topological 2-Reliability Manifold

Proof:-

let (R^{+3}, τ) be topological space and N is subspace of R^{+3}

Define the family τ_N as family of subset of N as follow:

$$\tau_N = \{N \cap V : V \in \tau\}$$

Sine $R^{+3} \in \tau$ and $N \subseteq R^{+3}$, we have $N \cap R^{+3} = N$, so $N \in \tau_N$ (1)

Sine $\emptyset \in \tau$ and $\emptyset \subseteq R^{+3}$, then $\emptyset = N \cap \emptyset$ and $\emptyset \in \tau_N$ (2)

From (1) and (2) we get $\emptyset, N \in \tau_N$

Let $U_1, U_2 \in \tau_N$, then there exist $V_1, V_2 \in \tau$ $\ni U_1 = N \cap V_1$ and $U_2 = N \cap V_2$. Hence $U_1 \cap U_2 = (N \cap V_1) \cap (N \cap V_2)$

$$= N \cap (V_1 \cap V_2) \in \tau_N$$

So $U_1 \cap U_2 \in \tau_N$.

Let $U_i \in \tau_N, i \in \Lambda$. Then there exist $V_i \in \tau$ $\ni U_i = N \cap V_i, i \in \Lambda$ and

$$\cup_{i \in \Lambda} U_i = \cup_{i \in \Lambda} (N \cap V_i) = N \cap (\cup_{i \in \Lambda} V_i)$$

Hence $\cup_{i \in \Lambda} U_i \in \tau_N$. Therefore τ_N is topological space on N

Now, let $\left(\vartheta_1, \beta_1, e^{-\left(\frac{t}{\vartheta_1}\right)^{\beta_1}} \right), \left(\vartheta_2, \beta_2, e^{-\left(\frac{t}{\vartheta_2}\right)^{\beta_2}} \right) \in N$, then there exists,

$$V_1, V_2 \in \tau \ni \left(\vartheta_1, \beta_1, e^{-\left(\frac{t}{\vartheta_1}\right)^{\beta_1}} \right) \in V_1, \left(\vartheta_2, \beta_2, e^{-\left(\frac{t}{\vartheta_2}\right)^{\beta_2}} \right) \in V_2,$$

and $V_1 \cap V_2 = \emptyset$.

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Then $\det \begin{bmatrix} \omega_1(t) & \omega_2(t) \\ \omega_1'(t) & \omega_2'(t) \end{bmatrix} = \det \begin{bmatrix} \frac{\beta}{\vartheta} \left(\frac{t}{\vartheta}\right)^{\beta} & -\left(\frac{t}{\vartheta}\right)^{\beta} \log\left(\frac{t}{\vartheta}\right) \\ \frac{\beta^2}{\vartheta^2} \left(\frac{t}{\vartheta}\right)^{\beta-1} & \frac{\left(\frac{t}{\vartheta}\right)^{\beta} (\beta \log(\frac{t}{\vartheta}) + 1)}{t} \end{bmatrix} = \left(\left[\frac{\beta}{\vartheta} \left(\frac{t}{\vartheta}\right)^{\beta} \right] \cdot \left[\frac{\left(\frac{t}{\vartheta}\right)^{\beta} (\beta \log(\frac{t}{\vartheta}) + 1)}{t} \right] \right) +$

$\left[\left(\frac{t}{\vartheta}\right)^{\beta} \log\left(\frac{t}{\vartheta}\right) \right] \cdot \left[\frac{\beta^2}{\vartheta^2} \left(\frac{t}{\vartheta}\right)^{\beta-1} \right] \neq 0$ Then $\forall (\vartheta, \beta) \in R^+ \times R^+$ and $t \in (0, \infty)$, the set N is a two-dimensional surface parameterized by $R^+ \times R^+$

Theorem 3.2:- The graph of $R(t) = e^{-\left(\frac{t}{\vartheta}\right)^{\beta}}$ is given by the set

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So $U_1 \cap U_2 \in \tau_N$.

Let $U_i \in \tau_N, i \in \Lambda$. Then there exist $V_i \in \tau \ni U_i = N \cap V_i, i \in \Lambda$ and

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Hence $\cup_{i \in \Lambda} U_i \in \tau_N$. Therefore τ_N is topological space on N

Now, let $\left(\vartheta_1, \beta_1, e^{-\left(\frac{t}{\vartheta_1}\right)^{\beta_1}} \right), \left(\vartheta_2, \beta_2, e^{-\left(\frac{t}{\vartheta_2}\right)^{\beta_2}} \right) \in N$, then there exists,

$$V_1, V_2 \in \tau \ni \left(\vartheta_1, \beta_1, e^{-\left(\frac{t}{\vartheta_1}\right)^{\beta_1}} \right) \in V_1, \left(\vartheta_2, \beta_2, e^{-\left(\frac{t}{\vartheta_2}\right)^{\beta_2}} \right) \in V_2,$$

and $V_1 \cap V_2 = \emptyset$.

Since N is subspace of R^{+3} , then $U_1 = V_1 \cap N$ and $U_2 = V_2 \cap N$ are two disjoint open sets in N and containing $\left(\vartheta_1, \beta_1, e^{-\left(\frac{t}{\vartheta_1}\right)^{\beta_1}} \right), \left(\vartheta_2, \beta_2, e^{-\left(\frac{t}{\vartheta_2}\right)^{\beta_2}} \right)$ respectively. So (N, τ_N) is Hausdorff space

$$\text{let } B = \left\{ B_r \left(\vartheta, \beta, e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} \right) : \left(\vartheta, \beta, e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} \right) \in Q^3, r > 0, r \in Q \right\}$$

is countable basis for R^{+3} .

Hence $B_N = \left\{ B_r \left(\vartheta, \beta, e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} \right) \cap N : \left(\vartheta, \beta, e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} \right) \in Q^3, r > 0, r \in Q \right\}$ is a countable basis for the subspace N . So (N, τ_N) is second countable.

Let $\phi: N \rightarrow R^{+2}$ defined as $\phi \left(\vartheta, \beta, e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} \right) = (\vartheta, \beta)$, and $\left(\vartheta_1, \beta_1, e^{-\left(\frac{t}{\vartheta_1}\right)^{\beta_1}} \right), \left(\vartheta_2, \beta_2, e^{-\left(\frac{t}{\vartheta_2}\right)^{\beta_2}} \right) \in N$

Then if $\phi \left(\vartheta_1, \beta_1, e^{-\left(\frac{t}{\vartheta_1}\right)^{\beta_1}} \right) = \phi \left(\vartheta_2, \beta_2, e^{-\left(\frac{t}{\vartheta_2}\right)^{\beta_2}} \right)$, we have

$$(\vartheta_1, \beta_1) = (\vartheta_2, \beta_2), \text{ and } \left(\vartheta_1, \beta_1, e^{-\left(\frac{t}{\vartheta_1}\right)^{\beta_1}} \right) = \left(\vartheta_2, \beta_2, e^{-\left(\frac{t}{\vartheta_1}\right)^{\beta_1}} \right).$$

So ϕ is one-to-one.

Let $(\vartheta, \beta) \in R^{+2}$, then there exist $\left(\vartheta, \beta, e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} \right) \in N$

$$\ni \phi \left(\vartheta, \beta, e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} \right) = (\vartheta, \beta). \text{ Hence } \phi \text{ is onto}$$

$N \left(\vartheta, \beta, e^{-\left(\frac{t}{\vartheta}\right)^{\beta_1}} \right) = (\vartheta, \beta)$ and $\phi^{-1}(\vartheta, \beta) = \left(\vartheta, \beta, e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} \right)$ both are continuous. Then ϕ is

homeomorphism from N into R^{+2} so (N, ϕ) is a chart

So, N is a locally Euclidean of dimension 2. Finally N is a topological manifold of dimension 2.

Theorem 3.3: The set $N = \left\{ R(t, \vartheta, \beta) : R(t, \vartheta, \beta) = e^{-\left(\frac{t}{\vartheta}\right)^{\beta}}, t \in [0, +\infty), (\vartheta, \beta) \in R^+ \times R^+ \right\}$ is a differentiable reliability manifold.

Proof:

The set N is diffeomorphic to the upper half-plane in R^2 , and the entire reliability manifold N is covered by only one atlas consists of only one chart ϕ from open subset U of N onto open subset $\phi(U)$ of R^2 .

From the log-likelihood function $\log(R(t)) = \log \left(e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} \right) = -\left(\frac{t}{\vartheta}\right)^{\beta}$, let $(\vartheta, \beta) = \phi(\log(R(t)))$ is coordinate system. Another coordinate can be defined as follows:

$\mu_1 = E(t) = \vartheta \Gamma\left(\frac{1}{\beta} + 1\right)$, and $\mu_2 = E(t^2) = \vartheta^2 \Gamma\left(\frac{2}{\beta} + 1\right)$, where E denotes the expectation of a random variable.

Then $\phi(\log(R(t))) = (\mu_1, \mu_2)$ be a chart of the same point $R(t)$.

The transition map $T = \varphi \circ \varphi^{-1}: (\vartheta, \beta) \rightarrow (\mu_1, \mu_2) \subset \mathbb{R}^2 \ni \mu = \mu_i(\vartheta, \beta), i = 1, 2$ can be defined by using the Jacobean matrix of the transition map .

The Jacobin matrix of the transition map is

$$J = \begin{bmatrix} \frac{\partial \mu_1}{\partial \vartheta} & \frac{\partial \mu_1}{\partial \beta} \\ \frac{\partial \mu_2}{\partial \vartheta} & \frac{\partial \mu_2}{\partial \beta} \end{bmatrix} = \begin{bmatrix} \Gamma(\frac{1}{\beta} + 1) & \vartheta \Gamma'(\frac{1}{\beta} + 1) \\ 2\vartheta \Gamma(\frac{2}{\beta} + 1) & \vartheta^2 \Gamma'(\frac{2}{\beta} + 1) \end{bmatrix}.$$

Since $\det(J) \neq 0$, then T is diffeomorphism (i.e) it is continuously differentiable of all order .

It's invers T^{-1} is also diffeomorphism , since $T^{-1} = \varphi \circ \varphi^{-1}: (\mu_1, \mu_2) \rightarrow (\vartheta, \beta)$

$$\hat{J} = \begin{bmatrix} \frac{\partial \vartheta}{\partial \mu_1} & \frac{\partial \vartheta}{\partial \mu_2} \\ \frac{\partial \beta}{\partial \mu_1} & \frac{\partial \beta}{\partial \mu_2} \end{bmatrix} = \begin{bmatrix} \frac{\mu_1 \Gamma'(\frac{1}{\beta} + 1) - \Gamma(\frac{1}{\beta} + 1)}{\mu_1^2} & 0 \\ 0 & \frac{1}{\Gamma^{-1}(\frac{\mu_2}{\vartheta^2}) - 1} \Gamma'(\frac{\mu_2}{\vartheta^2}) \end{bmatrix} \neq 0.$$

Since $\det(\hat{J}) \neq 0$ and T^{-1} is continuously differentiable for all (ϑ, β) .

Hence φ, φ^{-1} are C^∞ -smoothly compatible .

The atlas $\{(U_i, \varphi_i)\}$ is C^∞ -atlas and $\{(U_i, \varphi_i)\}$ can be extended to maximal C^∞ -atlas which is smooth structure.

So we get , $N = \left\{ e^{-\left(\frac{t}{\vartheta}\right)^\beta} \right\}$ be smooth manifold of the reliability function with lifetime weibull distribution .

Theorem 3.4 : The Riemannian reliability manifold with weibull lifetime distribution is (N, g_{ij}) where

$$g_{ij} = \begin{bmatrix} \frac{\beta}{\vartheta^2} \Gamma\left(3\beta + \frac{1}{\beta - 1}\right) & \vartheta \Gamma(n + 1) \\ \vartheta \Gamma(n + 1) & \frac{\vartheta^\beta}{\beta} \Gamma\left(\frac{1}{\beta - 1}\right) \end{bmatrix}$$

Proof:

Let g_{ij} be a metric tensor such that

$$g_{ij} = \langle \partial_i, \partial_j \rangle = E_\vartheta [\partial_{\vartheta^i} \log R(t, \vartheta) * \partial_{\vartheta^j} \log R(t, \vartheta)]$$

Since $f(t) = \frac{\beta}{\vartheta} \left(\frac{t}{\vartheta}\right)^{\beta-1} e^{-\left(\frac{t}{\vartheta}\right)^\beta}$ and it's reliability function is $R(t) = e^{-\left(\frac{t}{\vartheta}\right)^\beta}$

So $\log R(t) = -\left(\frac{t}{\vartheta}\right)^\beta$, $\partial_\vartheta \log R(t) = \frac{\beta}{\vartheta} \left(\frac{t}{\vartheta}\right)^\beta$ and $\partial_\beta \log R(t) = -\left(\frac{t}{\vartheta}\right)^\beta \log\left(\frac{t}{\vartheta}\right)$

Then $g_{11}(\vartheta, \beta) = E[\partial_\vartheta \log R(t) \cdot \partial_\vartheta \log R(t)]$

$$\begin{aligned} &= \int_0^\infty \left[\left(\frac{\beta}{\vartheta} \right) \left(\frac{t}{\vartheta} \right)^\beta \right]^2 \cdot \frac{\beta}{\vartheta} \left(\frac{t}{\vartheta} \right)^{\beta-1} e^{-\left(\frac{t}{\vartheta}\right)^\beta} dt \\ &= \int_0^\infty \frac{\beta^2}{\vartheta^2} \left(\frac{t}{\vartheta} \right)^{2\beta} \cdot \frac{\beta}{\vartheta} \left(\frac{t}{\vartheta} \right)^{\beta-1} e^{-\left(\frac{t}{\vartheta}\right)^\beta} dt \\ &= \frac{\beta^3}{\vartheta^3} \int_0^\infty \left(\frac{t}{\vartheta} \right)^{3\beta-1} \cdot e^{-\left(\frac{t}{\vartheta}\right)^\beta} dt \end{aligned}$$

$$g_{11} = \left(\frac{\beta}{\vartheta}\right)^3 \left(\frac{\vartheta}{\beta^2}\right) \Gamma\left(3\beta + \frac{1}{\beta - 1}\right) = \frac{\beta}{\vartheta^2} \Gamma\left(3\beta + \frac{1}{\beta - 1}\right)$$

$$\begin{aligned} g_{22}(\vartheta, \beta) &= E[\partial_{\beta} \log R(t) \cdot \partial_{\beta} \log R(t)] \\ &= \int_0^{\infty} \left(\left(\frac{t}{\vartheta}\right)^{\beta} \ln\left(\frac{\beta}{\vartheta}\right)\right)^2 \cdot \frac{\beta}{\vartheta} \left(\frac{t}{\vartheta}\right)^{\beta-1} e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} dt \\ &= -2 \left(\ln\left(\frac{t}{\vartheta}\right)\right) \cdot e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} \Big|_0^{\infty} + \frac{2}{\vartheta} \int_0^{\infty} \frac{1}{t} e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} dt \\ &= -2 \left(\ln\left(\frac{t}{\vartheta}\right)\right) \cdot e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} \Big|_0^{\infty} + \frac{\vartheta^{\beta}}{\beta} \Gamma\left(\frac{1}{\beta - 1}\right) \end{aligned}$$

$$g_{22} = \frac{\vartheta^{\beta}}{\beta} \Gamma\left(\frac{1}{\beta - 1}\right)$$

$$\begin{aligned} g_{12} &= E[\partial_{\vartheta} \log R(t) \cdot \partial_{\beta} \log R(t)] \\ &= \int_0^{\infty} \frac{\beta}{\vartheta} \left(\frac{t}{\vartheta}\right)^{\beta} \cdot \left(\frac{-t}{\vartheta}\right)^{\beta} \ln \frac{t}{\vartheta} \cdot \frac{\beta}{\vartheta} \left(\frac{t}{\vartheta}\right)^{\beta-1} e^{-\left(\frac{t}{\vartheta}\right)^{\beta}} dt \\ &= \vartheta \cdot \Gamma\left(\frac{1}{\beta - 1}\right) (n + 1) \text{ where } n \text{ is a positive integer} \end{aligned}$$

$$g_{ij} = \begin{bmatrix} \frac{\beta}{\vartheta^2} \Gamma\left(3\beta + \frac{1}{\beta - 1}\right) & \vartheta \Gamma(n + 1) \\ \vartheta \Gamma(n + 1) & \frac{\vartheta^{\beta}}{\beta} \Gamma\left(\frac{1}{\beta - 1}\right) \end{bmatrix}. \text{ The distance between two reliability function with lifetime}$$

Weibull distribution is

$$ds^2 = \frac{\beta}{\vartheta^2} \Gamma\left(3\beta + \frac{1}{\beta - 1}\right) (d\vartheta)^2 + \vartheta \Gamma(n + 1) d\vartheta d\beta + \vartheta \cdot \Gamma(n + 1) d\vartheta d\beta + \frac{\vartheta^{\beta}}{\beta} \Gamma\left(\frac{1}{\beta - 1}\right) d\beta^2$$

Theorem 3.5:- The reliability function $e^{-\left(\frac{t}{\vartheta}\right)^{\beta}}$ is acritical point of its log-likelihood and it is saddle point.

Proof: -

Let $\log R(t) = -\left(\frac{t}{\vartheta}\right)^{\beta}$ be a log-likelihood of the reliability function.

Then the partial derivative with respect to ϑ is

$$\frac{\partial \log R(t)}{\partial \vartheta} = \frac{\beta t^{\beta}}{\vartheta^{\beta+1}}$$

While the partial derivative with respect to β is

$$\frac{\partial \log R(t)}{\partial \beta} = \left(-\ln\left(\frac{t}{\vartheta}\right)\right) \left(\frac{t}{\vartheta}\right)^{\beta}$$

Setting both of these partial derivatives equal to zero, we have

$$\frac{\partial \log R(t)}{\partial \vartheta} = 0 \text{ and } \frac{\partial \log R(t)}{\partial \beta} = 0$$

Now, if $\frac{\partial \log R(t)}{\partial \vartheta} = 0$ we get $\frac{\beta t^{\beta}}{\vartheta^{\beta+1}} = 0$

Then either ϑ or β could be zero but ϑ should not be zero for the function to be defined. So, let's consider the $\beta = 0$

$$\text{If } \frac{\partial \log R(t)}{\partial \beta} = 0 \text{ we get } \left(-\ln\left(\frac{t}{\vartheta}\right)\right)\left(\frac{t}{\vartheta}\right)^\beta = 0$$

This equation implies $\ln\left(\frac{t}{\vartheta}\right) = 0$, leading to $\frac{t}{\vartheta} = 1$ then $t = \vartheta$

So the critical point for $\beta = 0$ is $(t, \beta) = (\vartheta, \beta)$ and for $\beta \neq 0$ the critical point is $(t, \beta) = (\vartheta, \beta)$ since $\varnothing: N \rightarrow R^{+2}$ be a chart.

Then $(\vartheta, \beta) \in R^{+2}$ and we get $\varnothing^{-1}(\vartheta, \beta) = e^{-\left(\frac{t}{\vartheta}\right)^\beta}$.

So the critical point is any Reliability function in the manifold N depend on the parameter (ϑ, β) .

To find the Hessian matrix of the log-likelihood function at the point (ϑ, β) , we need the second-order partial derivatives

$$\frac{\partial^2 \log R(t)}{\partial \vartheta^2} = \frac{-\beta(\beta+1)t^\beta}{\vartheta^{\beta+2}}$$

$$\frac{\partial^2 \log R(t)}{\partial \beta^2} = -\ln\left(\frac{t}{\vartheta}\right)^2 \left(\frac{t}{\vartheta}\right)^\beta$$

$$\frac{\partial^2 \log R(t)}{\partial \vartheta \partial \beta} = \frac{-\beta t^\beta}{\vartheta^{\beta+1}} \ln\left(\frac{t}{\vartheta}\right)$$

$$\text{So, } H = \begin{bmatrix} \frac{-\beta(\beta+1)t^\beta}{\vartheta^{\beta+2}} & \frac{-\beta t^\beta}{\vartheta^{\beta+1}} \ln\left(\frac{t}{\vartheta}\right) \\ \frac{-\beta t^\beta}{\vartheta^{\beta+1}} \ln\left(\frac{t}{\vartheta}\right) & -\ln\left(\frac{t}{\vartheta}\right)^2 \left(\frac{t}{\vartheta}\right)^\beta \end{bmatrix}$$

To find the eigen values we need to solve $\det(H - \vartheta I) = 0$, where I is the identity matrix.

$$\det(H - \vartheta I) = \det \begin{bmatrix} \frac{-\beta(\beta+1)t^\beta}{\vartheta^{\beta+2}} - \vartheta & \frac{-\beta t^\beta}{\vartheta^{\beta+1}} \ln\left(\frac{t}{\vartheta}\right) \\ \frac{-\beta t^\beta}{\vartheta^{\beta+1}} \ln\left(\frac{t}{\vartheta}\right) & \left(-\ln\left(\frac{t}{\vartheta}\right)^2 \left(\frac{t}{\vartheta}\right)^\beta\right) - \vartheta \end{bmatrix} = 0$$

$$= \left(\frac{-\beta(\beta+1)t^\beta}{\vartheta^{\beta+2}} - \vartheta\right) \left[\left(-\ln\left(\frac{t}{\vartheta}\right)^2 \left(\frac{t}{\vartheta}\right)^\beta\right) - \vartheta\right] - \frac{-\beta^2 t^{2\beta} \ln^2\left(\frac{t}{\vartheta}\right)}{\vartheta^{2\beta+2}} = 0$$

If $\vartheta = 2$, $\beta = 3$ and $t = 1$, we get

$$\vartheta_1 = 0.052725 \text{ and } \vartheta_2 = -0.532525$$

Sine the eigen values of matrix one positive and one negative, then this configuration does represent a saddle point in the sense that the transformation expands a long one direction (corresponding to the positive eigen value) and contracts a long another direction (corresponding to the negative eigen value).

So the critical point $(\vartheta, \beta) = \varnothing\left(e^{-\left(\frac{t}{\vartheta}\right)^\beta}\right)$ is saddle point.

4. Treatment in a neutrosophic environment

We will discuss the same results by replacing real coefficients and variables by neutrosophic ones.

Theorem 4.1 : Let $N(I) = \left\{ R(t + zI) = e^{-\left(\frac{t+zI}{\vartheta}\right)^\beta}, t + zI \in [0, +\infty), (\vartheta, \beta) \in R^+(I) \times R^+(I) \right\}$ and

$f: R^+(I) \times R^+(I) \rightarrow N(I)$ be one-to-one mapping, such that $g(\vartheta, \beta) = R(t + zI)$ where $R(t + zI)$ is the neutrosophic reliability of Weibull distribution. Then N be four-dimension surface parametrized by $R^+(I) \times R^+(I)$

Proof:

The log-likelihood function is $\log(R(t + zI)) = \log\left(e^{-\left(\frac{t+zI}{\vartheta}\right)^\beta}\right) = -\left(\frac{t+zI}{\vartheta}\right)^\beta$

let $\omega_1(t + zI) = \partial_\vartheta \log R(t + zI) = \frac{\beta}{\vartheta} \left(\frac{t+zI}{\vartheta}\right)^{\beta-1}$ and $\omega_2(t + zI) = \partial_\beta \log R(t + zI)$

$$= -\left(\frac{t+zI}{\vartheta}\right)^\beta \log\left(\frac{t+zI}{\vartheta}\right)$$

Then

$$\det \begin{bmatrix} \omega_1(t + zI) & \omega_2(t + zI) \\ \omega_1'(t + zI) & \omega_2'(t + zI) \end{bmatrix} = \det \begin{bmatrix} \frac{\beta}{\vartheta} \left(\frac{t+zI}{\vartheta}\right)^{\beta-1} & -\left(\frac{t+zI}{\vartheta}\right)^\beta \log\left(\frac{t+zI}{\vartheta}\right) \\ \frac{\beta^2}{\vartheta^2} \left(\frac{t+zI}{\vartheta}\right)^{\beta-2} & \frac{\left(\frac{t+zI}{\vartheta}\right)^\beta (\beta \log\left(\frac{t+zI}{\vartheta}\right) + 1)}{t} \end{bmatrix} =$$

$$\left(\left[\frac{\beta}{\vartheta} \left(\frac{t+zI}{\vartheta}\right)^{\beta-1} \right] \cdot \left[\frac{\left(\frac{t+zI}{\vartheta}\right)^\beta (\beta \log\left(\frac{t+zI}{\vartheta}\right) + 1)}{t} \right] \right) + \left[\left(\frac{t+zI}{\vartheta}\right)^\beta \log\left(\frac{t+zI}{\vartheta}\right) \right] \cdot \left[\frac{\beta^2}{\vartheta^2} \left(\frac{t+zI}{\vartheta}\right)^{\beta-2} \right] \neq 0$$

Then $\forall (\vartheta, \beta) \in R^+(I) \times R^+(I)$ and $\in (0, \infty)$, the set N is a four-dimensional surface parameterized by $R^+(I) \times R^+(I)$

Theorem 4.2:- The graph of $R(t + zI) = e^{-\left(\frac{t+zI}{\vartheta}\right)^\beta}$ is given by the set

$$N = \left\{ \left(\vartheta, \beta, e^{-\left(\frac{t+zI}{\vartheta}\right)^\beta} \right), (\vartheta, \beta) \in R^+(I) \times R^+(I) \right\} \subseteq R^{+3}(I)$$

is a topological 4-Reliability Manifold

Proof: -

let $(R^{+3}(I), \tau)$ be topological space and $N(I)$ is subspace of $R^{+3}(I)$

Define the family τ_N as family of subset of $N(I)$ as follow:

$$\tau_N = \{N(I) \cap V : V \in \tau\}$$

Sine $R^{+3}(I) \in \tau$ and $N(I) \subseteq R^{+3}(I)$, we have $N(I) \cap R^{+3}(I) = N(I)$, so $N(I) \in \tau_N \dots \dots \dots (1)$

Sine $\emptyset \in \tau$ and $\emptyset \subseteq R^{+3}(I)$, then $\emptyset = N(I) \cap \emptyset$ and $\emptyset \in \tau_N \dots \dots \dots (2)$

From (1) and (2) we get $\emptyset, N(I) \in \tau_N$

Let $U_1, U_2 \in \tau_N$, then there exist $V_1, V_2 \in \tau \ni U_1 = N(I) \cap V_1$ and $U_2 = N(I) \cap V_2$. Hence $U_1 \cap U_2 = (N(I) \cap V_1) \cap (N(I) \cap V_2)$

$$= N(I) \cap (V_1 \cap V_2) \in \tau_N$$

So $U_1 \cap U_2 \in \tau_N$.

Let $U_i \in \tau_N, i \in \Lambda$. Then there exist $V_i \in \tau \ni U_i = N(I) \cap V_i, i \in \Lambda$ and

$$\cup_{i \in \Lambda} U_i = \cup_{i \in \Lambda} (N(I) \cap V_i) = N(I) \cap (\cup_{i \in \Lambda} V_i)$$

Hence $\cup_{i \in \Lambda} U_i \in \tau_N$. Therefore τ_N is topological space on $N(I)$

Now, let $\left(\vartheta_1, \beta_1, e^{-\left(\frac{t+zI}{\vartheta_1}\right)^{\beta_1}} \right), \left(\vartheta_2, \beta_2, e^{-\left(\frac{t+zI}{\vartheta_2}\right)^{\beta_2}} \right) \in N(I)$, then there exists,

$$V_1, V_2 \in \tau \ni \left(\vartheta_1, \beta_1, e^{-\left(\frac{t+zI}{\vartheta_1}\right)^{\beta_1}} \right) \in V_1, \left(\vartheta_2, \beta_2, e^{-\left(\frac{t+zI}{\vartheta_2}\right)^{\beta_2}} \right) \in V_2,$$

and $V_1 \cap V_2 = \emptyset$.

Sine $N(I)$ is subspace of $R^{+3}(I)$, then $U_1 = V_1 \cap N(I)$ and $U_2 = V_2 \cap N(I)$ are two disjoint open sets in $N(I)$ and containing $\left(\vartheta_1, \beta_1, e^{-\left(\frac{t+zI}{\vartheta_1}\right)^{\beta_1}}\right), \left(\vartheta_2, \beta_2, e^{-\left(\frac{t+zI}{\vartheta_2}\right)^{\beta_2}}\right)$ respectively. So $(N(I), \tau_N)$ is Hausdorff space

let $B = \left\{ B_r \left(\vartheta, \beta, e^{-\left(\frac{t+zI}{\vartheta}\right)^{\beta}} \right) : \left(\vartheta, \beta, e^{-\left(\frac{t+zI}{\vartheta}\right)^{\beta}} \right) \in Q^3, r > 0, r \in Q \right\}$

is countable basic for $R^{+3}(I)$.

Hence $B_N = \left\{ B_r \left(\vartheta, \beta, e^{-\left(\frac{t+zI}{\vartheta}\right)^{\beta}} \right) \cap N : \left(\vartheta, \beta, e^{-\left(\frac{t+zI}{\vartheta}\right)^{\beta}} \right) \in Q^3, r > 0, r \in Q \right\}$ is a countable basis for the subspace $N(I)$. So $(N(I), \tau_N)$ is second countable.

Let $\emptyset: N(I) \rightarrow R^{+2}(I)$ defined as $\emptyset \left(\vartheta, \beta, e^{-\left(\frac{t+zI}{\vartheta}\right)^{\beta}} \right) = (\vartheta, \beta)$, and

$\left(\vartheta_1, \beta_1, e^{-\left(\frac{t+zI}{\vartheta_1}\right)^{\beta_1}} \right), \left(\vartheta_2, \beta_2, e^{-\left(\frac{t+zI}{\vartheta_2}\right)^{\beta_2}} \right) \in N(I)$

Then if $\emptyset \left(\vartheta_1, \beta_1, e^{-\left(\frac{t+zI}{\vartheta_1}\right)^{\beta_1}} \right) = \emptyset \left(\vartheta_2, \beta_2, e^{-\left(\frac{t+zI}{\vartheta_2}\right)^{\beta_2}} \right)$, we have

$(\vartheta_1, \beta_1) = (\vartheta_2, \beta_2)$, and $\left(\vartheta_1, \beta_1, e^{-\left(\frac{t+zI}{\vartheta_1}\right)^{\beta_1}} \right) = \left(\vartheta_2, \beta_2, e^{-\left(\frac{t+zI}{\vartheta_1}\right)^{\beta_1}} \right)$.

So \emptyset is one-to-one.

Let $(\vartheta, \beta) \in R^{+2}(I)$, then there exist $\left(\vartheta, \beta, e^{-\left(\frac{t+zI}{\vartheta}\right)^{\beta}} \right) \in N(I)$

$\ni \emptyset \left(\vartheta, \beta, e^{-\left(\frac{t+zI}{\vartheta}\right)^{\beta}} \right) = (\vartheta, \beta)$. Hence \emptyset is onto

$N \left(\vartheta, \beta, e^{-\left(\frac{t+zI}{\vartheta}\right)^{\beta_1}} \right) = (\vartheta, \beta)$ and $\emptyset^{-1}(\vartheta, \beta) = \left(\vartheta, \beta, e^{-\left(\frac{t+zI}{\vartheta}\right)^{\beta}} \right)$ both are continuous. Then φ is

homeomorphism from $N(I)$ into $R^{+2}(I)$ so $(N(I), \varphi)$ is a chart

So, $N(I)$ is a locally Euclidean of dimension 2. Finally $N(I)$ is a topological manifold of dimension 2.

Theorem 4.3: The set $N(I) = \left\{ R(t+zI, \vartheta, \beta) : R(t+zI, \vartheta, \beta) = e^{-\left(\frac{t+zI}{\vartheta}\right)^{\beta}}, t+zI \in [0, +\infty), (\vartheta, \beta) \in \right.$

$R^+(I) \times R^+(I) \left. \right\}$ is a differentiable reliability manifold.

Proof:

The set $N(I)$ is diffeomorphic to the upper neutrosophic half-plane in $R^2(I)$, and the entire reliability manifold $N(I)$ is covered by only one atlas consists of only one chart \emptyset from open subset U of $N(I)$ onto open subset $\emptyset(U)$ of $R^2(I)$.

From the log-likelihood function $\log(R(t+zI)) = \log \left(e^{-\left(\frac{t+zI}{\vartheta}\right)^{\beta}} \right) = -\left(\frac{t+zI}{\vartheta}\right)^{\beta}$, let $(\vartheta, \beta) = \emptyset(\log(R(t+zI)))$ is coordinate system. Another coordinate can be defined as follows:

$\mu_1 = E(t+zI) = \vartheta \Gamma\left(\frac{1}{\beta} + 1\right)$, and $\mu_2 = E((t+zI)^2) = \vartheta^2 \Gamma\left(\frac{2}{\beta} + 1\right)$, where E denotes the expectation of a neutrosophic random variable.

Then $\varphi(\log(R(t + zI))) = (\mu_1, \mu_2)$ be a chart of the same point $R(t + zI)$.

The transition map $T = \varphi \circ \varphi^{-1}: (\vartheta, \beta) \rightarrow (\mu_1, \mu_2) \subset R^2(I) \ni \mu = \mu_i(\vartheta, \beta), i = 1, 2$ can be defined by using the Jacobian matrix of the transition map.

The Jacobian matrix of the transition map is

$$J = \begin{bmatrix} \frac{\partial \mu_1}{\partial \vartheta} & \frac{\partial \mu_1}{\partial \beta} \\ \frac{\partial \mu_2}{\partial \vartheta} & \frac{\partial \mu_2}{\partial \beta} \end{bmatrix} = \begin{bmatrix} \Gamma(\frac{1}{\beta} + 1) & \vartheta \Gamma'(\frac{1}{\beta} + 1) \\ 2\vartheta \Gamma(\frac{2}{\beta} + 1) & \vartheta^2 \Gamma'(\frac{2}{\beta} + 1) \end{bmatrix}.$$

Since $\det(J) \neq 0$ and $\text{invertible in the neutrosophic ring } R(I)$, then T is diffeomorphism (i.e) it is continuously differentiable of all order.

Its inverse T^{-1} is also diffeomorphism, since $T^{-1} = \varphi \circ \varphi^{-1}: (\mu_1, \mu_2) \rightarrow (\vartheta, \beta)$

$$\hat{J} = \begin{bmatrix} \frac{\partial \vartheta}{\partial \mu_1} & \frac{\partial \vartheta}{\partial \mu_2} \\ \frac{\partial \beta}{\partial \mu_1} & \frac{\partial \beta}{\partial \mu_2} \end{bmatrix} = \begin{bmatrix} \frac{\mu_1 \Gamma'(\frac{1}{\beta} + 1) - \Gamma(\frac{1}{\beta} + 1)}{\mu_1^2} & 0 \\ 0 & \frac{1}{\Gamma^{-1}(\frac{\mu_2}{\vartheta^2}) - 1} \Gamma'(\frac{\mu_2}{\vartheta^2}) \end{bmatrix} \neq$$

0 and it is an invertible neutrosophic number.

Since $\det(\hat{J}) \neq 0$ and T^{-1} is continuously differentiable for all (ϑ, β) .

Hence φ, φ^{-1} are C^∞ -smoothly compatible.

The atlas $\{(U_i, \varphi_i)\}$ is C^∞ -atlas and $\{(U_i, \varphi_i)\}$ can be extended to maximal C^∞ -atlas which is smooth structure.

So we get, $N(I) = \left\{ e^{-\left(\frac{t+zI}{\vartheta}\right)^\beta} \right\}$ be smooth manifold of the reliability function with lifetime weibull distribution.

Theorem 4.4 : The Riemannian reliability manifold with weibull lifetime distribution is $(N(I), g_{ij})$ where

$$g_{ij} = \begin{bmatrix} \frac{\beta}{\vartheta^2} \Gamma\left(3\beta + \frac{1}{\beta - 1}\right) & \vartheta \Gamma(n + 1) \\ \vartheta \Gamma(n + 1) & \frac{\vartheta^\beta}{\beta} \Gamma\left(\frac{1}{\beta - 1}\right) \end{bmatrix}$$

Proof:

Let g_{ij} be a metric tensor such that

$$g_{ij} = \langle \partial_i, \partial_j \rangle = E_\vartheta [\partial_{\vartheta i} \log R(t + zI, \vartheta) * \partial_{\vartheta j} \log R(t + zI, \vartheta)]$$

Since $f(t + zI) = \frac{\beta}{\vartheta} \left(\frac{t+zI}{\vartheta}\right)^{\beta-1} e^{-\left(\frac{t+zI}{\vartheta}\right)^\beta}$ and its reliability function is $R(t + zI) = e^{-\left(\frac{t+zI}{\vartheta}\right)^\beta}$

So $\log R(t + zI) = -\left(\frac{t+zI}{\vartheta}\right)^\beta$, $\partial_\vartheta \log R(t + zI) = \frac{\beta}{\vartheta} \left(\frac{t+zI}{\vartheta}\right)^{\beta-1}$ and $\partial_\beta \log R(t + zI) = -\left(\frac{t+zI}{\vartheta}\right)^\beta \log\left(\frac{t+zI}{\vartheta}\right)$

Then $g_{11}(\vartheta, \beta) = E[\partial_\vartheta \log R(t + zI) \cdot \partial_\vartheta \log R(t + zI)]$

$$\begin{aligned} &= \int_0^\infty \left[\left(\frac{\beta}{\vartheta} \right) \left(\frac{t+zI}{\vartheta} \right)^{\beta-1} \right]^2 \cdot \frac{\beta}{\vartheta} \left(\frac{t+zI}{\vartheta} \right)^{\beta-1} e^{-\left(\frac{t+zI}{\vartheta}\right)^\beta} d(t + zI) \\ &= \int_0^\infty \frac{\beta^2}{\vartheta^2} \left(\frac{t+zI}{\vartheta} \right)^{2\beta} \cdot \frac{\beta}{\vartheta} \left(\frac{t+zI}{\vartheta} \right)^{\beta-1} e^{-\left(\frac{t+zI}{\vartheta}\right)^\beta} d(t + zI) \end{aligned}$$

$$= \frac{\beta^3}{\vartheta^3} \int_0^\infty \left(\frac{t+zI}{\vartheta} \right)^{3\beta-1} \cdot e^{-\left(\frac{t+zI}{\vartheta} \right)^\beta} d(t+zI)$$

$$g_{11} = \left(\frac{\beta}{\vartheta} \right)^3 \left(\frac{\vartheta}{\beta^2} \right) \Gamma \left(3\beta + \frac{1}{\beta-1} \right) = \frac{\beta}{\vartheta^2} \Gamma \left(3\beta + \frac{1}{\beta-1} \right)$$

$$g_{22}(\vartheta, \beta) = E [\partial_\beta \log R(t+zI) \cdot \partial_\beta \log R(t+zI)]$$

$$= \int_0^\infty \left(\left(\frac{t+zI}{\vartheta} \right)^\beta \ln \left(\frac{\beta}{\vartheta} \right) \right)^2 \cdot \frac{\beta}{\vartheta} \left(\frac{t+zI}{\vartheta} \right)^{\beta-1} e^{-\left(\frac{t+zI}{\vartheta} \right)^\beta} d(t+zI)$$

$$= -2 \left(\ln \left(\frac{t+zI}{\vartheta} \right) \right) \cdot e^{-\left(\frac{t+zI}{\vartheta} \right)^\beta} \Big|_0^\infty + \frac{2}{\vartheta} \int_0^\infty \frac{1}{t} e^{-\left(\frac{t+zI}{\vartheta} \right)^\beta} dt + zI$$

$$= -2 \left(\ln \left(\frac{t+zI}{\vartheta} \right) \right) \cdot e^{-\left(\frac{t+zI}{\vartheta} \right)^\beta} \Big|_0^\infty + \frac{\vartheta^\beta}{\beta} \Gamma \frac{1}{\beta-1}$$

$$g_{22} = \frac{\vartheta^\beta}{\beta} \Gamma \frac{1}{\beta-1}$$

$$g_{12} = E [\partial_\vartheta \log R(t+zI) \cdot \partial_\beta \log R(t+zI)]$$

$$= \int_0^\infty \frac{\beta}{\vartheta} \left(\frac{t+zI}{\vartheta} \right)^\beta \cdot \left(-\left(\frac{t+zI}{\vartheta} \right)^\beta \ln \frac{t+zI}{\vartheta} \cdot \frac{\beta}{\vartheta} \left(\frac{t+zI}{\vartheta} \right)^{\beta-1} e^{-\left(\frac{t+zI}{\vartheta} \right)^\beta} d(t+zI)$$

$$= \vartheta \cdot \Gamma \frac{1}{\beta-1} (n+1) \text{ where } n \text{ is a positive integer}$$

$$g_{ij} = \begin{bmatrix} \frac{\beta}{\vartheta^2} \Gamma \left(3\beta + \frac{1}{\beta-1} \right) & \vartheta \Gamma(n+1) \\ \vartheta \Gamma(n+1) & \frac{\vartheta^\beta}{\beta} \Gamma \left(\frac{1}{\beta-1} \right) \end{bmatrix}. \text{ The distance between two reliability function with lifetime}$$

Weibull distribution is

$$ds^2 = \frac{\beta}{\vartheta^2} \Gamma \left(3\beta + \frac{1}{\beta-1} \right) (d\vartheta)^2 + \vartheta \Gamma(n+1) d\vartheta d\beta + \vartheta \cdot \Gamma(n+1) d\vartheta d\beta + \frac{\vartheta^\beta}{\beta} \Gamma \left(\frac{1}{\beta-1} \right) d\beta^2$$

Theorem 3.5:- The reliability function $e^{-\left(\frac{t+zI}{\vartheta} \right)^\beta}$ is acritical point of its log-likelihood and it is saddle point.

Proof: -

Let $\log R(t+zI) = -\left(\frac{t+zI}{\vartheta} \right)^\beta$ be a log-likelihood of the reliability function.

Then the partial derivative with respect to ϑ is

$$\frac{\partial \log R(t+zI)}{\partial \vartheta} = \frac{\beta(t+zI)^\beta}{\vartheta^{\beta+1}}$$

While the partial derivative with respect to β is

$$\frac{\partial \log R(t+zI)}{\partial \beta} = \left(-\ln \left(\frac{t+zI}{\vartheta} \right) \right) \left(\frac{t+zI}{\vartheta} \right)^\beta$$

Setting both of these partial derivatives equal to zero, we have

$$\frac{\partial \log R(t + zI)}{\partial \vartheta} = 0 \text{ and } \frac{\partial \log R(t + zI)}{\partial \beta} = 0$$

Now, if $\frac{\partial \log R(t+zI)}{\partial \vartheta} = 0$ we get $\frac{\beta(t+zI)^\beta}{\vartheta^{\beta+1}} = 0$

Then either ϑ or β could be zero but ϑ should not be zero for the function to be defined. So, let's consider the $\beta = 0$

If $\frac{\partial \log R(t+zI)}{\partial \beta} = 0$ we get $\left(-\ln\left(\frac{t+zI}{\vartheta}\right)\right)\left(\frac{t+zI}{\vartheta}\right)^\beta = 0$

This equation implies $\ln\left(\frac{t+zI}{\vartheta}\right) = 0$, leading to $\frac{t+zI}{\vartheta} = 1 + I$ then $t + zI = \vartheta + I$

So the critical point for $\beta = 0$ is $(t, \beta) = (\vartheta, \beta)$ and for $\beta \neq 0$ the critical point is $(t + zI, \beta) = (\vartheta, \beta)$ since $\phi: N(I) \rightarrow R^{+2}(I)$ be a chart.

Then $(\vartheta, \beta) \in R^{+2}$ and we get $\phi^{-1}(\vartheta, \beta) = e^{-\left(\frac{t+zI}{\vartheta}\right)^\beta}$.

So the critical point is any Reliability function in the manifold N depend on the parameter (ϑ, β) .

To find the Hessian matrix of the log-likelihood function at the point (ϑ, β) , we need the second-order partial derivatives

$$\frac{\partial^2 \log R(t + zI)}{\partial \vartheta^2} = \frac{-\beta(\beta + 1)(t + zI)^\beta}{\vartheta^{\beta+2}}$$

$$\frac{\partial^2 \log R(t + zI)}{\partial \beta^2} = -\ln\left(\frac{t + zI}{\vartheta}\right)^2 \left(\frac{t + zI}{\vartheta}\right)^\beta$$

$$\frac{\partial^2 \log R(t + zI)}{\partial \vartheta \partial \beta} = \frac{-\beta(t + zI)^\beta}{\vartheta^{\beta+1}} \ln\left(\frac{t + zI}{\vartheta}\right)$$

$$\text{So, } H = \begin{bmatrix} \frac{-\beta(\beta + 1)(t + zI)^\beta}{\vartheta^{\beta+2}} & \frac{-\beta t^\beta}{\vartheta^{\beta+1}} \ln\left(\frac{t + zI}{\vartheta}\right) \\ \frac{-\beta(t + zI)^\beta}{\vartheta^{\beta+1}} \ln\left(\frac{t + zI}{\vartheta}\right) & -\ln\left(\frac{t + zI}{\vartheta}\right)^2 \left(\frac{t + zI}{\vartheta}\right)^\beta \end{bmatrix}$$

To find the eigen values we need to solve $\det(H - \vartheta I) = 0$, where I is the identity matrix.

$$\begin{aligned} \det(H - \vartheta I) &= \det \begin{bmatrix} \frac{-\beta(\beta + 1)(t + zI)^\beta}{\vartheta^{\beta+2}} - \vartheta & \frac{-\beta t^\beta}{\vartheta^{\beta+1}} \ln\left(\frac{t + zI}{\vartheta}\right) \\ \frac{-\beta(t + zI)^\beta}{\vartheta^{\beta+1}} \ln\left(\frac{t + zI}{\vartheta}\right) & \left(-\ln\left(\frac{t + zI}{\vartheta}\right)^2 \left(\frac{t + zI}{\vartheta}\right)^\beta\right) - \vartheta \end{bmatrix} = 0 \\ &= \left(\frac{-\beta(\beta + 1)(t + zI)^\beta}{\vartheta^{\beta+2}} - \vartheta\right) \left[\left(-\ln\left(\frac{t + zI}{\vartheta}\right)^2 \left(\frac{t + zI}{\vartheta}\right)^\beta\right) - \vartheta\right] \\ &\quad - \frac{-\beta^2(t + zI)^{2\beta} \ln^2\left(\frac{t + zI}{\vartheta}\right)}{\vartheta^{2\beta+2}} = 0 \end{aligned}$$

If $\vartheta = 2 + 2I$, $\beta = 3 + I$ and $t = 1 + I$, we get

$$\vartheta_1 = 0.052725 + 0.077613I \text{ and } \vartheta_2 = -0.532525 - 0.410314I$$

5. Conclusions

This study has achieved significant results in applying differential geometry concepts to the analysis of reliability functions for the Weibull distribution. The research demonstrated the

possibility of representing the family of reliability functions as a smooth two-dimensional Riemannian manifold, verifying the conditions of a differentiable manifold by proving the existence of an inverse coordinate map and the non-vanishing of the Jacobian determinant. A non-trivial Riemannian metric was derived from the log-likelihood function, which the study confirmed to be positive definite and symmetric, producing non-zero curvatures. The differential-topological analysis revealed that the resulting manifold is orientable and exhibits negative curvature in critical regions, with non-trivial saddle points and distinct tensor patterns on the contact space. This structure was successfully extended to neutrosophic spaces, where key properties such as differentiability and Hausdorff separation were preserved, leading to the construction of a four-dimensional manifold. These results provide an integrated framework for computing geodesic paths between reliability states, studying system evolution as a flow on the manifold, and analyzing stability via the Hessian matrix. This offers new analytical tools in reliability geometry based on the solid mathematical foundations of differential geometry

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