- 1 The Transient Variation of the Complexes of the Low Latitude Ionosphere within the
- 2 Equatorial Ionization Anomaly Region of Nigeria.
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12 Abstract

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The quest to find an index for proper characterization and description of the dynamical response 13 of the ionosphere to external influences and its various internal irregularities has led to the study 14 of the day to day variations of the chaoticity and dynamical complexity of the ionosphere. This 15 study was conducted using Global Positioning System (GPS) Total Electron Content (TEC) time 16 17 series, measured in the year 2011, from 5 GPS receiver stations in Nigeria which lies within the 18 Equatorial Ionization Anomaly region. The nonlinear aspect of the TEC time series were 19 obtained by detrending the data. The detrended TEC time series were subjected to various analyses for phase space reconstruction and to obtain the values of chaotic quantifiers which are 20 Lyapunov exponents LE, correlation dimension, and Tsallis entropy for the study of dynamical 21 complexity. The results show positive Lyapunov exponents for all days which indicate chaoticity 22 23 of the ionosphere with no definite pattern for both quiet and disturbed days. However Values of LE were lower for the storm period compared to its nearest relative quiet periods for all the 24 stations. Considering all the days of the year the daily/transient variations show no definite 25 26 pattern for each month but day to day values of Lyapunov exponent for the entire year show a wavelike semiannual variation pattern with lower values around March, April, September and 27 October, a change in pattern which demonstrates the self-organized critical phenomenon of the 28 29 system. This can be seen from the correlation dimension with values between 2.7 and 3.2 with 30 lower values occurring mostly during storm periods demonstrating a phase transition from higher

dimension during the quiet periods to lower dimension during storms for most of the stations. 31 The values of Tsallis entropy show similar variation pattern with that of Lyapunov Exponent 32 with a lot of agreement in their comparison, with all computed values of Lyapunov exponent 33 correlating with values of Tsallis entropy within the range of 0.79 to 0.82. These results show 34 that Lyapunov quantifiers can be used together as indices in the study of the variations of the 35 dynamical complexity of the ionosphere. The presence of chaos and high variations in the 36 dynamical complexity, even at quiet periods in the ionosphere may be due to the internal 37 dynamics and inherent irregularities of the ionosphere which exhibit non-linear properties. 38 39 However, this inherent dynamics may be complicated by external factors like Geomagnetic 40 storms. This may be the main reason for the drop in the values of Lyapunov exponent and Tsallis 41 entropy during storms. The results also show a strong interplay between determinism and stochasticity, as the ionosphere shows its response to changes in solar activities and in its internal 42 43 dynamics. The dynamical behavior of the ionosphere throughout the year as described by these quantifiers, were discussed in this work. 44

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1.0 Introduction

The behavior of natural systems like the ionosphere is a function of changes that occur in the underlying dynamics that exists in such system. These underlying dynamics however can be sometimes complex and nonlinear due to superposition of different changes in dynamical variables that constitute it. When the dynamical states of a system changes suddenly due to sudden changes in the external factor affecting the system, then such a system is said to be deterministic.

However, there is no totally deterministic system in nature, because all natural systems exhibit a mixture of both deterministic properties. Although few natural systems have been found to be low dimensional deterministic in the sense of the theory, the concept of low-dimensional chaos has been proven to be fruitful in the understanding of many complex phenomena (Hegger et al., 1999) The degree of determinism or stochasticity in most natural systems is dependent on how much the system can be influenced by external factors, the nature of these external factors among others .The ionosphere like every other natural system posses its intrinsic dynamics and it can also be influenced by other external factors. The typical characteristics of a dynamical system

like the ionosphere is expected to naturally show the interplay between determinism and stochasticity simply because of the fact that the ionosphere which has an inherent internal dynamics is also influenced by the influx of stochastic drivers like the solar wind, since it is influenced by external dynamics like every other natural system. This has made pure determinism impossible in the ionosphere, a situation that is common to all natural system and its surrounding.

The intensity of the solar wind coming into the ionosphere varies with the solar activity and an extreme solar activity can lead to geomagnetic storms and substorms drive in high intensity plasma wind at enormous speed and it serves as major stochastic driver leading to storm. The solar wind is driven from the sun into the ionospheric system during the quiet and storm and during relatively quiet periods of each month of the year. However other processes which include various factors like local time variations of the neutral winds, ionization processes, production-recombination rates, photoionization processes, plasma diffusion and various electrodynamics processes. (Unnikrishnan, 2010). The mesosphere—and the lower thermospheric dynamics as reported by Kazimirovsky and Vergasova (2009) and also the influence of gravity waves as reported by Sindelarova (2009) can also be of great influence on the internal dynamics of the ionosphere.

Therefore, it is of great importance to study the chaoticity and dynamical complexity of the ionosphere and its variations in all geophysical conditions. However a good number of investigations have been carried out on concept of chaos in the upper atmosphere before now which includes the study on magnetospheric dynamics and the ionosphere. The study of chaos in magnetospheric index time series such as AE and AL were initially carried out by (Vasiliadis et al., 1990, Shan et al, 1999; Pavlos et al, 1992). These previous efforts made by the aforementioned researchers has led to the development of the concept of investigating and revealing the chaoticity and the complex dynamics of the ionosphere, and as a result, studies on the chaoticity of the ionosphere have been conducted, by some investigators like Bhattacharyya (1990) who studied chaotic behavior of ionospheric diversity fluctuation using amplitude and phase scintillation data, and found the existence of low dimension chaos. Also, Wernik and Yeh (1994) further revealed the chaotic behavior of the ionospheric turbulence using scintillation data and numerical modeling of scintillation at high latitude. They showed that the ionospheric

turbulence attractor (if it exists) cannot be reconstructed from amplitude scintillation data and their measured phase scintillation data adequately reproduce the assumed chaotic structure in the ionosphere. Also Kumar et al., (2004) reported the evidence of chaos in the ionosphere by showing the chaotic nature of the underlying dynamics of the fluctuations of TEC power spectrum indicating exponential decay and the calculated positive value of Lyapunov exponent. This is also supported by the results of the comparison of the chaotic characteristics of the time series of variations of TEC with the pseudochaotic characteristic of the colored noise time series. Xuann et al., (2006) studied chaos properties of ionospheric total electron content (TEC) using TEC data from 1996 to 2004, and analyze possibility to predict it by using chaos. They found the presence of chaos in the TEC measured in the study area, as indicated by the positive Lyapunov exponent computed from their data. The correlation dimension was 3.6092 from their estimation. They were also able to show that the TEC time series can be predicted using chaos.

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Also, Unnikrishnan et al (2006a,b) have analyzed the deterministic chaos in mid latitude and Unnikrishnan (2010), Unnikrishnan and Ravindran (2010), analyzed some TEC data from some Indian low latitude stations for quiet period and major storm period and found in Their results the presence of chaos which was indicated by a positive Lyapunov exponent, and they also inferred that storm periods exhibits lower values compared to quiet periods. The dynamical complexity of magnetospheric processes and the ionosphere have been studied by a number of researchers. Balasis et al., (2008) investigated the dynamical complexity of the magnetosphere by using Tsallis entropy as a dynamical complexity measure in D_{st} time series also Balasis et al., (2009) investigated the dynamical complexity in D_{st} further by considering different entropy measures. Coco et al (2011) using the information theory approach studied the dynamical changes of the polar cap potential which is characteristic of the polar region ionosphere by considering three cases (i) steady IMF B_z >0, (ii) steady IMF B_z <0 and (iii) a double rotation from negative to positive and then positive to negative Bz. They observed a neat dynamical topological transition when the IMF Bz turns from negative to positive and vice versa, pointing toward the possible occurrence of an order/disorder phase transition, which is the counterpart of the large scale convection rearrangement and of the increase of the global coherence. Further studies in chaotic behavior and nonlinear dynamics is however needed to improve our understanding of the

dynamical behavior of the ionosphere of low latitude ionosphere especially over Africa during quiet and storm for different season of the year some as to be able to characterize chaoticity for different season of the year for quiet and storm periods. Recently Ogunsua et al 2014 studied comparatively the chaoticity of the equatorial ionosphere over Nigeria using TEC data, considering five quietest day classification and five most disturbed day classification. They were able to show the presence of chaos as indicated the positive Lyapunov exponents and also were able to show that Tsallis entropy can be used as a viable measure of dynamical complexity in the ionosphere with potions showing lower values of Tsallis entropy indicating lower dynamical complexity, with a good relationship with Lyapunov exponents. They found a phase transition from higher dimension during quiet days to Lower dimension during storm.

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The low latitude region where Nigeria is situated is known as the equatorial anomaly region, where the magnetic field B is almost totally parallel to the equator. Off the equator the E region dynamo electric field maps -along the magnetic field up to the F-region altitude in the low latitude, the this eastward electric field (E) of the F region-interacts with the magnetic field B at the F region during the day. This results in the electrodynamic lifting of the F-region plasma over the equator, which known as EXB drift. The uplifted plasma over the equator moves along the magnetic line in response to gravity, diffusion and pressure gradients and hence, the fountain effect. The fountain effect being controlled by the EXB drift shows the dynamics of the diurnal variation equatorial anomaly (Abdu, 1997; Unnikrishnan 2010). There is a reduction in the F region ionization density at the magnetic equator and also much enhanced ionization density at the two anomaly crests within ±15° of the magnetic latitude north and south of the equator (Rama Rao et al., 2006). The equatorial ionization anomaly and other natural processes which includes various ionization processes and recombination; influx of solar wind, photoionization processes and so many-many other factors that occur due to variations in solar activities, have a great influence on the systems of the ionosphere, due to their effects on internal dynamics of the ionosphere. This portrays the ionosphere as a typical natural system with continuous interaction with its external environment which led to the study of the influence of the sun on the ionosphere (Ogunsua et al., 2014).

The ionosphere possesses a significant level of nonlinear variations that requires more investigation which can be studied and characterized using non linear approach like the chaoticity and dynamical complexity for the study of its dynamics. The need to study the daily variation in the dynamical complexity of the ionosphere arises from the established knowledge and understanding which shows that the ionosphere is a complex system with so many variations that can arise from various dynamical changes that can be due to various changes in different processes that contribute to the behavior and nature of the ionosphere. Rabiu et al., (2007) affirmed that characterizing the ionosphere is of utmost importance due to the numerous complexities associated with the region. These numerous changes interestingly at times do not occur on the same scale for day to day. The concept of chaos as applied to ionospheric and magnetospheric studies on quiet and stormy conditions are limited.

Most investigations have been based on only quiet and storm conditions for all studies carried out, none of the previous works involved the quiet and disturbed day classification of geophysical conditions until recently by Ogunsua et al.,(2014), and also, the day to day variation of these phenomena have not been considered.

The comparative study of chaoticity and dynamical complexity recently conducted by Ogunsua et al., (2014) on the low latitude ionosphere over Nigeria using the Tsallis entropy for the first time as a quantifier compared with Lyapunov exponent suggested the applicability of these quantifiers in this present work as proxies for the internal dynamics of the ionosphere. Also the day to day variations of these phenomena were studied to reveal the possible underlying seasonal variation of these dynamics.

2.0 Data and Methodology

The data used for this study is the global positioning system (GPS) total electron content (TEC) data obtained from 5 GPS satellite receiver stations. Table 1 shows the coordinates of the stations. These receivers take the measure of slant TEC within 1m² columnar unit of the cross section along the ray path of the satellite and the receiver which is given by

176 STEC=
$$\int_{receiver}^{Satellite} Ndl$$
 (1)

- 177 The observation of the total number of free electron along the ray path are derived from the
- 178 frequency L₁(1572.42 MHz) and L₂(1227.60 MHz) of Global Positioning System(GPS),that
- 179 provide the relative ionosphere delay of electromagnetic waves travelling through the medium
- 180 (Saito et al., 1998). The Slant TEC is projected to vertical TEC using the thin shell model
- assuming the height of 350m (Klobuchar, 1986).
- 182 $VTEC = STEC \cdot \cos[\arcsin(R_e \cos\Theta/R_e + h_{max})]$ (2)
- Where $R_e = 6378km$ (radius of the earth), $h_{max} = 350km$ (the vertical height assumed from
- the satellite) and $\Theta = elevation$ angle at ground station

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- 186 In this study, 5 GPS TEC measuring stations lying within the low latitude region were
- 187 considered, as shown in table 1. The TEC data obtained for January to December 2011 were
- 188 considered for this study and the data are given at 1min sampling time. The TEC data were
- subjected to various analyses which will be discussed in the next section. The day to day
- 190 variations of the chaotic behavior and dynamical complexity were studied for the entire year.
- 191 The surrogate data tests for non linearity were also conducted for both the dynamical and
- 192 geometrical aspects.

3.0 Methods of Data Analysis and Results

3.1 Time series analysis

- 195 Time series can be seen as a numerical account that describes the state of a system, from which it
- was measured. A given time series, S_n can be defined as a sequence of scalar measurement of a
- particular quantity taken as series at different portion in time for a given time interval(Δt). The
- 198 time series describe the physical appearance of an entire system, as seen in Fig 1. However it
- 199 may not always describe the internal dynamics of that system. A system like the ionosphere
- 200 possesses a dominant dynamics that can be seen as diurnal so the data should be treated so as to
- 201 be able to see its internal dynamics. The measured TEC time series were plotted to see the
- dynamics of the system. A typical plot of TEC usually has a dominant dynamics (see fig 1)
- 203 which may be seen as the diurnal behavior, however, it can also be seen that there is also a
- 204 presence of fluctuations (which appear to be nonlinear) in the system as a result of the internal

dynamics of the ionosphere and space plasma system, due to different activities in the ionosphere. Therefore there is need to minimize the influence of the diurnal variations since we are more interested in the nonlinear internal dynamics of the system in this study, to do so the TEC time series was detrended by carrying out the following analysis below:

Since for the given daily data of 1minute sampling time there are 1440 data points per day. Then there exists a time series t_i , where i = 1,2,3.....1440 represents the observed time series, and there also exists a set of u_i where i = 1,2,3.....1440, such that the diurnal variation reduced time

212 is given by

$$T_i = t_i - u_i \tag{3}$$

Where i = 1,2,3,...,j = mod(i,1440), if $mod(j,1440) \neq 0$, and j = 1440 if d(j,1440) = 0.

This method will give the detrended time series represented by T_i obtained from the original

TEC data as shown in fig 2. This method is similar to that used by (Unnikrishnan et al., 2006,

Unnikrishnan 2010), the further explanations on the dynamical results can be found in (Kumar et

al., 2004). The detrended time series were subjected to further analyses for the Phase space

reconstruction and also to obtain the values of Lyapunov exponents, correlation dimension,

220 Tsallis entropy and the implementation of surrogate data test.

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3.1.1 Phase Space reconstruction and Non Linear Time Series Analysis

The study of chaoticity and dynamical complexity in a dynamical system requires a non linear approach, due to the fact that systems described by these phenomena can be referred to as nonlinear complex systems. The magnetosphere and the ionosphere are good examples of such systems. To be able to study such phenomena some nonlinear time series analysis can be carried out on the time series data describing such a system. The detrended time series of TEC measurement is subjected to some nonlinear time series data analysis to obtain the mutual information and false nearest neigbours, embedding dimension and delay coordinates for the phase space reconstruction, and the evaluation of other chaotic quantifiers namely: Lyapunov Exponents, Correlation dimension, recurrence analysis and Entropy.

- 232 The phase space reconstruction helps to reveal the multidirectional aspect of the system. The
- 233 phase space reconstruction is based on embedding theorem, such that the phase space is
- reconstructed to show the multidimensional nature as follows:

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$$Y_n = (s_{n-(m-1)\tau}, s_{n-(m-2)\tau}, ..., s_{n-\tau}, s_n)$$
 (4)

- where Y_n are vector in phase space. The proper choice of embedding dimension (m) and delay
- Time (τ) are essential for phase space reconstruction (Fraser and Swinney,1986; Kennel et
- 238 al.,1992).

- 239 If the plot showing the time delayed mutual information shows a marked minimum that value
- 240 can be considered as a responsible time delay; Fig 3 shows the mutual information plotted
- 241 against time delay. Likewise, the minimal embedding dimension, which correspond to the
- 242 minimum number of the false nearest neighbours Fig 4 can be treated as the optimum value of
- embedding dimension in (Unnikrishnan et al.,2006, Unnikrishnan, 2010). It was observed that
- for all the daily detrended TEC time series the choice of $\tau \ge 30$ and $m \ge 4$ values of delay and
- embedding dimension above these values are suitable for analysis of data for all stations. The
- 246 choice of $\tau = 30$ and m = 5 were mostly used to analyze the dynamical aspects for all the
- stations. The reconstructed Phase space trajectory is shown in Fig 5

3.1.2 Lyapunov Exponents

- 249 The Lyapunov exponent has been a very important quantifier for the determination of chaos in a
- 250 dynamical system. This quantifier is also used for the determination of chaos in time series,
- representing natural systems like the ionosphere and magnetosphere (Unnikrishnan 2008, 2010).
- 252 A positive Lyapunov exponent indicates divergence of trajectory in one dimension, or alternative
- 253 an expansion of volume, which can also be said to indicate repulsion, or attraction from a fixed
- 254 point. A positive Lyapunov exponent indicates that there is evidence of chaos in a dissipative
- 255 deterministic system, where the positive Lyapunov exponent indicates divergence of trajectory in
- one direction or expansion of value and a negative value shows convergence at trajectory or
- 257 contraction of volume along another direction.

- The largest Lyapunov exponent (λ_1) can be used to determine the rate of divergence as indicated
- 259 by (Wolf et al.,1985)
- 260 Where

$$\lambda_1 = \lim_{r \to \infty} \frac{1}{t} \ln \frac{\Delta x(t)}{x(0)} = \lim_{r \to \infty} \frac{1}{t} \sum_{i=1}^t \ln \left(\frac{\Delta x(t_i)}{\Delta x(t_{i-1})} \right)$$
 (5)

- 262 The Lyapunov exponent was computed for the TEC values measured from Different stations.
- The evolution in state space was scanned with $\tau = 30$, m = 5, is shown in fig 6. The day to day
- 264 variations of the Lyapunov exponent was computed for the entire year to so as to study the
- 265 annual trend of variation. This was implemented using the method introduced by Rosenstein
- 266 (1993), and Hegger et al., (1994), both algorithms use very similar methods. Lyapunov
- 267 exponents were also computed for varying time delay at constant embedding dimension and also
- 268 for varying embedding dimension, to check for the stability with changes in trajectory. These can
- be seen in fig 6b and 6c. The day to day values of Lyapunov exponent plotted for the Enugu
- 270 station and for Toro station are shown in fig 7a to 7b. The plots of the day to day values show the
- transient variation of the ionosphere and a wavelike yearly pattern.

3.1.3 Correlation Dimension

- 273 Another relevant method to study the underlying dynamics or internal dynamics of a system is to
- evaluate the dimension of the system. The correlation dimension gives a good approximation of
- 275 this as suggested by Grassberger and Procaccia (1983a; b). The correlation dimension is
- 276 preferred over the box counting dimension because it takes into account the density of points on
- 277 the attractor (Strogatz 1994). The correlation dimension D is defined as

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$$D = \lim_{r \to 0} \frac{\ln C(r)}{\ln r}$$
 (6)

- The term C(r) is the correlation sum for radius (r) where for a small radius (r) the correlation
- sum can be seen as $C(r) \sim r^d$ for $r \to 0$. The correlation sum is dependent of the embedding
- dimension (m) of the reconstructed phase space and it is also dependent of the length of the time
- 282 series N as follows

283
$$C(r) = \frac{2}{N(N-1)} \sum_{i=1}^{N} \sum_{j=i+1}^{N} \Theta(r - ||y_i - y_j||)$$
 (7)

- Where Θ is the Heaviside step function, with $\Theta(H) = 0$ if $H \le 0$ and $\Theta(H) = 1$ for H > 0.
- 285 The correlation dimension was computed using the Theiler algorithm approach, with Theiler
- window (w) at 180. The Theiler window was chosen to be approximately equal to the product of
- 287 m and τ . A similar approach to the computation of correlation dimension was used by
- 288 Unnikrishnan and Ravindran (2010) to determine the correlation dimension of detrended TEC
- data for some stations in India which lies within the equatorial region, like Nigeria. Ogunsua et
- al., (2014) also used similar methods for some detrended TEC from Nigerian stations.
- 291 The correlation dimension for data taken for the quietest day of October 2011 and the most
- disturbed day of October 2011 from Birnin Kebbi GPS TEC measuring station were represented
- by Fig 8a and Fig 8b respectively. The correlation dimension saturates at $m \ge 4$ for the quietest
- day of the month and at $m \ge 5$ for the most disturbed day. In this illustration the most disturbed
- day of this month fall within the storm period of October 2011. The classification of days into
- 296 quiet and disturbed days in the month of October 2011 enables us to compare the quiet and storm
- 297 periods together while comparing the quiet days with some relatively disturbed days.

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3.1.4 Computation of Tsallis Entropy and Principles of Nonextensive Tsallis Entropy

Entropy measures are very important statistical techniques that can be used to describe the dynamical nature of a system. The Tsallis entropy can be used to describe the dynamical complexity of a system and to also understand the nonlinear dynamics like chaos which may exist in a natural system. The use of entropy measure as a method to describe the state of a physical system has been employed into information theory for decades. The computation of entropy allows us to describe the state of disorderliness in a system, one can generalize this same concept to characterize the amount of information stored in more general probability distributions (Kantz & Shrieber 2003, Balasis et al.,2009). The concept of information theory is basically concerned with these principles. The information theory gives us an important approach to time series analysis. If our time series which is a stream of numbers, is given as a source of information such that this numbers are distributed according to some probability distribution, and transitions between numbers occur with well-defined probabilities. One can deduce same average behaviour of the system at a different point and for the future. The term entropy is used in both physics and information theory to describe the amount of uncertainty or

313 information inherent in an object or system (Kantz and schrieber 2003). The state of an open system is usually associated with a degree of uncertainty that can be quantified by the 314 Boltzmann-Gibbs entropy, a very useful uncertainty measure in statistical mechanics. However 315 316 Boltzmann-Gibbs entropy cannot, describe non-equilibrium physical systems with large 317 variability and multifractal structure such as the solar wind (Burgala et al., 2007, Balasis et al., 2008). One of the crucial properties of the Boltzmann-Gibbs entropy in the context of classical 318 319 thermodynamics is extensivity, namely proportionality with the number of elements of the 320 system. The Boltzmann-Gibbs entropy satisfies this prescription if the subsystems are 321 statistically (quasi-) independent, or typically if the correlations within the system are essentially 322 local. In such cases the system is called extensive. In general however, the situation is not of this 323 type and correlations may be far from negligible at all scales. In such cases, the Boltzmann-324 Gibbs entropy is nonextensive (Balasis et. al., 2008, 2009). These generalizations above were 325 proposed by Tsallis (1988), who was inspired by the probabilistic description of multifractal geometries. Tsallis (1988, 1998) introduced an entropy measure by presenting an entropic 326 327 expression characterized by an index q which leads to a nonextensive statistics,

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$$S_q = k \frac{1}{q-1} \left(1 - \sum_{i=1}^W p_i^q \right)$$
 (8)

Where p_i are the probabilities associated with the microscopic configurations, W is their total number, q is a real number, and k is Boltzmann's constant. The value q is a measure of the nonextensitivity of the system: $q \to 1$ corresponds to the standard extensive Boltzmann-Gibbs statistics. This is the basis of the so called nonextensive statistical mechanics, which generalizes the Boltzmann-Gibbs theory. The entropic index q characterizes the degree of nonadditivity reflected in the following pseudoadditivity rule:

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$$\frac{S_q(A+B)}{k} = \left[\frac{S_q(A)}{k} \right] + \left[\frac{S_q(B)}{k} \right] + (1-q)[S_q(A)/k] \left[\frac{S_q(B)}{k} \right].$$
 (9)

The cases q > 1 and q < 1, correspond to subadditivity (or subextensivity) and superadditivity (or superextensivity), respectively and q = 1 represents additivity (or extensivity). For subsystems that have special theory probability correlations, extensivity is not a valid for Boltzmann-Gibbs entropy, but may occur for S_q with a particular value of the index q. Such systems are sometimes referred to as nonextensive (Boon and Tsallis, 2005, Balasis et al 2008,

2009). The parameter q itself is not a measure of the complexity of the system, but measures the degree of nonextensivity of the system. It is the time variations of the Tsallis entropy for a given $q(S_q)$ that quantify the dynamic changes of the complexity of the system. Lower S_q values characterize the portions of the signal with lower complexity. In this presentation we estimate S_q on the basis of the concept of symbolic dynamics and by using the technique of lumping (Balasis et al. 2008, 2009).

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Considering the fact that Tsallis entropy has been extensively used for magnetospheric studies to obtain interesting result on the dynamical complexity by Balasis et al., (2008; 2009), we find it necessary to consider its application to the study of ionospheric dynamics. It is also necessary to compare the results obtained on the computation of Tsallis entropy to that of Lyapunov exponent. A comparison in the relationship between the values of Lyapunov exponent and Tsallis entropy were carried out to show their relationship as measures of complexity. This is based on the fact that Tsallis entropy has been linked to have a significant degree of response to edge of chaos and chaotic regimes dynamical systems due to its non extensive nature (Baranger et al., 2002; Anastasiadis et al., 2005); and also linked to weak chaos and Vanishing Largest Lyapunov exponent (Kalogeropoulos et al.,2012; 2013). It has been established that Lyapunov exponent varies directly as the Tsallis entropy (complexity) of a system, based on the variation of

Baranger et al., (2002) were able to show that, in the non extensive case Tsallis entropy has been found to vary directly as Kolmogorov-Sinai generated from Lyapunov exponents for Logistic map and dynamical system in the threshold of chaos where $\lambda = 0$, with direct variation when q =1 during chaotic regime. They were able to show that for all cases of positive Lyapunov exponent λ there is an average exponential increase of any small initial distance which can be given as

the entropic index q introduced by Tsallis et al (1988) and the nature of the system's dynamics.

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$$\xi(t) = (x_t - x_t)/(x_0 - x_0)$$
 (10)

For $\xi(t) \equiv exp(\lambda t)$ where x_t and x_t are positions of two 2 initially closed trajectories.

They were able to further relate q to the exponential increase in small distances at the edge of chaos $\lambda = 0$ as

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$$\xi(t) = \left[1 + (1-q)\lambda_q t\right]^{1/(1-q)}$$
 (11)

369 given that

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$$\exp(x) = [1 + (1-q)x]^{1/(1-q)}$$
 (12)

- 371 A similar Tsallis generalization was made for Lyapunov exponent in Coraddu et al., (2005),
- 372 further explaining that the exponential behavior for the chaotic regime is recovered for $q \to 1$:
- $\lim_{q\to 1} exp_q(\lambda_q t) = \exp(\lambda t)$ generalized exponentials shows similar behavior
- 374 However, Anastasiadis et al., (2005) explored different q index values for complex networks for
- 375 $\lambda < 0$ (periodic case) or $\lambda = 0$ (edge of chaos) and $\lambda > 0$ (chaotic regime) where they found
- 376 q=2 to be appropriate for a well distinguish variation in Tsallis entropy between chaos and
- edge of chaos regime, more details can be found in the paper.
- 379 From the established connection between Lyapunov exponent and Tsallis entropy stated above,
- 380 Ogunsua et al., (2014) were able to investigate the similarities in their response to the complex
- 381 dynamics of the ionosphere, and this informs the further use of the two quantities as indices to
- study the day to day variation of ionospheric behaviour in this work.
- 383 The values of these entropy measures were also computed in order to study the dynamical
- 384 complexity of the system under observation (the ionosphere). The day to day values of Tsallis
- entropy were computed for the entire year for different stations. The day to day values of Tsallis
- and for Toro station are shown in fig 9(a and b). The plots
- 387 of the day to day values show the transient variation of the ionosphere and a wavelike yearly
- 388 pattern.

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3.2 Non linearity Test using surrogate data

- 390 The test for non-linearity using the method of surrogate data according to Kantz and Schreber
- 391 (2003) has proven to be a good test for non-linearity in time series describing a system. It has
- 392 been accepted that the method of surrogate data test could be a successful tool for the
- 393 identification of nonlinear deterministic structure in an experimental data (Pavlos et al., 1999).
- 394 This method involves creating a test of significance of difference between linearly developed

surrogate and original nonlinear time series to be tested. The test is done by carrying out the computation of the same quantity on both surrogates and the original time series and then checking for the significance of difference between the results obtained from the surrogates with the original data. Theiler et al (1992) suggested the creation of surrogate data by using Monte Carlo techniques for accurate results. According to this method, typical characteristic of data under study are compared with those of stochastic signals (surrogates), which have the same auto-correlation function and the power spectrum of the original time series. It can be safely concluded from the test of significance carried out on the surrogate and the original data that, a stationary linear Gaussian Stochastic model cannot describe the process under study provided that the behaviour of the original data and the surrogate data are significantly different.

In this work 10 surrogate data were generated from the original data set. The geometrical and dynamical characteristics of the original data were then compared to that of the surrogates using the statistical method of significance of difference which can be defined as

$$S = \frac{\alpha_{Surr} - \alpha_{Original}}{\sigma} \tag{13}$$

Where α_{Surr} is the mean value of the computed quantity for the surrogate data and $\alpha_{Original}$ is the same quantity computed for the original TEC data, σ is the standard deviation of the same quantity computed for the surrogate data. The significance of difference considered for the null hypothesis to be rejected here is greater than 2, which enables us to be able to reject the null hypothesis that the original TEC data describing the ionospheric system can be modeled using a Gaussian linear stochastic model with confidence greater than 95%.

The surrogate data test for all stations used in this study show that the Lyapunov exponent of the surrogate data for the selected days in October are shown in the Table below The results show that the surrogate data test for Lyapunov exponent show a significance of difference greater than 2 for all the selected days for all the stations. Similar results were obtained for Mutual Information, Fraction of False Nearest Neighbours and Correlation Dimension. This result gives us the confidence to reject the null hypothesis that the data used cannot be modeled using a linear Gaussian stochastic model, which shows that the system is a nonlinear system with

some level of determinism. Fig. 10 shows the plots comparing the mutual information plotted against time delay for the original detrended data blue with the mutual information for the surrogate data for TEC data measured at Lagos for the quietest day of March 2011, while Fig. 11 is comparing fraction of false nearest neighbours for the same set of data. Tables 2a shows the values of Lyapunov exponents for both original detrended and its surrogate data for TEC measured in Lagos during the quietest days and Table 2b shows the values of Lyapunov exponents for both original detrended and its surrogate data for TEC measured in Lagos during the most disturbed days of October 2011.

3.3 Trend filtering using the moving average approach for the daily Values

The trend of a fluctuating time series can be made clearer to reveal the general pattern of that time series, and to make the fluctuating pattern of the daily variation of the chaoticity and dynamical complexity measures clearer in the work, the moving average method has been employed. The method of moving average filtering has found its applications geophysics (e.g. Bloomfeld 1992; Bloomfield and Nychka 1992; Baillie and Chung 2002), and in other areas like financial time series analysis, microeconomics, biological sciences and medical sciences. The various fields mentioned require different trend filtering method depending on the structure of the time series to be analyzed. Different filtering processes that can be used to reveal the trend includes the moving average filters, exponential filters, band-pass filtering, median filtering etc.

Suppose we have a time series z[t] such that $t = 1,2,3 \dots n$, where 'n' could assume any value. If z[t] consists of a consistent varying trend component that appears over a longer period of time t given as u[t] and a more rapidly varying component v[t]. The goal of trend filtering in any research is to estimate either of the two components (Kim et al., 2009). The purpose of trend filtering in this work is to further reveal the general slow varying trend that appears to be obvious in the daily variation of the values of the chaoticity and dynamical complexity of the ionosphere, which might appear to be obviously varying with the yearly solar activity (a quantity with slow varying trend). To make u[t] which represents the general slow varying trend smoother and in the process reduce v[t] we apply the moving average filter.

- 452 If we assume z[t] to be our time series representing the daily variation of the values of the
- 453 chaoticity and dynamical complexity of the ionosphere, then our smoothing with weighting
- 454 vector/filter w_i will create the new sequence u_i as

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$$u[t] = z[t] * w[n] = \frac{1}{2k+1} \sum_{i=-k}^{k} x[n-1].$$
 (14)

- 456 In this work the Savitzy-Golay method of smoothing proposed by Savitzky and Goley (1967),
- 457 which is a generalized form of moving average was applied to the trend smoothing of the daily
- 458 variation of the chaoticity and dynamical complexity of the ionosphere. In this case it performs a
- least square fit to a small set of L(=2k+1) consecutive data to a polynomial and then takes
- 460 midpoint of the polynomial curve as output. The smoothed time series in this work will now be
- 461 given as

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$$u[t] = z[t] * \omega[n] = \frac{\sum_{i=-k}^{k} A_i * x[n-i]}{\sum_{i=-k}^{k} A_i}$$
 (15)

- where, $\omega[n] = \frac{A_n}{\sum_{i=-k}^k A_i}$, $-k \le n \le k$ such that A_i controls the order of polynomial. A similar
- method was described in Reddy et al., (2010).
- 465 The smoothed daily variation and the original data and the plot of the smoothed variation only,
- 466 for the Lyapunov exponents of the detrended TEC measured at the Enugu and Toro are shown in
- 467 fig 12(a and b). The smoothed day to day variation for Tsallis entropy for the detrended TEC
- measured at Enugu and Toro stations respectively are shown in fig 13(a and b).

469 4.0 DISCUSSION

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- 470 The results presented in the work reveals the dynamical characteristics of the ionosphere. These
- 471 characteristics are being discussed in this section, considering the time series treatment and phase
- 472 space reconstruction; the study of chaos using chaotic quantifiers and the use and comparison of
- 473 dynamical complexity measures in terms of their response to the variations on ionospheric
- 474 dynamics. Also being discussed, is the implication of the nonlinearity test using the surrogate
- data and the comparison of the two quantifiers and their viability as indices for the continuous
- 476 study and characterization of the ionosphere

The time series analysis shows the appearance of some degree of nonlinearity in the internal dynamics of the ionosphere. The time series plot in Fig. 1 shows the rise in TEC to peak at the sunlit hours of the day, however it can be seen that the rising to the peak exhibited by the ionosphere, which is the dominant dynamics during the day, make it impossible to clearly see the internal dynamics of the system from the TEC time series plot. It can be seen that the TEC time series curve is not a smooth curve with tiny variations, which probably describes a part of the internal dynamics. These visible tiny variations around the edges of the time series plot can be regarded as rate of change of TEC which is a phenomenon that can describe the influence of scintillations in the ionosphere these variations are however more obvious during the night time between 1100th and 1440th minutes of the day (that is, between about 1800 and 2400 hours of the day). It should be noted here that scintillations has been described as a night time phenomena associated with spread-F, and it occurs around pre-midnight and post-midnight periods (Vyas and Chandra 1994; Vyas and Dayanandan 2011; Mukherjee et al., 2012; Bhattacharyya and Pandit 2014). The detrended data shows the internal dynamics of the system more clearly, with a pattern similar to the values around night period mentioned earlier. The post-sunset values (especially at night time) in Fig.1 show a pattern similar pattern with the detrended TEC plot in Fig 2. It has been established that TEC does not decrease totally throughout the night as expected normally through simple theory that TEC builds up during the day, but it shows some anomalous enhancements and variations and this can occur under a wide range of geophysical conditions (Balan and Rao, 1987; Balan et al., 1991; Unnikrishnan and Ravindran, 2010). The delay representation of the phase space reconstruction shows a trajectory that is clustered around its origin, for all the stations, which can be seen as an indication of the possible presence of chaos. The degree of closeness of these trajectories however varies for different days from one station to another, resulting from varying degrees of variations in stochasticity and determinism. The varying degrees of variations in stochasticity and determinism can be attributed to the daily variations and local time variations of photoionization, recombination, influx of solar wind and other factors that may influence the daily variations of TEC (Unnikrishnan 2010).

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The positive values of Lyapunov exponent indicate the presence of chaos (Wolf et al., 1985; Rosenstein et al., 1993; Hegger et al., 1999; Kantz and Schreiber, 2003). The presence of chaos was revealed by the positive Lyapunov exponent computed from all stations and this as a result

of the fact that ionosphere is a dynamic system controlled by many parameters including acoustic motions of the atmosphere electromagnetic emission and variations in the geomagnetic field. Because of its extreme sensitivity to solar activity, the ionosphere is a very sensitive monitor of solar events. The ionospheric structure and peak densities in the ionosphere vary greatly with time (sunspot cycle, seasonally and diurnally), with geographical location (polar, auroral zones, mild-latitudes, and equatorial regions), and with certain solar-related ionospheric disturbances. During and following a geomagnetic storm, the ionospheric changes around the globe, as observed from ground site can appear chaotic (Fuller-Rowell et al., 1994; Cosolini and Chang, 2001; Unnikrishnan and Ravindran, 2010). The recorded presence of chaos as indicated by the positive values of Lyapunov exponent was found in all the computations, for all the TEC values obtained for the selected days from all the measuring stations used in this work. This can be expected as it agrees with results from previous works that show that there is a reasonable presence of chaos in the ionosphere, even in the midst of the influence of stochastic drivers like solar wind (Bhattacharyya, 1990; Wernik and Yeh, 1994; Kumar et al., 2004; Unnikrishnan et al., 2006a,b; Unnikrishnan, 2010). However the values of Lyapunov exponents vary from day to day due to variations in ionospheric processes for different days on the same latitude as seen in Fig. 7(a and b) with Fig. 12(a and b) showing the day to day variation (upper panel) and the smoothed curve of the day to day variation (lower panel) for the entire year. There are also latitudinal variations due to spatial variations in the various ionospheric processes taking place simultaneously. The ionosphere is said to have a complex structure due to these varying ionospheric processes. The higher values of Lyapunov exponent during months of low solar activity (the solstices) is an

The higher values of Lyapunov exponent during months of low solar activity (the solstices) is an evidence that that the rate of exponential growth in infinitesimal perturbations in the ionosphere leading to chaotic dynamics might be of higher degree during most of the days of those months compared to days of the months with high solar activities showing lower values of Lyapunov

exponents (Unnikrishnan 2010, Unnikrishnan and Ravindran, 2010).

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The results of the correlation dimension values computed are within the range of 2.8 to 3.5 with the lower values occurring mostly during the storm periods. The lower dimension during the storm periods compared to the quiet days may be due to the effect of a stochastic drivers like strong solar wind and solar flares, that occurs during geomagnetic storms on the internal

dynamics of the ionosphere, this could have been as a result of the fact that the internal dynamics must have been suppressed by the external influence. The restructuring of the internal dynamics of the ionosphere might be responsible for low dimension chaos during storm and also the lower values of other measures like the Lyapunov exponents. The relatively disturbed day however might have a higher dimension so long as it is not a storm period, and sometimes a relatively disturbed day of the month might be a day with storm and in this case there is usually a lower value of chaoticity and sometimes lower values of correlation dimension as well. The lower value of chaoticity and dimension in ionosphere during storms indicates a phase transition from higher values during the quiet periods to lower values during storm periods which may be due to the modification of the ionosphere by the influx of high intensity solar wind during the storm period (Unnikrishnan et al., 2006a, b; Unnikrishnan 2010; Unnikrishnan and Ravindran, 2010).

The surrogate data test shows significance of difference greater than 2 for all the computed measures which enables rejection of the null hypothesis that the ionospheric system can be represented with a linear model for all the data used from the stations. However it was discovered that the lower significance of difference corresponds to the lower values of Lyapunov exponents during storm and extremely disturbed periods (see tables 2 and 3). This may be due the rise in stochasticity during the storm period as a result of drop in values of computed quantities like Lyapunov exponents. Our ability to reject the Null hypothesis for all stations however shows the presence of determinism and confirmconfirms that the underlying dynamics of the ionosphere is mostly non-linear. This further validates the presence of chaos since the surrogate data test for non-linearity show that out detrended TEC is not a Gaussian (linear) stochastic signal (Unnikrishnan 2010).

The Tsallis entropy was able to show the deterministic behavior of the ionosphere considering its response during storm periods compared to other relatively quiet periods as the rapid drop in values of Tsallis entropy during storm show that there is a transition from higher complexity during quiet period to lower complexity during storms, this response in the values of Tsallis entropy is similar to the response of Lyapunov exponent values during storm. This reaction to storm shown by the values of Tsallis entropy computed for TEC was also described by the

reaction of Tsallis entropy computed for Dst during storm periods (Balasis et al., 2008, 2009). A closer observation of the day-to-day variability within a month shows that the values were much lower for storm periods compared to the nearest relative quiet period. For example, the storm that occurred on the 25th of October resulted in lower values of Lyapunov exponent and Tsallis entropy compared to relatively quiet days close to it. The reaction to storm may be due to the influence of stochastic driver like strong solar wind flowing into the system as a result of solar flare or CMEs that produces the geomagnetic storms. Although there is always an influence of corpuscular radiation in form of solar wind flowing from the sun into the ionosphere, the influence is usually low for days without storm compared to days with geomagnetic storms as a result of solar flares, CMEs etc (Unnikrishnan et al., 2006a,b; Unnikrishnan, 2010, Ogunsua et al 2014).

The presence of chaos and high variations in the dynamical complexity, even at quiet periods in the ionosphere may be due to the internal dynamics and inherent irregularities of the ionosphere which exhibit non-linear properties. However, this inherent dynamics may be complicated by external factors like Geomagnetic storms. This may be the main reason for the drop in the values of Lyapunov exponent and Tsallis entropy during storms. According to Unnikrishnan et al., (2006a,b), geomagnetic storms are extreme forms of space weather, during which external driving forces, mainly due to solar wind, subsequent plasmasphere -ionosphere coupling, and related disturbed electric field and wind patterns will develop. This in turn creates many active degrees of freedom with various levels of coupling among them, which alters and modifies the quiet time states of ionosphere, during a storm period. This new situation developed by a storm, may modify the stability/instability conditions of ionosphere, due to the superposition of various active degrees of freedom.

The observation from the day-to-day variability of Lyapunov exponent and Tsallis entropy also show irregular pattern for all stations. These irregular variations might be due to the same factors mentioned before (i.e internal irregularities due to so many factors described and also due to variation in the influx of the external stochastic drivers). The day-to-day variability for the entire year shows a "wavelike" pattern with the values dropping to lower values during the equinox months especially during March-April equinox. The wavelike pattern has been found to be

similar for different stations as seen in Figs. 7 & 12 and Figs. 9 &13 for Lyapunov exponents and Tsallis entropy respectively. Figs.9 and 13 show the smoothed curves for Lyapunov exponent and Tsallis entropy respectively, with the drop in values at equinoxes showing more clearly. The phase transition in chaoticity and dynamical complexity is also responsible for the wavelike variations, with values of Lyapunov exponent and Tsallis entropy dropping during the equinoxial months, and this may be due to the influence of the daily influx of the solar wind having higher values during equinoxes due to the proximity of the Earth to the sun during this period compared to the solstice months.

 The wavelike pattern observed has been described to be as a result of self organized critical (SOC) phenomenon, a phenomenon which has been found to exist in the magnetosphere and the same could exist in the ionosphere, since the magnetosphere couples the ionosphere tightly to the solar wind (Lui, 2002). This was first suggested by (chang et al., 1992,1998,1999; Consolini et al., 1996 Chapman et al., 1998; Freeman and Watkins 2002 and; Koselov and Koselova, 2001. Uritsky et al., (2003) and Chang et al., (1992) pointed out that the existence of SOC in plasma sheet in the tail of the magnetosphere and the entire magnetospheric system is described by the manner in which the magnetospheric dynamics exhibits a number of scale free-statistical relation. This has been verified in many ways from the observations made on local and global characteristics of geomagnetic perturbations as seen in Freeman and Watkins 2002.

Perrault and Akasofu (1978) argued that the scale free component of the magnetosphere can be possibly as a result of external perturbations like solar wind. Therefore we can describe the SOC as a specific slowly driven many-body system characterized by an intermittent scale-free response to the external perturbations and global instability, which implies that the system can adjust to rate of changes, as in the case of magnetospheric system without losing its signatures of critical dynamics (Bak et al., 1987; Chang et al., 1992, Uritsky et al., 2003).

Similar effects can occur in the ionosphere since the ionosphere is coupled to the ionosphere as mentioned earlier. Therefore ionosphere experiences the effects of solar wind as it impacts the magnetosphere. The lower values of chaoticity at the equinoxes have been suggested to be as a result of the fact that the internal dynamics of the system adjusts itself to the perturbation from the influx of the solar wind which maximizes at the equinoxes. The suppression of the internal dynamics of the ionosphere as a result of its modification by external stochastic drivers like the solar wind has been described by (Unnikrishnan et al., 2006; 2010; Ogunsua et al., 2014). The resulting wavelike pattern might be more obvious at the equatorial region due to the proximity of the region to the sun which lies directly above the equator during the equinoxes.

Although there is a scale-free response as mentioned before, the suppressed internal dynamics does not change its signatures as the ionospheric system retains it chaotic dynamics but only at a lower level. This is described by the drop in the values of the two parameters describing the chaoticity and dynamical complexity of the ionosphere, that is, the Lyapunov exponent and Tsallis entropy.

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The variation along the latitude also shows the inconsistence and complexity of the ionospheric processes. This is the reason why for the same day of the month the values of Lyapunov exponent vary from one station to another. Lyapunov exponent however, appears to respond better to changes in solar activities compared to Tsallis entropy with more distinct results. This may be due to the fact that Tsallis entropy being not only a measure of complexity, but also a measure of disorderliness in a system might not be as perfect in describing chaos as Lyapunov exponent. Kalogeropoulos (2009) and Baranger et.al (2002) observed that Tsallis entropy has a relationship that is not totally linear in all cases at different level of chaos with Lyapunov exponent as a measure of chaos.

There are also many variations in the internal dynamics of the ionosphere that could lead to changes in chaotic behavior. The variations of Lyapunov exponents during quiet days might be as a result of different variations in the intrinsic dynamics of the ionosphere. Difference in variation pattern at different stations for the same quiet day might also be due to the same reason. It can be affirmed that the ionosphere is a complex system that varies with a short latitudinal or longitudinal interval such that even stations with one or two degrees of latitudinal differences might record different values on the same day for both quiet and disturbed periods and that the same might also occur for storm periods. This is illustrated by the different pattern of variation of TEC recorded from different stations within such a close range as used in this study.

These Latitudinal variation in the values of Lyapunov exponents and Tsallis entropy can be further described by the behavior of the TEC because there can be a more sporadic rate of change in TEC as seen in the time series plots as a result of irregularities in the internal dynamics of the ionosphere, which might be as a result of plasma bubbles. Irregularities develop in the evening hours at F region altitudes of magnetic equator, in the form of depletions, frequently referred to as bubbles. The edges of these depletions are very sharp resulting in large time rate of TEC in the equatorial ionosphere, even during magnetically quiet conditions. The large gradient of the equatorial ionization persists in the local post-sunset hours till about 2100 h LT. (DasGupta et al., 2007; Unnikrishnan and Ravindran, 2010). The TEC data for one station might experience an extremely sharp rate of change in TEC that may be due to some plasma bubbles in that region while the TEC from the other station stays normal. These variations in the various

internal dynamics like plasma bubbles leading to scintillation can cause variations in the dynamical response of the TEC. Hence, the irregular variation in the values of the Lyapunov exponent and Tsallis entropy even in quiet periods for two relatively close stations may be due to these irregularities. This might also be responsible for the quiet days in the same station having lower values of Lyapunov exponent compared to higher values recorded for disturbed days without the external influence of storms.

The variations of these chaos and dynamical complexity parameters might also be as a result of the anomalous TEC enhancements that might occur at nights (Balan and Rao (1987); Balan et al., 1991). These effects can also be seen more clearly in the Tsallis entropy values for the five period window for quiet day of January, 2011, because the night time value is higher and it also show a much higher series of fluctuations during this period compared to other periods. As mentioned in Unnikrishnan and Ravindran (2010), the irregular changes in the dynamical characteristics of TEC from the results of Lyapunov exponent and Tsallis entropy also may be due to the collisional Raleigh-Taylor instability which may give rise to a few large irregularities in L band measurements (Rama Rao et al., 2006; Sripathi et al., 2008) all these can be seen as internal factors responsible for variations in the dynamical response of TEC as recorded from the values of the Lyapunov exponents and Tsallis entropy completed for days without storm which might be quiet or disturbed according to classification and also could account for higher values of these qualifiers during disturbed days compared quiet days. During storms however, the values were much lower.

The relationship between Lyapunov exponent and Tsallis entropy can further be seen from this work as the two quantifies exhibit similarities in their response to the dynamical behavior of the ionosphere with phase transition at the same periods of time for all stations. A further investigation of this relationship shows that all the daily values of Tsallis entropy correlates positively with the values of Lyapunov exponent at values between 0.78 and 0.83.

The ability of these quantifiers to clearly reveal the ionospheric dynamical response to solar activities and changes in its internal dynamics due to other factors is a valid proof of the

authenticity of the use of these chaotic and dynamical measures, as indices for ionospheric studies.

5.0 Conclusion

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The chaotic behaviour and dynamical complexity of low latitude ionospheric behaviour over some parts of Nigeria was investigated using TEC time series measured at five different stations namely Birnin Kebbi (geographic coordinates 12°32′N, 4°12′E; dip latitude 0.62°N), Torro (geographic coordinates $10^{\circ}03'N$, $9^{\circ}04'E$; dip latitude $-0.82^{\circ}N$), Enugu (geographic coordinates 6°26'N, 7°30'E; dip latitude -3.21°N), Lagos (geographic coordinates 6°27'N, $3^{\circ}23'E$; dip latitude $-3.07^{\circ}N$) and Yola (geographic coordinates $9^{\circ}12'N$, $12^{\circ}30^{\circ}E$; dip latitude -1.39°N) within the low latitude region. The detrended TEC time series data obtained from the GPS data measurement were studied for chaoticity using phase space reconstruction techniques, computation of Lyapunov exponents and correlation dimension. Tsallis entropy was used for the study of dynamical complexity of the ionospheric system described by the TEC data. The detrended TEC time series were subjected to further analysis for phase space reconstruction from which the choice of time delay of 30 was obtained and an embedding dimension of 5 were considered in this study. The evidence of the presence of chaos in all the time series data was obtained for all the data used, as indicated by the positive Lyapunov exponent. The results of Tsallis entropy show the variations in the dynamical complexity of the ionosphere, which may be due to geomagnetic storms and other phenomena like changes in the internal irregularities of the ionosphere. The response of the Tsallis entropy to various changes in the ionosphere also shows the deterministic nature of the system. The results of the Tsallis entropy show a lot of similarities with that of the Lyapunov exponents, with both results showing a phase transition from higher values in the solstices to lower values during the equinoxial months. The values of Lyapunov exponent were found lower for the days of the months in which storm was recorded relative to the nearest relatively quiet days which agree with previous works by other investigators. A similar pattern of results was obtained for the computed values of Tsallis entropy. The random variations in the values of chaoticity in the detrended TEC describing the internal dynamics of the ionosphere as seen in the result obtained from both Lyapunov exponent and Tsallis entropy

depicts the ionosphere as a system with a continuously changing internal dynamics, which shows

that the ionosphere is not totally deterministic but also has some elements of stochasticity influencing its dynamical behaviour.

The phase transition in the systems of the ionosphere resulting in the lower values of the chaoticity and dynamical complexity quantifiers during the geomagnetic storms and the equinoxial months is the evidence that the ionosphere can be greatly modified by stochastic drivers like solar wind and other incoming particle systems. It can also be seen that the results of Tsallis entropy follow the same pattern with Lyapunov exponent, which shows show that both can be use simultaneously and comparatively as measures of chaos and dynamical complexity as the correlation of all the values obtained for both quantities give values between 0.78 and 0.81.

Although the knowledge of being able to characterize the ionospheric behaviour using the two m ajor quantifiers shows their ability to measure level of determinism when used together, the relat ionship between these two quantifiers calls for more research, in the use of these qualifiers, to en able proper description and characterization of the state of ionosphere. The response of both Tsal lis entropy and Lyapunov exponents to changes in the ionosphere shows that the two quantifiers can be used as indices to describe the processes/dynamics of the ionosphere.

Even though we cannot conclude totally until further investigations have been carried out on various properties of the ionosphere describing its dynamics. It can be safely established that this study has created roadmap for the use of the chaoticity and dynamical complexity measures as indices to describe the process/dynamics of the ionosphere.

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766	Acknowledgement
767	The GPS data used for this research were obtained from the public archives of the Office of the
768	Surveyor General of the Federation (OSGoF) of the Federal Government of Nigeria, which is the
769	mapping agency of Nigeria.
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Table 1: Coordinates of the GPS stations

Station Name	Geographic Coordinates		Dip latitude (° <i>N</i>)
	Long (°E)	$Lat(^{\circ}N)$	
Birnin Kebbi	4° 12′E	12° 32′N	0.62°N
Torro	9° 04′E	10° 03′N	−0.82°N
Yola	12° 30′E	9° 12′N	-1.39°N

Lagos	3° 23′E	6° 27′N	−3.07° <i>N</i>	969
Enugu	7°30′E	6° 26′N	−3.21° <i>N</i>	970

Table 2a: Results of Surrogate data test for Lyapunov exponent for TEC data for the quietest days of October 2011 at Birnin Kebbi station.

Original Data	Surrogate data
0.1165	0.3921 ± 0.0420
0.0931	0.2029 ± 0.0756
0.1041	0.3860 ± 0.0741
0.0498	0.2891 ± 0.0598
0.1420.	0.3621 ± 0.0504

Table 2b: Results of Surrogate data test for Lyapunov exponent for TEC data for the most disturbed days of October 2011 at Birnin Kebbi station.

Surrogate data	980
0.3039 ± 0.0541	981
0.3156 ± 0.0428	
0.2527 ± 0.0296	982
0.3662 ± 0.0468	983
0.3100 ± 0.0416	004
	0.3039 ± 0.0541 0.3156 ± 0.0428 0.2527 ± 0.0296 0.3662 ± 0.0468

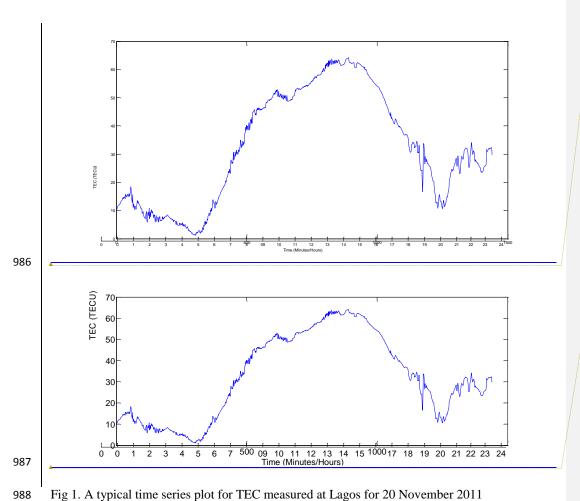


Fig 1. A typical time series plot for TEC measured at Lagos for 20 November 2011

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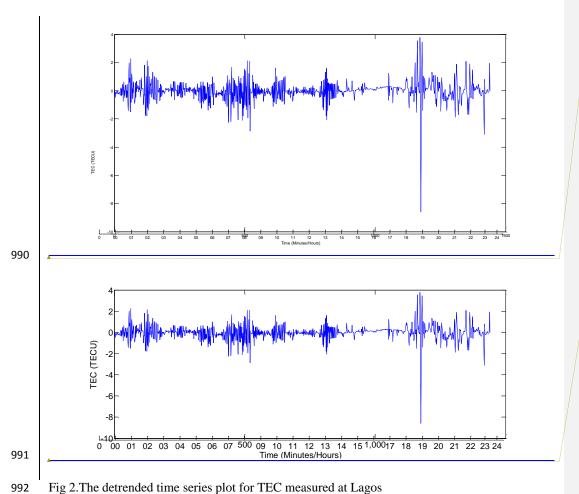


Fig 2.The detrended time series plot for TEC measured at Lagos

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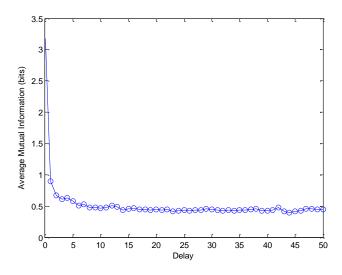


Fig. 3 Average mutual information against time Delay for TEC measured at Yola

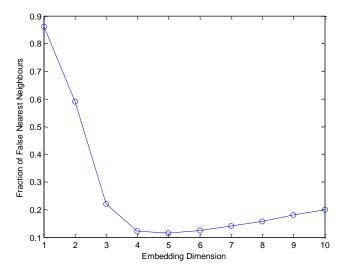


Fig. 4 Fraction of false nearest neighbours against embedding dimension for TEC measured at yola

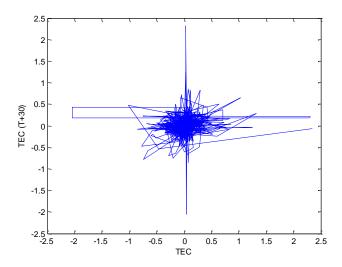


Fig.5 The Delay representation of the phase space reconstruction of the detrended TEC

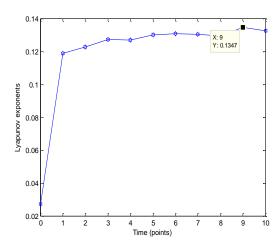
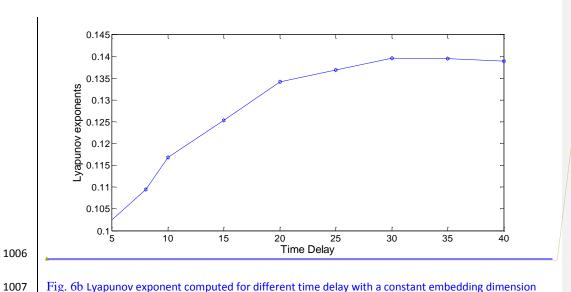
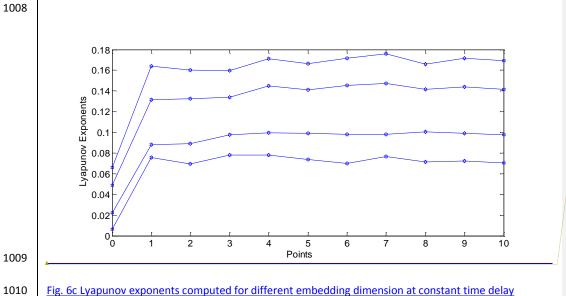


Fig. 6a Lyapunov Exponent computed and its evolution, computed as the state space trajectory scanned with tau=30, m=5 for detrended time series measured at Yola with Largest Lyapunov Exponent equal to 0.1347



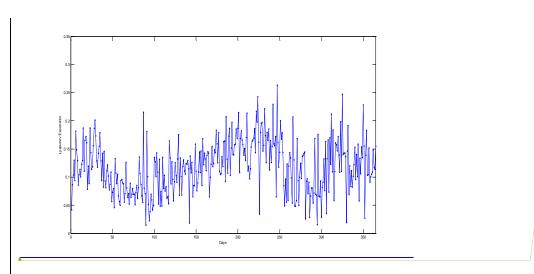
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Fig. 6b Lyapunov exponent computed for different time delay with a constant embedding dimension



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Fig. 6c Lyapunov exponents computed for different embedding dimension at constant time delay



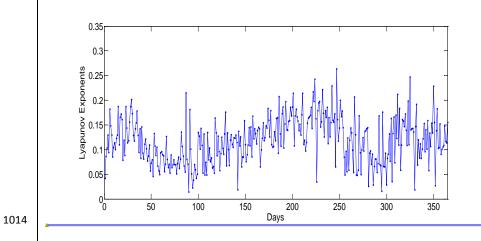
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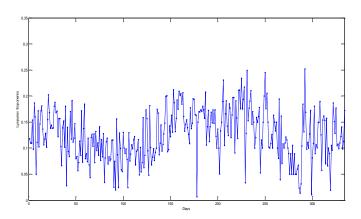
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Fig. 7a The transient variations of Lyapunov exponents for 365 days of 2011 for detrended TEC measured at Enugu



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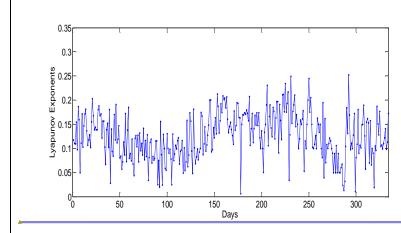


Fig. 7b The transient variations of Lyapunov exponents for 334 days (Jan1 –Nov30) of 2011 for detrended TEC measured at Toro

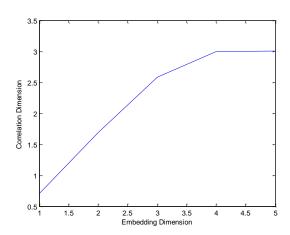


Fig. 8a The correlation dimension of the detrended TEC for the quietest day of October at Birnin Kebbi which saturates at $m \ge 4$ and $\tau = 39$

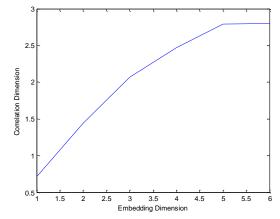


Fig. 8b The correlation dimension of the detrended for the most disturbed day of October at Birnin Kebbi which saturates at $m \ge 5$ and $\tau = 34$

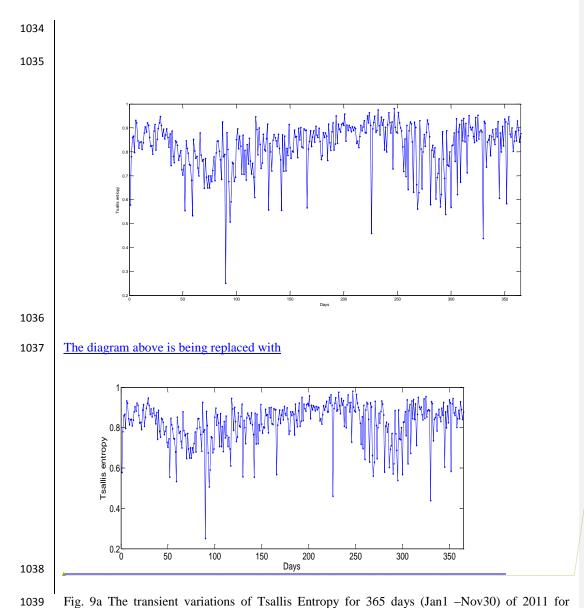
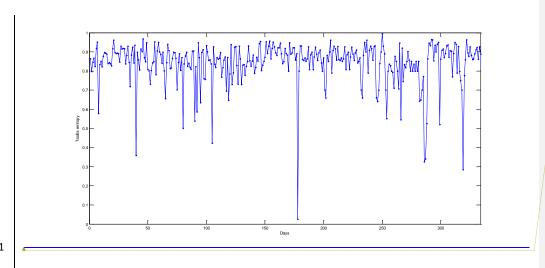


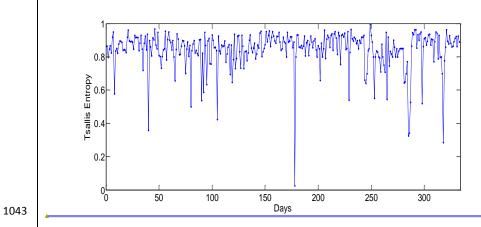
Fig. 9a The transient variations of Tsallis Entropy for 365 days (Jan1 –Nov30) of 2011 for detrended TEC measured at Enugu



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Fig. 9b The transient variations of Tsallis Entropy for 334 days (Jan1 –Nov30) of 2011 for detrended TEC measured at Toro

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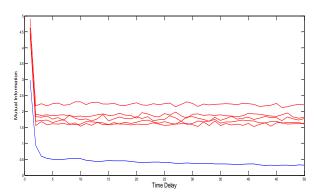


Fig 10 Mutual information plotted against time delay for the original detrended data in (blue curve) with the mutual information for the surrogate data (red curve) for TEC data measured at Lagos for the quietest day of march 2011

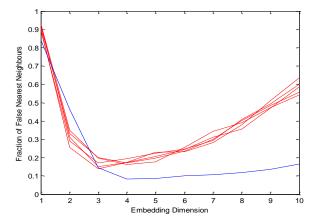
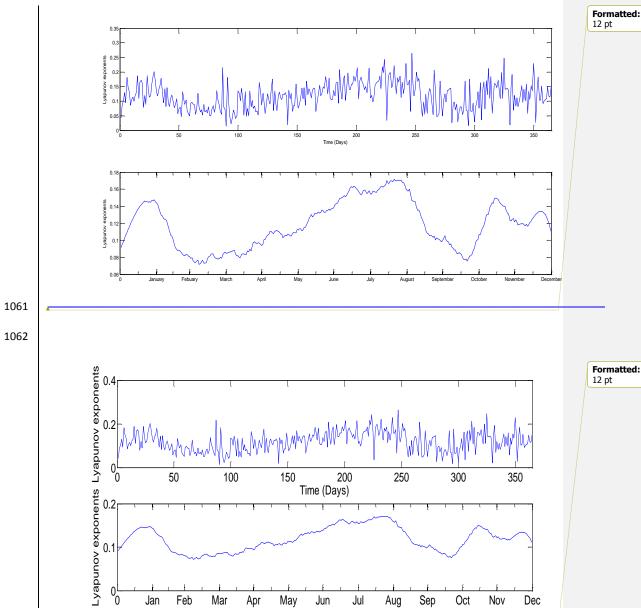


Fig 11 Fraction of false nearest neighbours plotted against time embedding dimension for the original detrended data in (blue curve) with the mutual information for the surrogate data (red curve) for TEC data measured at Lagos for the quietest day of march 2011





200 Time (Days)

May Jun Time(Months)

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Aug

Sep

Jul

300

Oct

Nov

150

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0 0

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Jan

Feb

100

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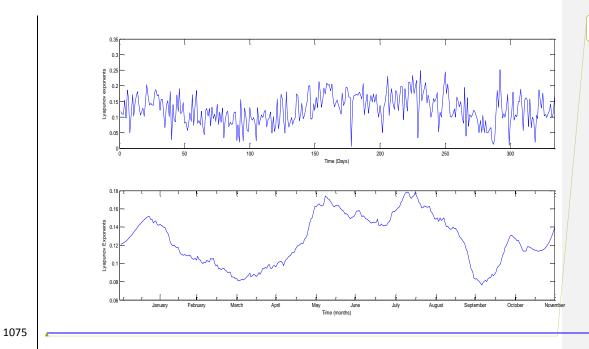
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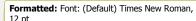
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Fig. 12a Daily variation of Lyapunov exponents for TEC measured at the Enugu station for the year 2011 showing the Original data (Upper Panel) and the smoothed Plot of daily variation of Lyapunov exponents for TEC measured at the Enugu station for the year 2011 (Lower panel)





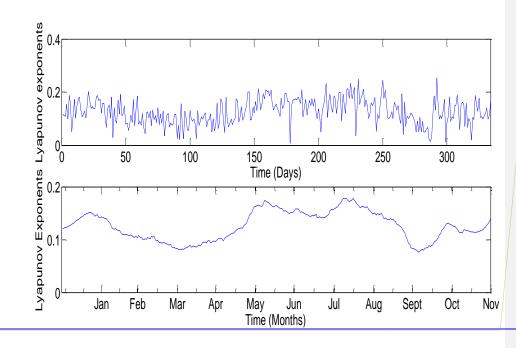


Fig 12b Daily variation of Lyapunov exponents for TEC measured at the Toro station for the year 2011 showing the Original data (Upper Panel) and the smoothed Plot of daily variation of Lyapunov exponents for TEC measured at the Toro station for the year 2011 (Lower panel)

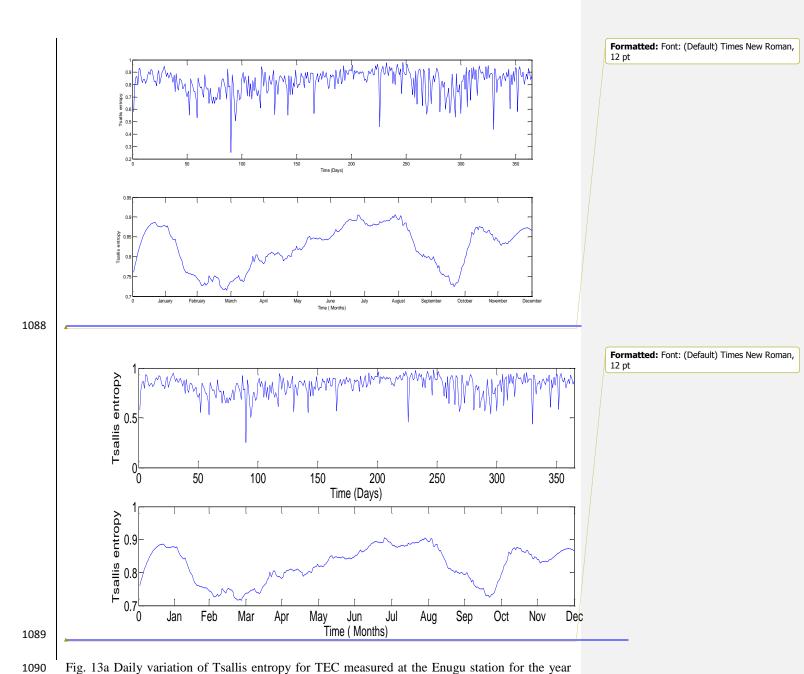
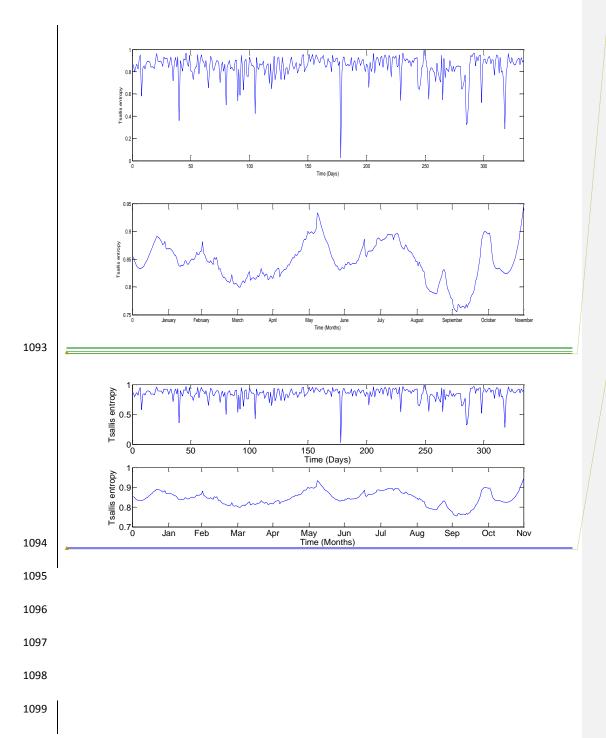


Fig. 13a Daily variation of Tsallis entropy for TEC measured at the Enugu station for the year 2011 showing the Original data (Upper Panel) and the smoothed Plot of daily variation of Lyapunov exponents for TEC measured at the Enugu station for the year 2011 (Lower panel)



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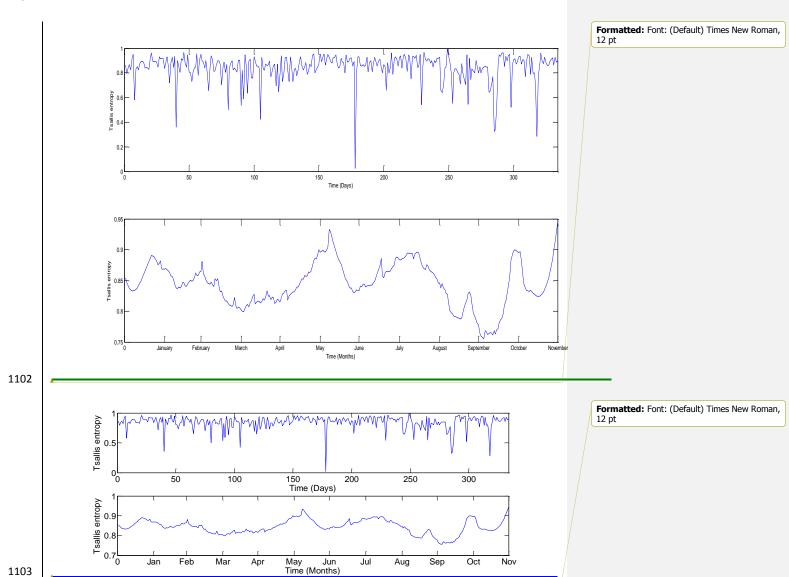


Fig. 13b Daily variation of Tsallis entropy for TEC measured at the Toro station for the year 2011 showing the Original data (Upper Panel) and the smoothed Plot of daily variation of Lyapunov exponents for TEC measured at the Enugu station for the year 2011 (Lower panel)

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- 1175 COMMENT, RESPONSE AND EXPECTED CHANGES TO MANUSCRIPT
- 1176 BASED ON THE FIRST REFEREE REPORT (NUMBER 1) ON THE
- 1177 **PAPER TITILED:**
- 1178 THE TRANSIENT VARIATION OF THE COMPLEXES OF THE LOW
- 1179 LATITUDE IONOSPHERE WITHIN THE EQUATORIAL IONIZATION
- 1180 ANOMALY REGION OF NIGERIA".

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1. REVIEW COMMENTS BY K. UNNIKRISHNAN

This study was conducted using Total Electron Content (TEC) time series for 2011, measured from 5 GPS receiver stations in Nigeria by employing chaotic\non linear analysis. The detrended TEC time series were reconstructed and the values of chaotic quantifiers namely, Lyapunov exponents LE, correlation dimension, and Tsallis entropy were estimated to reveal dynamical complexity of the system. Authors aim to link the chaotic quantifiers and ionospheric behaviour over Nigeria using nonlinear techniques, which is further verified by surrogate data test, and they produced some interesting results. The paper is well written and worth publishing. I strongly recommend for the publication of this article after minor revision based on the points listed below: 1) The convergence of the computed Lyapunov exponents should be discussed by showing whether they are stable with the change of the embedding dimension, and the time delay. This aspect is very important since the computations of LE depend strongly on the ability to track the dynamical trajectories in the embedded space. For this, in the revised version, authors may present plots for LE versus time delay, by keeping embedding dimension a constant, and also between LE and embedding dimensions, at constant time delay. 2) The wavelike pattern exhibited by LE and Tsallis entropy with the drop in values at equinoxes (Figures 7, 9, 12 and 13) possibly due to self-organized critical phenomenon of the system, is an interesting observation. It would be better if authors could present some more clarifications to link the selforganized critical phenomenon of equatorial ionosphere and the observed wavelike pattern of LE and Tsallis entropy. 3) Axis title and labels for Figures 1,2,7,9,10,11, and 13 are very small in size. Please redraw them with more clarity.

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2. AUTHORS COMMENTS

1206 The authors will hereby appreciate the honest observations on our paper and attempts have been made to include the revision points in the main published paper. 1207 1208 3. AUTHORS CHANGES TO THE MANUSCRIPT 1209 1210 1211 The first point based on the Stability of Lyapunov exponent at different time delay and at different embedding dimension has been included. Please see Fig 1-2. 1212 II. The second point based on clarification to be made on the Reflection of Self 1213 Organized Criticality (SOC) in the ionospheric dynamics will be considered with the 1214 inclusion of a new subtopic on SOC in the discussions of the published paper which 1215 is explained in the write up below with references. 1216 1217 1218 1219 Reflection of Self Organized Criticality (SOC) in the ionospheric dynamics 1220 The wavelike pattern observed has been described to be as a result of self organized critical 1221 (SOC) phenomenon, a phenomenon which has been found to exist in the magnetosphere and the 1222 same could exist in the ionosphere, since the magnetosphere couples the ionosphere tightly to the 1223 solar wind (Lui, 2002). This was first suggested by (chang et al., 1992, 1998, 1999; Consolini et al., 1996 Chapman et al., 1998; Freeman and Watkins 2002 and; Koselov and Koselova, 2001. 1224 Uritsky et al., (2003) and Chang et al., (1992) pointed out that the existence of SOC in plasma 1225 1226 sheet in the tail of the magnetosphere and the entire magnetospheric system is described by the 1227 manner in which the magnetospheric dynamics exhibits a number of scale free-statistical 1228 relation. This has been verified in many ways from the observations made on local and global 1229 characteristics of geomagnetic perturbations as seen in Freeman and Watkins 2002. 1230 Perrault and Akasofu (1978) argued that the scale free component of the magnetosphere can be 1231 possibly as a result of external perturbations like solar wind. Therefore we can describe the SOC as a specific slowly driven many-body system characterized by an intermittent scale-free 1232 response to the external perturbations and global instability, which implies that the system can 1233 1234 adjust to rate of changes, as in the case of magnetospheric system without losing its signatures of 1235 critical dynamics (Bak et al., 1987; Chang et al., 1992, Uritsky et al., 2003). 1236 Similar effects can occur in the ionosphere since the ionosphere is coupled to the ionosphere as

mentioned earlier. Therefore ionosphere experiences the effects of solar wind as it impacts the magnetosphere. The lower values of chaoticity at the equinoxes have been suggested to be as a

result of the fact that the internal dynamics of the system adjusts itself to the perturbation from

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the influx of the solar wind which maximizes at the equinoxes. The suppression of the internal 1240 1241 dynamics of the ionosphere as a result of its modification by external stochastic drivers like the solar wind has been described by (Unnikrishnan et al., 2006; 2010; Ogunsua et al., 2014). The 1242 resulting wavelike pattern might be more obvious at the equatorial region due to the proximity of 1243 the region to the sun which lies directly above the equator during the equinoxes. 1244 1245 Although there is a scale-free response as mentioned before, the suppressed internal dynamics does not change its signatures as the ionospheric system retains it chaotic dynamics but only at a 1246 lower level. This is described by the drop in the values of the two parameters describing the 1247 chaoticity and dynamical complexity of the ionosphere, that is, the Lyapunov exponent and 1248 1249 Tsallis entropy. 1250 1251 References 1252 Uritsky, V. M., Klimas A. J., and Vassiliadis D.,: Evaluation of spreading critical exponents 1253 from the spatiotemporal evolution of emission regions in the nighttime aurora, Geophys. Res. Lett., 30(15), 1813, doi:10.1029/2002GL016556, 2003. 1254 1255 1256 Bak, P., C. Tang, and K. Wiesenfeld, Self-Organized Criticality-an Explanation of 1/F Noise, Phys. Rev. Lett., 59(4), 381-384, 1987. 1257 1258 1259 Chang, T., Low-Dimensional Behavior and Symmetry-Breaking of Stochastic-Systems Near Criticality-Can These Effects Be Observed in Space and in the Laboratory, IEEE Trans. On 1260 Plasma Sci., 20(6), 691-694, 1992. 1261 1262 1263 Chang, T.: Sporadic localized reconnection and multiscale intermittent turbulence in the magnetotail, AGU Monograph No. 104, Geospace Mass and Energy Flow, (Eds) Horwitz, J. L., 1264 Gallagher, D. L., and Peterson, W. K., p. 193, (American Geophysical Union, Washington, D. 1265 C.), 1998. 1266 1267 1268 Chang, T., Self-organized criticality, multi-fractal spectra, sporadic localized reconnections and intermittent turbulence in the magnetotail, Phys. of Plasmas, 6(11), 4137–4145, 1999. 1269 1270 Chapman, S. C., Watkins, N. W., Dendy, R. O., Helander, P., and Rowlands, G.: A simple 1271 1272 avalanche model as an analogue for magnetospheric activity, Geophys, Res. Lett., 25, 2397-1273 2400, 1998. 1274 Consolini, G., Marcucci, M. F., and Candidi, M.: Multifractal structure of auroral electrojet index 1275 data, Phys. Rev. Lett., 76 (21), 4082-4085, 1996. 1276 1277 Freeman, M. P., and N. W. Watkins, The heavens in a pile of sand, Science, 298, 979–980, (1) 1278

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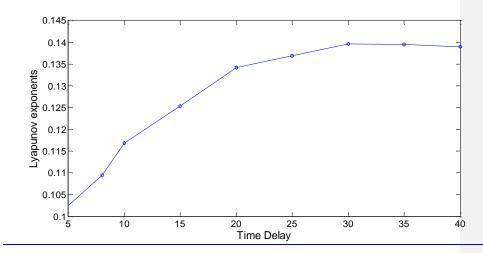


Fig 1 Different Lyapunov exponent computed for different time delay with a constant embedding dimension

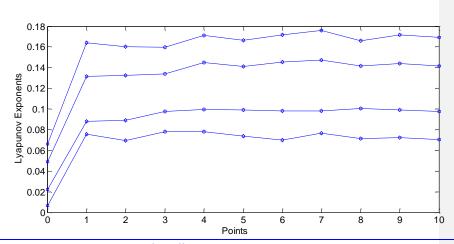


Fig 2 Different Lyapunov exponent computed for different embedding dimension with a constant time delay

III. Finally the comment on the small font size of the title and axis labels for Figures 1,2,7,9,10,11, and 13 will be readjusted for the main published paper.

The authors will also like to state here that additions and subtraction to the main published paper will be subject to editor's recommendations

1328 COMMENT, RESPONSE AND EXPECTED CHANGES TO MANUSCRIPT BASED ON
1329 THE FIRST REFEREE REPORT (NUMBER 2) ON THE PAPER TITILED:

1330 "THE TRANSIENT VARIATION OF THE COMPLEXES OF THE LOW LATITUDE
1331 IONOSPHERE WITHIN THE EQUATORIAL IONIZATION ANOMALY REGION OF

1332 **NIGERIA**".

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1. ANONYMOUS REFEREES COMMENTS

The paper entitled 'The Transient Variation of the Complexes of the Low Latitude Ionosphere within the Equatorial Ionization Anomaly Region of Nigeria' by Rabiu et al. attempts to study the utility of Lyapunov exponents and Tsallis entropy computed from the total electron content (TEC) derived from GPS observations in characterization of the dynamical response of the dip equatorial ionosphere to external influences. One major shortcoming of the paper is the assumption on the part of the authors that the only external influence is due to magnetospheric forcing seen during magnetic storms. There is also significant forcing from below the ionosphere, which causes day-to-day variability, even during magnetically quiet periods, in the occurrence of the equatorial plasma bubble that produces the largest changes in the de-trended time series for daily TEC at a low latitude station, after the diurnal variation has been removed, as has been done in the present paper. It is not clear from the results presented in the paper, that non-linear dynamics of the low latitude ionosphere is mainly determined by geomagnetic storms and substorms. On the whole, the quality of the paper is poor, with some glaring errors mentioned in the next paragraph. The paper has several basic scientific issues that need to be addressed and corrected before the authors even proceed to present results pertaining to the nonlinear dynamics of equatorial and low-latitude ionosphere. These are listed below:

- 1. On p 1960, lines 13-14 are incorrect. In the dip equatorial region and at the low dip latitudes (all below 3.5 degrees) where the stations considered by the authors are located, the magnetic field B is horizontal and perpendicular to the dip equator and not parallel to the equator as stated by the authors
- by the authors.
- 2. On p 1960, lines 14-17: 'Off the equator map along F region' are meaningless. What do the authors mean by 'the eastward electric field (E) of the E-region interacts with the magnetic field B during the day'? There is no E X B drift of the E region plasma as a whole because in the E region only the electrons are magnetized while ion motion is influenced more by collisions with neutrals.
- 3. On p1864, the authors fail to mention what the set u_i consists of and how do they obtain this
 set. Moreover, Ti in equation (3) is not the diurnal variation reduced time.

1364	5. What does the delay representation of the time series shown in Figure 5 represent?
1365	6. What are Δx and r in Equation (5)?
1366	7. In equation (6), limit has to be calculated for $r \to 0$, and not $r \to \infty$.
1367 1368 1369	8. Instead of the lengthy write-up on Tsallis Entropy, which has been better described in cited references, the authors should mention the formula that they have used to calculate the Tsallis Entropy from their data.
1370 1371 1372	9. Equation (14) is incorrect and need not be given. Authors should write Eq. (15) correctly: they are summing over i and x is characterized by n? The authors should take greater care to write correct equations.
1373 1374	10. Page 1876, lines 1-2. There is no such thing as 'acoustic motions of the atmosphere electromagnetic emission'!
1375 1376	11. On p 1876, line 23. The solstices are not necessarily months of low solar activity. Some major magnetic storms have occurred during the solstices.
1377 1378	2. AUTHORS RESPONSE TO COMMENTS
1379	We appreciate the referee's comment there are other external influences is other than
1380	magnetospheric forcing seen during magnetic storms but also significant forcing from below the
1381	ionosphere. We like to state that this was considered by the authors but it was silent in the writer
1382 1383	up for instance gravity waves were meant to be mentioned as part of the influences on the internal dynamics of the ionosphere. Thank you for the comment necessary additions will be
1384	made.
1385 1386	 Please refer to references for proper clarification on the direction of B field close to the equator.
1387 1388	2. The comment was right E X B drift only in the F region; necessary corrections will be effected in the main paper where necessary.
1389	3. Please refer to other cited literatures where such techniques have been used.
1390 1391	 Please refer to all given references on the techniques of embedding and phase space reconstruction.

4. Authors fail to mention the formulae they have used to calculate the mutual information and

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the number of false nearest neighbours.

- 5. Please refer to all given references on the techniques of embedding and phase space 1392 1393 reconstruction. 6. Please refer to the reference (Wolf et al 1985) on the computation of Lyapunov 1394 1395 exponents. 7. Lyapunov exponents should be computed for $r \to \infty$. Please refer to the reference (Wolf 1396 et al 1985) on the computation of a Lyapunov exponents 1397 8. The write up was necessary as it gives relationship between Lyapunov exponent Tsallis 1398 entropy as recommended by previous referees in the previous paper since both 1399 parameters are being used together and also comparatively. The mathematical 1400 expressions describing the Tsallis entropy have been given in the text please consult 1401 1402 referees for further understanding. 9. The equations are necessary, (14) is the basic moving average method equation and the 1403 1404 correct equations are $u[t] = z[t] * w[n] = \frac{1}{2k+1} \sum_{i=-k}^{k} x[n-i].$ (14)1405 1406 $u[t] = z[t] * \omega[n] = \frac{\sum_{i=-k}^{k} A_i * x[n-i]}{\sum_{i=-k}^{k} A_i}$ 1407 (15)1408 Refer to literatures. 1409
 - 10. Please see reference and other Literatures like the Physics of the ionosphere and magnetosphere by John Ashworth Ratcliffe to understand the concept of acoustic motions in the upper atmosphere.
 - 11. Yes you are right there can be major storms in solstice on that point. However, many literatures have proven that the effect of solar wind and activities on the earth are usually higher during the equinoxes. The statement can be rephrased.
- We appreciate the referee's comments and we shall look into a few points we find relevant but we will need the referee to refer to literatures where necessary. Thank you.

3. AUTHOR'S CHANGES TO MANUSCRIPT

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1419 1420 <u>I.</u> The authors will mention external factors from below the ionosphere in the published paper

1421	II. More clarity will be shown comment 2 based on the fact that the field due to the dynamo
1422	in the E-region maps along the magnetic field to the F region altitudes above the equator.
1423	III. Typographical errors in equations 14 and will be corrected
1424	IV. Comment 11 will be looked into statement will be suitably rephrased with editors consent
1425	The authors hereby state that the additions and inclusion will be subject to editor's
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1447 COMMENT, RESPONSE AND EXPECTED CHANGES TO MANUSCRIPT BASED ON
1448 THE FIRST REFEREE REPORT (NUMBER 3) ON THE PAPER TITILED:

"THE TRANSIENT VARIATION OF THE COMPLEXES OF THE LOW LATITUDE IONOSPHERE WITHIN THE EQUATORIAL IONIZATION ANOMALY REGION OF NIGERIA"

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1. ANONYMOUS REFEREES COMMENT

1454 Dear Editor,

I have reviewed the discussion paper "The transient variation of the complexes of the low latitude ionosphere within the equatorial ionization anomaly region of Nigeria" by Rabiu et al. submitted for possible publication in Nonlinear Processes in Geophysics (NPG) and found it unacceptable for publication in NPG. The paper studies using Lyapunov exponents and Tsallis entropy the day to day variations in 2011 of the chaoticity and dynamical complexity of the ionosphere over Nigeria. My major concern is that there is a very high degree of similarity with their previous work recently published in NPG (Ogunsua et al., 2014) that studied using the same methodology and techniques the chaoticity and dynamical complexity (as before) of the ionosphere over the same region during quiet and disturbed days of the same year (the similarity report of the present paper resulted in a similarity index of 52% with just one source, i.e., their previous NPG work!). There is nowhere apparent in the paper (abstract, discussion, conclusions) the differences or the improvements from their previous publication in the same journal (apart the fact that in 2011 they used 3 stations and now 5 and they now deal with the day to day variation instead of quiet and disturbed days that they considered before). For instance, I have found around 10 references in the text to be missing from the References' list that I was able to find in the paper by Ogunsua et al. (2014)! Another major issue to me concerns their time series analysis using the Tsallis entropy. I was unable to find the Tsallis q index value they used to calculate the corresponding entropy. My guess is that they have used the q value of 2 indicated by Anastasiadis et al. (2005). However, in the references they have cited there is a methodology to infer an optimum q value for the problem in hand. Even that would have been some improvement from their previous published work but it is missing.

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2. AUTHORS RESPONSE

In response to the Anonymous Referees report #3 we appreciate his/her efforts to understand what the Authors of the paper are presenting. This report responds to the issues raised by the referees which include (1) The appearance of similarity between the new paper and Ogunsua et al., (2014) (2) The optimum value of Tsallis entropy q index used in this work. (3) The error in

references, which involves the references found in Ogunsua et al. (2014) that were not included in the new paper. We respond by saying that the understanding shown by the writer of the referees report #3 is not in line with the focus of the work.

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The Authors will like to react to the fact that the paper appearing to be similar to the previous one, is important for the research at hand based on the fact that the findings in this paper is a follow up to the research from the previous paper. However the research concepts and focus are totally different. The previous paper considered different geographical conditions by looking at the 5 international most quiet day and international most disturbed day classification while in this case the day to day variations are being considered which help understand the ionospheric response in quiet, relatively disturbed and storm. Here we'll use this opportunity to show that the day to day variation is not the same as the most quiet and most disturbed day classification as they can be applied to test and reveal different things in the variation of the upper atmosphere. This day-to-day variations have been found to reveal majorly different things that were not revealed in the previous paper, these include the wavelike patterns revealing troughs during equinoxes resulting from effects of solar activities on the ionosphere in the entire year, it was pointed out that this drop in values at equinoxes are based on self organization phenomenon (SOC) that could occur in the upper as described by (Consolini, Chang et al.,). Although the part of the methodology that is concerned, with computation of the nonlinear parameters might be similar It does not mean that the focus of the paper is the same. We cannot change the nonlinear analysis techniques. After the nonlinear analysis were carried out the results were further analyzed to show some clarity in the ionospheric response in terms of the variations of this parameter. The results in this present work differs completely form the previous work as it does not only revealed the unstable day to day transient variation of the chaoticity and dynamical complexity of the ionosphere but it also shows a seasonal trend of variation the even requires more studies.

Therefore final assessment of result is different in this paper compared to Ogunsua et al. (2014). Hence the authors are considering further the sensitivity of the ionosphere to daily changes in the internal dynamics and in different external factors. All these result cannot be put together in the same paper as there will be a lumping up of different ideas and concepts which might distastefully destroy the notion of building one research upon another. The Authors will like to point out to the anonymous referee that methodologies of different papers can be slightly similar but the papers could come with different concepts and focus such that the two papers can examine different behavior or responses of a system being studied as the case may be. Also the same set of data could be used to examine different responses of a system to the same phenomena.

On the second point the authors only pointed out the relationship between Tsallis entropy and Lyapunov exponent and the relationship between the q index and Lyapunov exponent to justify the essence of using the two parameters comparatively this was written in Ogunsua et al 2014 due to the referee's recommendation, which we have found to be very useful more on the

optimum value of q index can be found in different texts and references specializing in the area as cited. It should be hereby noted that this work is mainly based on its application the responses to these parameters, which have been shown in the work. Therefore it will not dwell too much on the methodology of optimum value of q index as the references like Anastasiadis et al., 2005 has been cited in the paper to help the reader.

Similar things compared to Ogunsua et al. 2014 were reproduced in some areas of this new paper to re-establish the fact that both quantities can be used as indices to explain the processes and dynamics of the ionosphere and its reaction to the external influences. To be able to establish this completely, several tests using similar analytical techniques should be carried out using different concepts that have been studied before or looking at different behaviours ranging from transient to seasonal behaviours and other responses of the ionosphere before it can be totally established that the two parameters can be used as indices for continual interpretation of the ionospheric processes and dynamics. And this new paper written as a follow up to Ogunsua et al. 2014 is another step in that direction.

The Authors will hereby maintain that the claims in report #3 are not enough for the anonymous referee to make such statement on the acceptability of the paper. The authors will however take note of the indication of the optimum value of Tsallis entropy used even though the main idea behind this work is to see the ionosphere dynamical response, using these parameters were being established. Secondly, the references will be carefully checked and corrected appropriately by the authors. All needed adjustments will be applied where necessary.

3. AUTHOR'S CHANGES IN THE MANUSCRIPT

Authors will disclose the optimum value of the q index used for computations in this work. But this is subject to the editor's recommendation however we strongly believe that this is not the main focus of the paper. The authors will also like to state the every other addition or removal of texts based on this reviewers comment will be subject to editor's recommendation, even though referees view does not align with the focus of the paper.