



An IoT based Smart Positioning and Decision Management Application with Real Time Kinematics Information

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ABSTRACT

An IoT based smart positioning measures are useful to the society for accurate mapping of boundaries and to do the localization of geo graphical area. The existing approach uses GNSS technologies, and the boundary values can be improved further by applying advanced technique. In this paper, a novel approach for optimization of the positioning technique is proposed. A digital classification and optimization of area mapping, enhancement of correlation data, analysis through expert system and clustering of delineation zones using machine learning algorithm is implemented as part of the solution. The precision accuracy optimization is achieved by error correction of zonal map boundaries obtained through GNSS satellite system. The real time kinematic technique is applied to achieve the better accuracy. The experiment is carried out in real time at Nagamangala region and real time data received from the radio frequency sensors, satellites are analysed, calibrated, and improved further to obtain the precision accuracy up to 99.9% with the accuracy of 2mm. The proposed solution is useful in real time sectors to manage the data related to real time tracking, safety critical applications, health care, military, mining, geological surveys, and agricultural applications.

Keywords:

Real Time Kinematics, IoT Application, Smart Positioning and Management, Sensor, RFID, Real Time Navigation, Localization, Geo Data, Precision Decision Making

1. Introduction

The Satellite Geo spatial data is obtained from the remote sensing device that has the raw GNSS data in WG84 based format [1]. The raw GNSS data is humungous, and data is generated dynamically within a milli second. The extraction of NGS CORS data from the generated data is challenging and requires optimal algorithm [2,3]. After the data acquisition, data processing and data transformation is to be performed further by using ML/DL based algorithms [4,5, and 6]. Precision agriculture needs accuracy in the creation of management zones [7]. Kriging

interpolation method though gives the solution, precision accuracy is still a challenge in agriculture, Land Survey, and other field specific applications [8]. Real Time Kinetics (RTK) is the mechanism used for deriving positioning accuracy in any area or zones [9, 10]. The proposed solution in this paper uses the RTK based solution to determine the accuracy of GNSS positioning.

Considering the diversion of the points of projections, one needs map the directions across the dimensions of coordinate systems. The spectral difference between the coordinate points interprets the accuracy level of the projection points [11]. Point level fusion mapping can be done to extract the coherent information of coordinates and then by applying the decision rule to obtain new coordinate projection space [12]. Real Time Kinetics can be used in IoT applications such as Smart Cities, Agriculture, Environment, Robotics, and Autonomous Vehicles etc. [13]. Real Time Kinetics gives the positioning references using reference station that is fixed and movable station [14]. Real Time Kinetics works with dual frequency, and it mitigates the positioning errors by differentiating the positions between reference station and movable station. In Vietnam and other European country, dual frequency based real time kinetic measurement technique is used [15, 16]. With optimized real time kinetics protocol implementation through GNSS receiver and NTRIP caster framework, it is possible to obtain the precision accuracy [17]. The similar approach is incorporated in this paper with optimization in real time kinetics.

Trajectory Planning method is used in Unmanned Aerial Vehicle delineation. A Trust Region Filtered Sequential Convex Programming is proposed to build UAV trajectories [18]. The image fusion space with geo plots needs projection mapping calculations and involves complicated computations [19]. Mixed Integer Linear Programming. To solve the trajectory planning, Random Tree Exploration and Informatics interpretation can be used [20]. The computation methods and algorithms logics increase the complexities for plotting the trajectories [21]. Convex programming methods are suggested for improving the trajectory generation [12 and 22]. Comprehensive Analysis is needed to achieve geo fencing through China's BDS-3 based carrier shift methods for range specific GNSS data [23]. Tukey and K-Means methods are suggested by Massimiliano et al. to implement Single Point Positioning on GIS applications [24]. Determining spatial differentiation of geographical area using Ecological Redline Area is proposed by Deng Sui Chen et al. [25]. To develop tractable simulations trial and error-based methods are followed in actual model scaling analysis [26]. A polynomial model is the common method for correction in geometric. The order of the polynomial and remote sensing is not co related and hence machine learning algorithm such as ENVI deep learning model is proposed by Weicheng Xu et al. [27]. Imperative areas are detected through down sampled Global Nighttime Light using multi source spatial variables as proposed by Yang Ye et al. [28].

Multi spectrum Instrument based satellite positioning data on the land area was studied and solution to spatial variability is discussed in the paper of Mortz K. Lehmann et al. [29]. Monitoring forest area based on multi source remote sensing by observing phenological surfaces of the environment is studied by Yali Zhang et al. [30]. A sequence of extraction of points using Gated Recurrent Unit framework and neighborhood network hood algorithm is proposed by Yi he et

al. [31]. Horizontal coordinate points for the transformation of points in longitudinal axis of the original coordinate system can be computed by using gradient component derivation method as proposed by Andreas A. Beckert et al. [32]. Satellite images are analyzed through remote sensing computations using optimized algorithm was suggested by Hu et al. [33]. High powered trajectory optimization to provide tri axis commands and attitude control approach is proposed by Mazinan et al [34]. However, scale dependent spatial pattern quantification is challenging and needs power law predictions as proposed by Qun Ma et al [35]. With back tracking the traces of encroachment space patterns across Southern Great Plain, a threshold value for accuracy assessment was studied by Xuebin et al. [36]. It is proven that utilization of GIS based methodologies are needed in various regions, zones and tracking etc, [37]. The delineation of land area may not be accurate at all locations [17, 38]. Bumairiyemu Maimaiti et al suggested to do mapping of classifications, expansion scaling and relevant parameter calculations done through ArcGIS software [39]. Diversified land zones and mapping of the areas through consistent characterization with an accuracy of 4.77m is implemented by Yang, F. and Zeng, Z [40]. Intense investigation is done by Shreshtha et al., in the mountain region of Asia where complexities are high due to unstable slopes. However, positioning accuracy is still a challenge and requires research.

Precision accuracy for Galileo constellation positioning can be determined by single frequency modeling and computation as proposed by Bahadur, B [13]. Post Processing Kinematics and Real Time Kinematics are the methods suggested to achieve accuracy in unmanned vehicles and the outcome shows that Root Mean Square of 0.0189m is achieved by Nicola et al and team. [14]. In urban areas, Global UrbanNet based framework is useful as proposed by Yanfei Zhong et al and team in their work [41]. Taddia, Y et al., proposed Network RTK and DJI Phantom RTK modes to achieve GNSS accuracy and have achieved 5cm offset with vertical residuals [42]. In our paper, a similar approach is adapted to achieve an offset accuracy of 2mm. The proposed IoT application is useful in use cases such as agriculture precision management, real time object tracking management and in safety critical applications and management.

2. Methodology

Data Format

The Smart positioning application makes use of the dynamic data generated by active satellites and is represented in the format called NGS CORS data. The raw data received from the satellite constellations are managed by real time kinematics technology, RFID sensor and Geo Informatics approaches. The positional data parameters such as Longitude, Latitude, Altitude, time, mode of fixing the position, accuracy data in 2D and 3D and live satellite information etc. A sample of data format received through the simulation software is represented in Fig. 1.



Fig.1DataFormat

The real challenge of engineering a complex system is to aggregate the electric, electronic, mechanical, and computing elements. The tool simulations or modeling is needed to study and analyze the behavior and performance of the elements. The signaling strength, gain and better impedance with low cost was needed for deriving delineation solution at its first place. The optimization technique needed to improve the harmonic response and analysis of antenna to be performed. A monopole antenna is analyzed, and mitigations are proposed by Jadon et al. [43] and Chand P S P etal[44]. The hardware module of RTK consists of RTK basestation and RTK Rover station along with the fabrication of patch antenna designed exclusively for the proposed IoT device.

Pattern Analysis of the Antennas

The design, fabrication, and pattern analysis of the antenna such as Pifa antenna, Microstrip element Antenna, Patch Microstrip inset Antenna, Patch Microstrip Circular Antenna etc. are performed in real time to achieve better accuracy. The elevation and pattern analysis of Patch and Pifa antenna is depicted in Fig.2 and Fig.3.

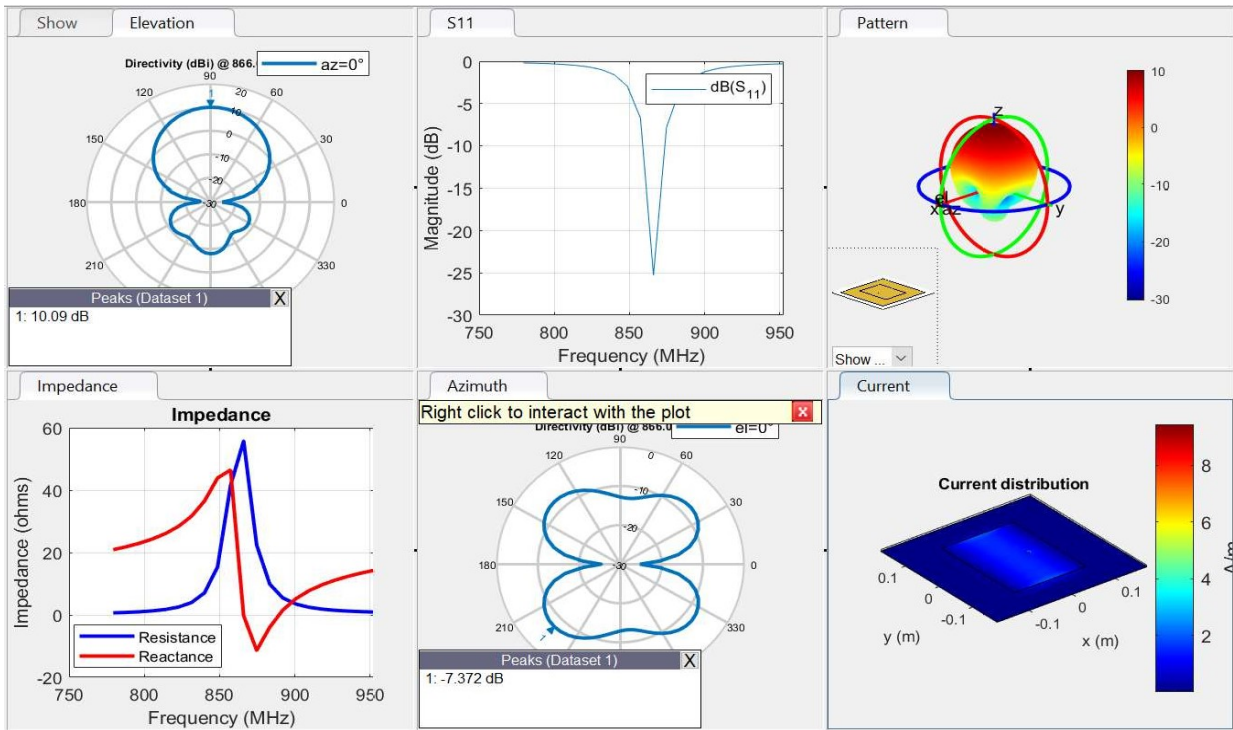


Fig2. Pattern of Patch Microstrip element Antenna with 155mm x 155mm dimension

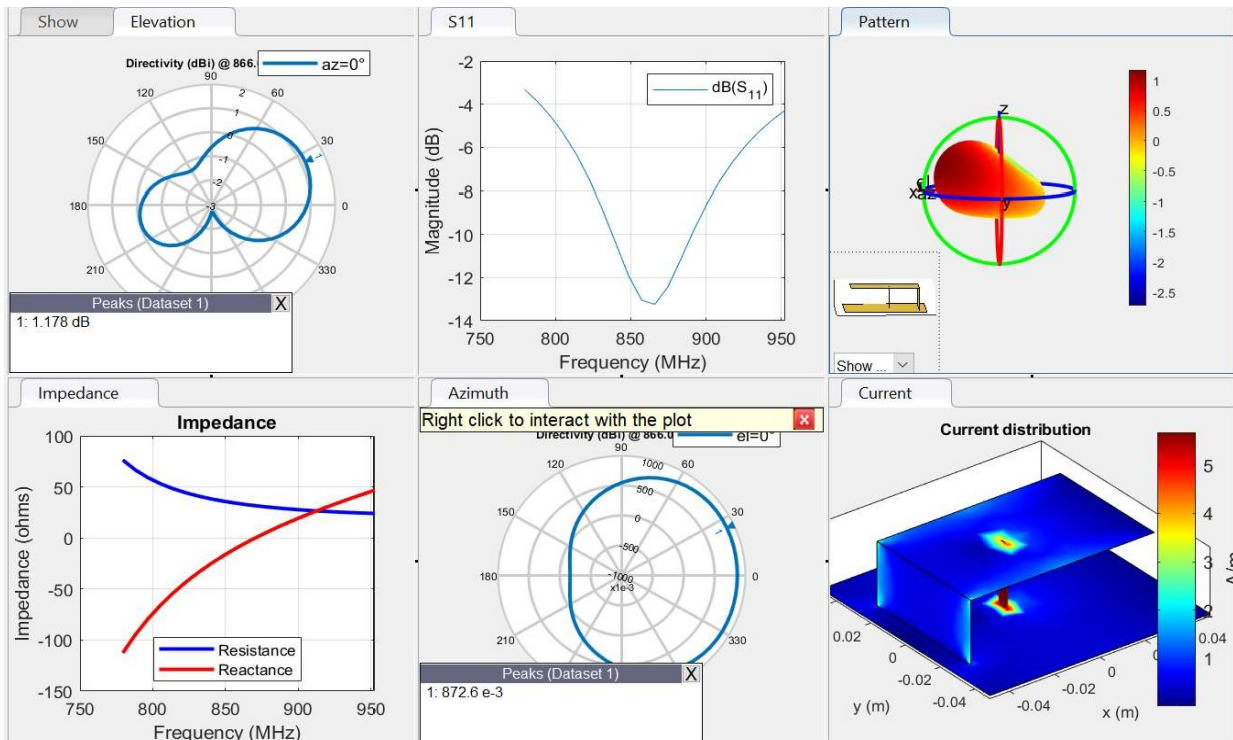


Fig.3 Pattern of Pifa Antenna with 45mm (x-axis) x 50mm (y-axis) x 20mm (z-axis)

Patch Antenna with Micro strip element is designed and manufactured further to improve the gain of signals. Dipole antenna, Fractal antenna, Helix, loop and spiral antenna are redesigned, and

It is verified that in set fed antenna gives the better result in comparison to other antennas as depicted in Figs. 4 to 7.

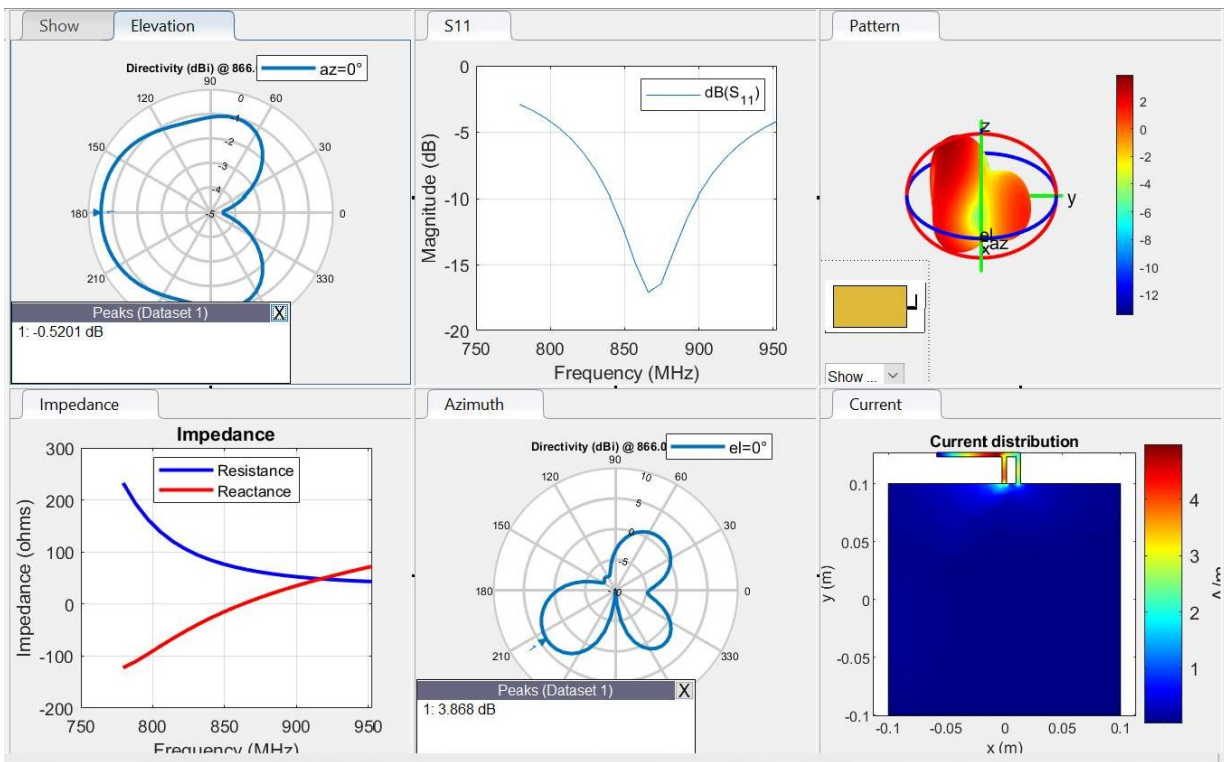


Fig.4 Pattern of Inverted F coplanar Antenna with 100mm x 100mm dimension

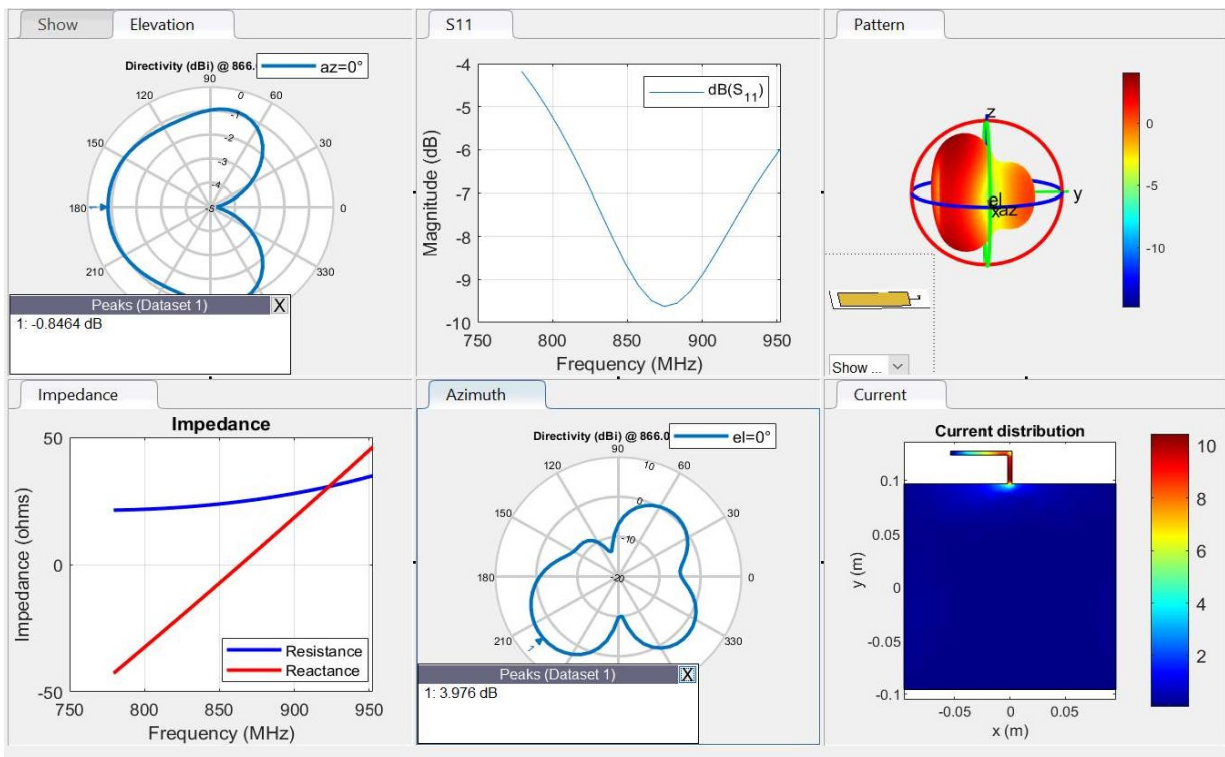


Fig.5 Pattern of Inverted L coplanar antenna with 100mm×100mm dimension

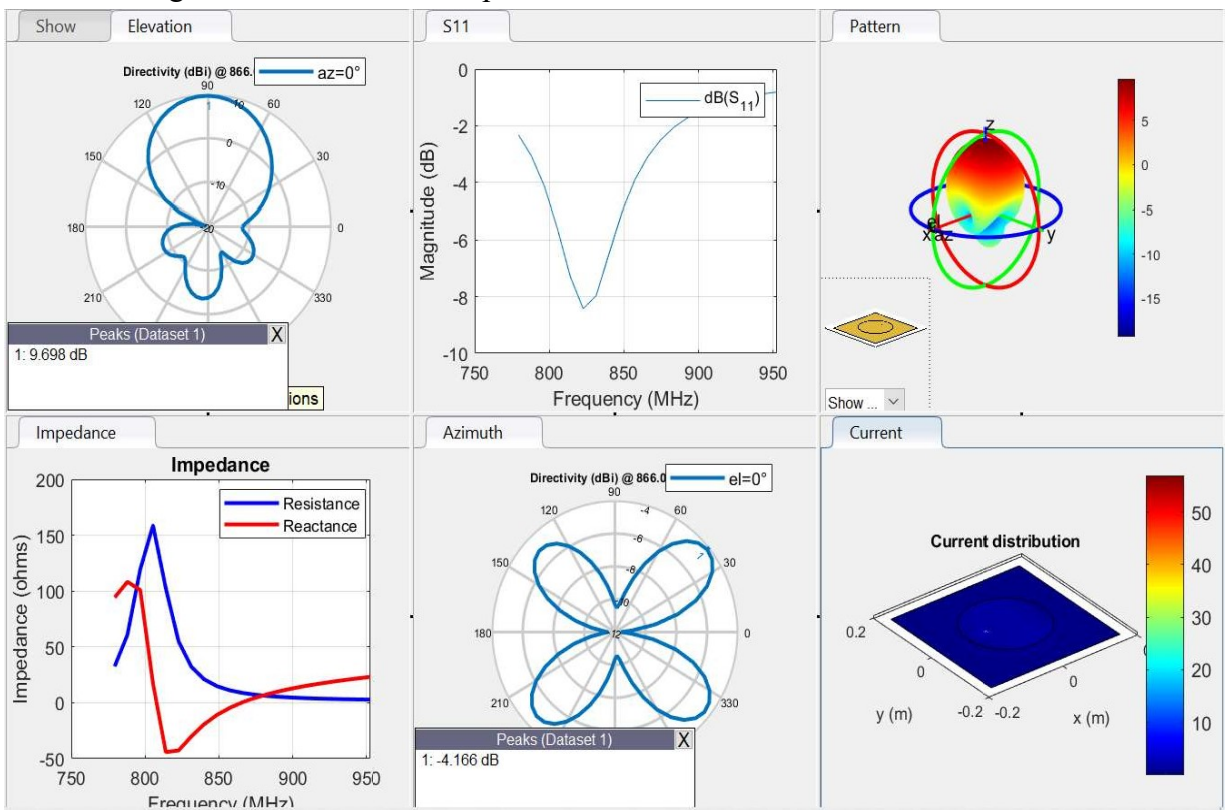


Fig.6 Pattern of Patch Microstrip circular antenna with 155mm×155mm dimension

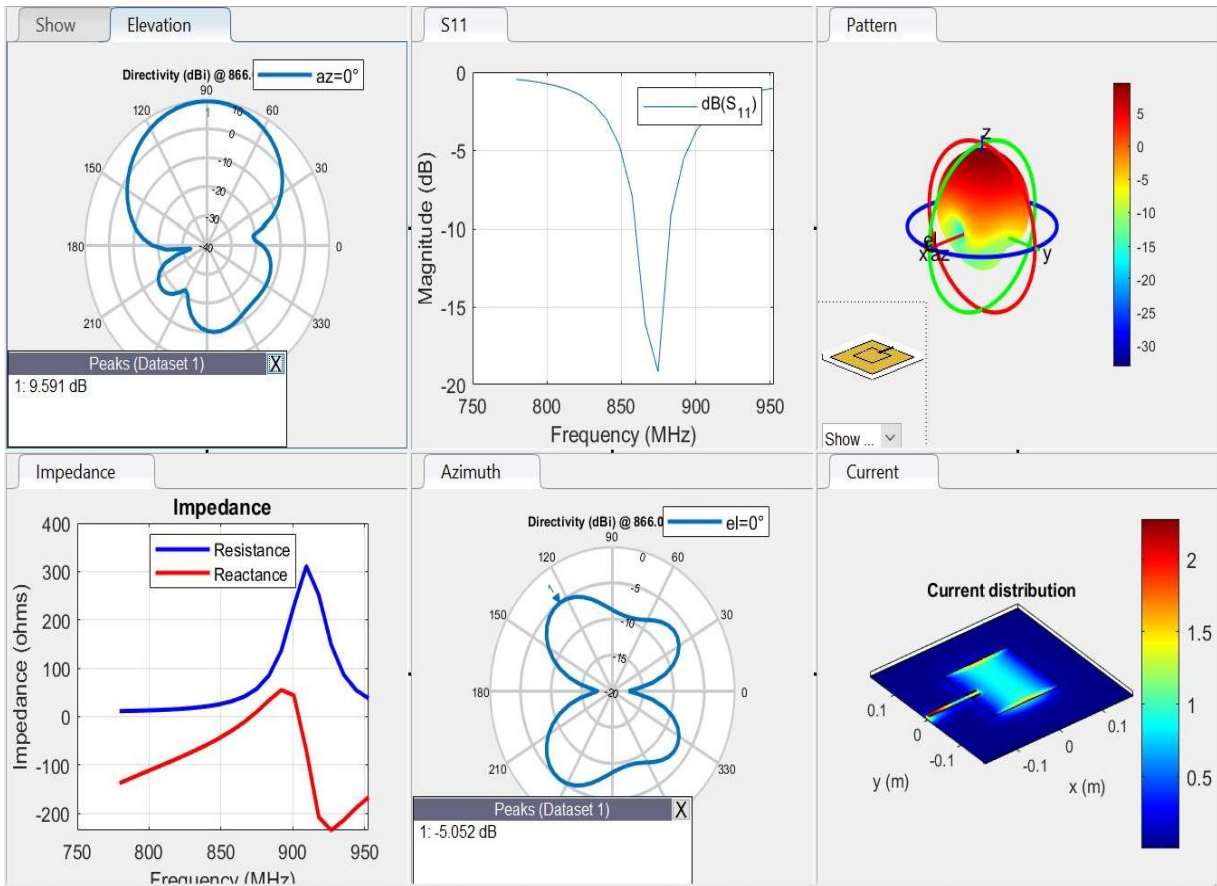


Fig7. Pattern of patch microstrip inset fed antenna with 155mm x 155mm dimension

It is evident from the pattern analysis that the microstrip antenna with patch inset fed antenna is efficient and can be used in the proposed system to improve the accuracy. The comparative analysis of the antenna is shown in Table 1.

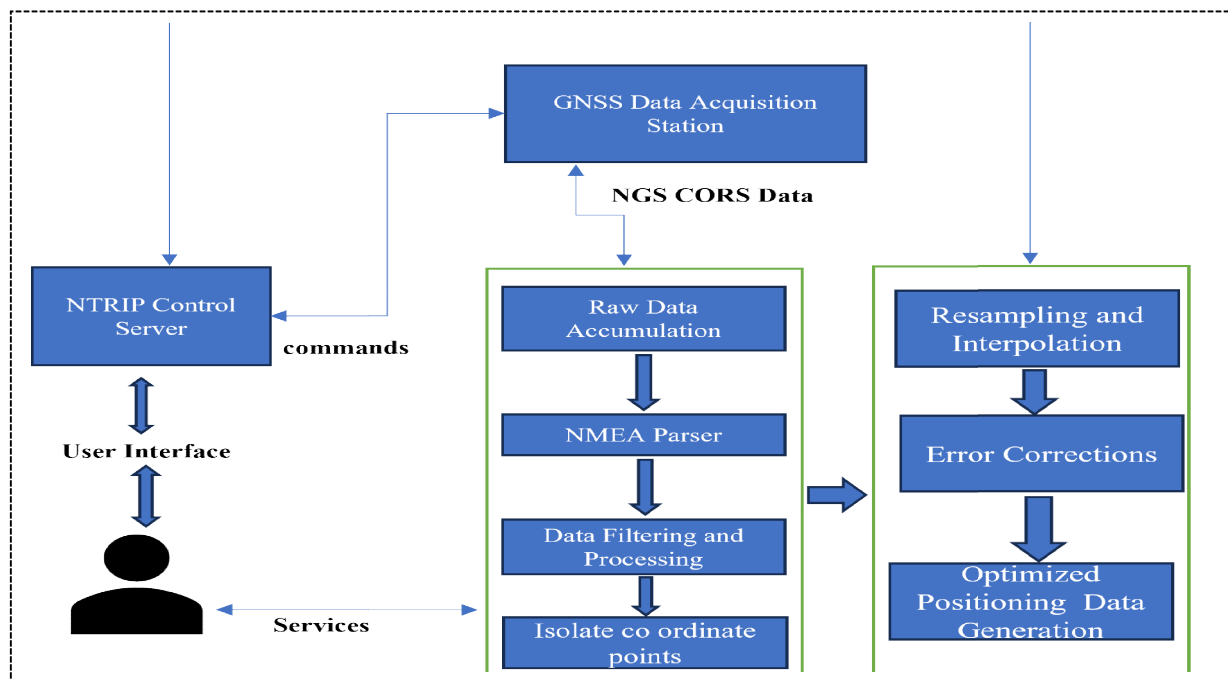
Table 1
Pattern Analysis of Antennas

Antenna Types	Dimension	Impedance @950 MHz	S11 Frequency @ 865 MHz	Distribution
Patch Microstrip element	155*155m	-20 to +60	25dB	high
Pifa Antenna	50*20mm	-100 to +100	13dB	low
Inverted F Co Planar	100*100m	-100 to +200	16dB	low

Inverted LCoPlanar	100* 100mm	-50to+50	9dB	low
PatchMicroStrip Circular	155*1 55mm	-50 to +150	8dB	low
PatchMicroStrip Insetfed	155*1 55mm	-200to +400	25dB	High

Framework for Decision Making and Information Management

The process framework used in this paper for obtaining the precision accuracy is represented in Fig. 8. The model consists of CLI driven control server, Data Acquisition station, and User Services. The processing activities at the services interface consists of parsing logic with optimization, Isolation of coordinate points, Interpretations, Corrections, Generation of new coordinate points and finally distribution of data. Real Time Kinematics based positioning is a technique that performs positioning computation. The hardware integrates geodata receiver comprising Rover and Ranger stations and is communicated through the server. The IoT system is deployed on the locality identified and obtains data from satellite. The received raw data is finetuned further and is optimized dynamically at the edge server. The satellite response generated is extracted automatically once every 4 milliseconds and is displayed on mobile device through android application. A wireless sensor tested to record the metadata and calibrate the timestamp according to the recorded data segment is useful as default gateway [50]. The process



framework used in this paper for obtaining the precision accuracy is represented in Fig. 8.

Fig. 8 Data Processing Framework

The model consists of CLI driven control server, Data Acquisition station, and User Services. The processing activities at the services interface consists of parsing logic with optimization, Isolation of coordinate points, Interpretations, Corrections, Generation of new coordinate points and finally distribution of data. Finally, data processing and decision is performed through the data processing logic.

Let us gather the data with distance record. After recording distance every 4 seconds thereafter record the elapsed time, t and distance, $dist(t)$ until the final distance value 1 cm. Let us consider exponential function as shown in Eq. (1)

$$dist(t) = Ee^{-pt} \quad (1)$$

where, E and p are the positive constants representing error correction value. Let E_r be the exponential spatial threshold value. Then, the Eqn. 1 is updated as shown in Eq. (2)

$$dist(t) = Ee^{-pt} + E_r \quad (2)$$

The initial recorded distance value is $dist(0)$. Then the deviation is calculated as per Eq. (3)

$$dist(t) = (dist(0) - E_r)e^{-pt} + E_r \quad (3)$$

If the function fits all of the data, then any of the ordered pair should satisfy the function. For example, use pair $(d/2, t/2)$ from the recorded data and the value of error correction constant p can be set with 0.01. Then, the distance amount can be derived as per Eq. (4).

$$dist(t) = (d_0)e^{-0.01t} + d_1 \quad (4)$$

This step to be repeated to find the constant p for all the pairs. Once the distance is determined, the time(t) vector can be determined as per the Eq. (5).

$$time(t) = 2\pi \div \sqrt{E_r \times dist_t} \quad (5)$$

During period, $time(t)$, the correlation variation per time is obtained as shown in Eq. (6)

$$2\pi \times \frac{dist(t)}{time(t)} \quad (6)$$

Distortion correction can be done by using first order linear equation. The coordinations system points (x, y) are related by mapping functions obtained from CORS data set, (u, v) as given in

Eq.(7).

$$u=f(x,y) \quad \text{and} \quad v=g(x,y) \quad (7)$$

where f and g are polynomial.

The translation due to distortion is represented by 1st degree polynomial. The new value u and v for n^{th} order coordinate sets $\{a_n, b_n\}$ is given by Eqn. 8

$$\begin{aligned} u_n &= a_0 + a_1x + a_2y + \dots + a_{n-1}x + a_ny \\ v_n &= b_0 + b_1x + b_2y + \dots + b_{n-1}x + b_ny \end{aligned} \quad (8)$$

The new coordinate system may not have one to one mapping with actual points. The actual to new coordinate difference can be calibrated and is dependent on the degree of the polynomial used in the mapping function and average mapping range is $(n+1)(n+2)/2$. With resampling and interpolation one can correct the misalignment and distortion. Furthermore, error correction and modeling can be done with the optimized computational procedure or technique. Several contiguous coordinate framed data are resampled, interpolated and modelled to achieve approximation values needed for accuracy [52]. The projection of coordinate system and the distortion angle along x and y varies. Precision rate with respect to inclination of coordinate system deviation is represented as shown in Fig. 9.

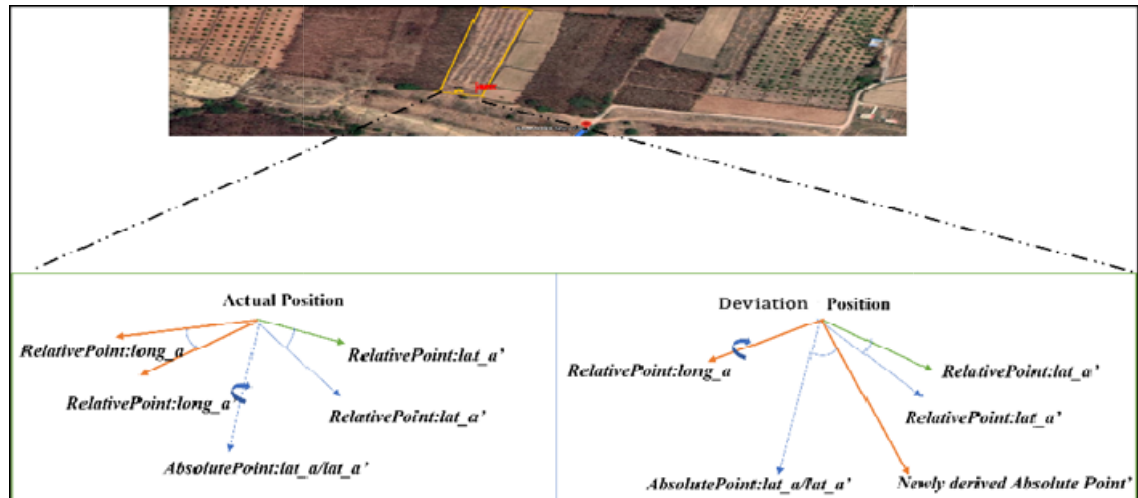


Fig.9. Coordination System Delineation and Deviation Correction

Let the displacement be $\Delta\lambda$ and is represented as shown in Eq.(9)

$$\Delta\lambda = \frac{360}{(\text{Correction} * \text{No of derivations})} \times \text{time}(t) \quad (9)$$

The consecutive data is calibrated to get complete deviation analysis. The periodic repetition parameter Q can be designated as per Eq. (10).

$$Q = 1 + \frac{K}{(N)} \times \Delta\lambda(10)$$

where, K represents distortions observed and N represents number of repetitions. The ratio K/N represents successive deviation parameter. The back propagation based neural network method with optimized Artificial Bee Colony algorithm is used in this paper to optimize the parameter and dynamic threshold technique is used to predict the future approximation of location points.

3. Results

In the proposed IoT system, the both the stations are placed at the distance of 40 kms as shown in Fig. 10. For testing, the RTK base station and RTK rover station is set up at two different locations with the distance of upto 40 km. Once the station is ready, the GNSS receivers start collecting the data via Bluetooth.

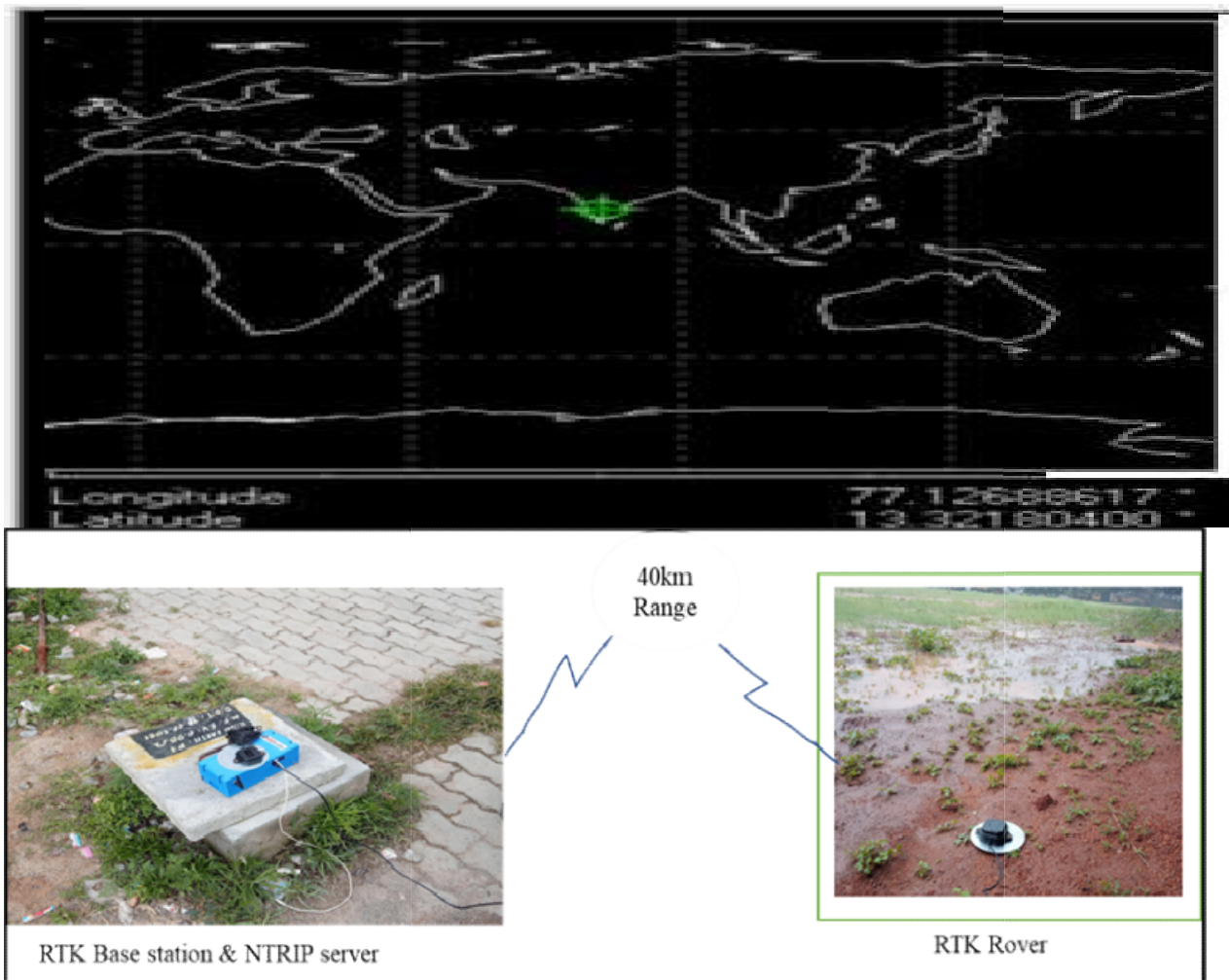


Fig. 10 Setup of RTK Base station and RTK Rover with 40 km distance as shown in Digital Satellite Map

The GPS log is captured in text file and is preprocessed before classification process is applied. The region where the internet facility is not available, the instrumentation profiles are obtained directly from the satellite through GNSS receiver. The constellations captured as shown in Fig.

11. The Geo raw data is collected in the console from the active satellites and smart delineation line in yellow with 2mm accuracy.



Fig. 11 Result of the Smart Positioning with delineation digital marking

The result of the positioning is tracked from the remote location and precision accuracy deviation in mm is depicted in the testing area at the remote server as shown in Fig. 12.

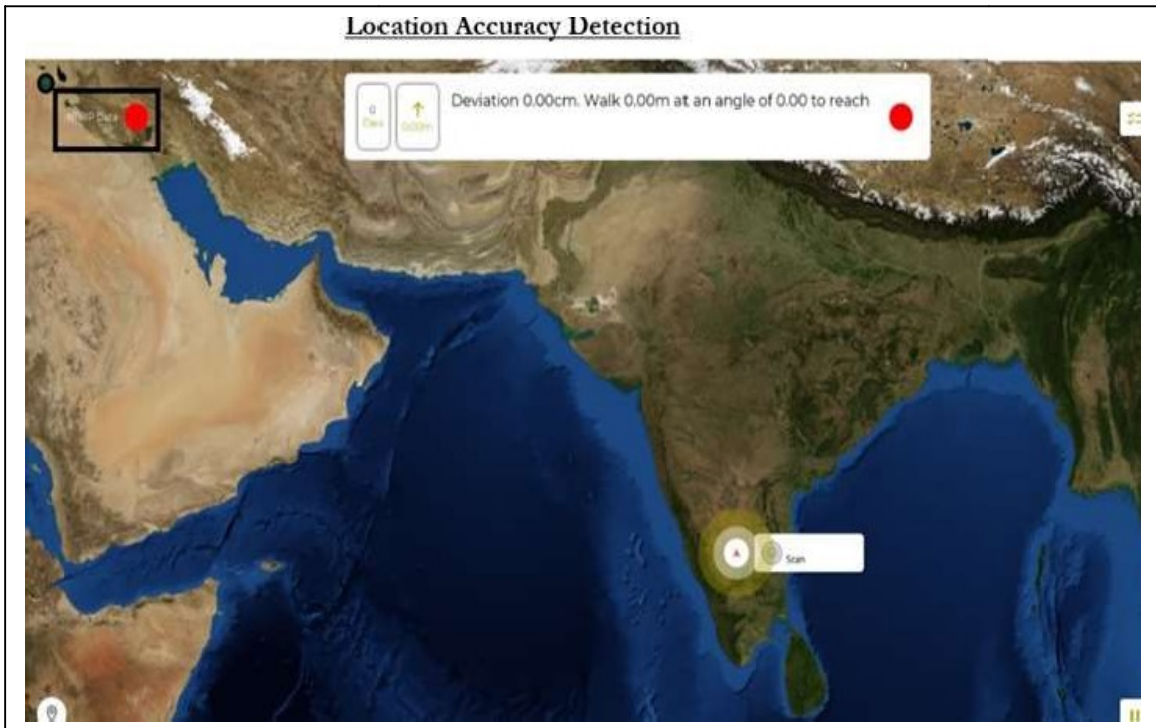


Fig. 12 Positioning Accuracy at 2mm precision and tracking from remote location away at 40 km. The experiment carried out using Real Time Kinetics proves that the location accuracy up to centimeter level is achievable and it is easy to carry out land survey in the area without Internet. The results of parameter simulations and it is proven that fractal boundary antenna is suitable for all satellite application. The data is calibrated once every 4 seconds and each time dynamically the optimization logic is applied. The deviation chart depicts that 100 percent of the data are corrected and positioning accuracy is achieved. The distortions observed are analyzed and as shown in Fig. 13, the location deviations variations are captured and 4% of the location data distortions are observed.

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$GNGGA,101049.00,1319.30817,N,07707.61304,E,1,12,0.75,839.1,M,-85.1
$GNGSA,A,3,05,29,25,20,02,13,15,,,,,1.56,0.75,1.37,1*08
$GNGSA,A,3,75,85,84,,,,,,1.56,0.75,1.37,2*04
$GNGSA,A,3,26,33,,,,,,1.56,0.75,1.37,3*02
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$GNRMC,101050.00,A,1319.30821,N,07707.61310,E,0.055,,010622,,,A,V*1

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Fig.13.RecordofGNSSliveandrawdatafetchingatthetestingarea

The relative point correction is to be done on altitude and relative point coordinates. The amount of relative point corrected is shown in Fig. 14

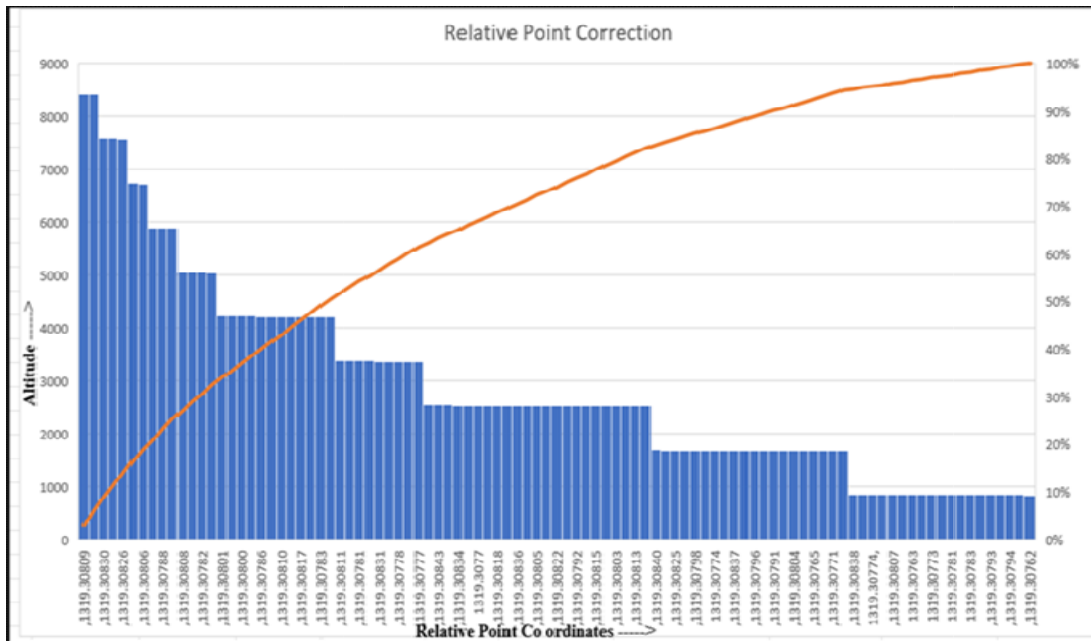


Fig.14RelativePointCorrectionStatistics

It is observed that all the relative point coordinates and the relative points are corrected with the optimization algorithm to achieve accuracy of 2mm at the testing area. The number of distortions observed across the coordinate system is shown in Fig. 15.

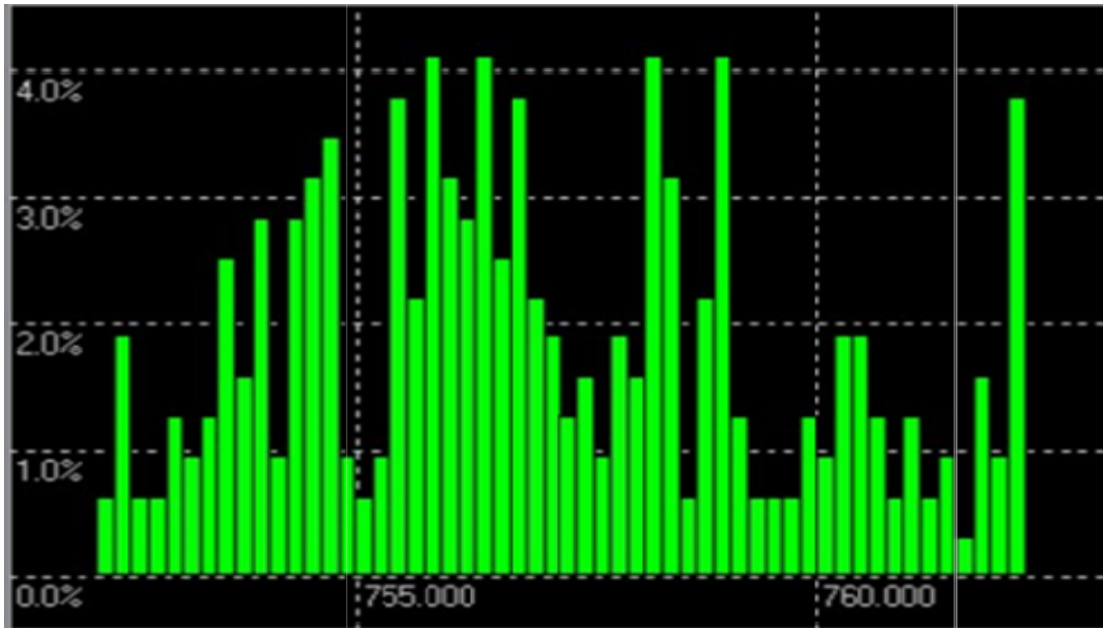


Fig.15 Distortions fetched while parsing the positioning data

4. Conclusions

Many IoT applications used in the field of Precision Agriculture, Land Survey, Surveillance Systems, Autonomous Vehicles, Human to Machine and Machine to Machine interaction requires live tracking of real time data on the ground and under the ground. Using GPS based geofencing method, though it is possible to fetch the nearby location, an accuracy upto centimetre and then upto millimetre level is not possible. Such requirement needs extended technology such as Real Time Kinetics. GNSS based accuracy using Real Time Kinetics is efficient in tracking and managing the IoT assets. In this paper, an optimized real time kinetics framework that gives the result of upto 2mm accuracy is illustrated. The GNSS raw data sampled and is pre-processed further with the finetuning of patch antenna. The research is carried out at Nagamangala area to demonstrate the precision accuracy and 99.9% of accuracy result is achieved.

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Data Availability Statement

None

Conflict of Interest

The author declares that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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