

# Android GNSS Measurements under Spoofing and Interference

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## ABSTRACT

This study explores the behavior of GNSS navigation on a commercial Android smartphone by analyzing raw measurements collected in both static and dynamic conditions. Using Google’s GNSS Logger and MATLAB-based post-processing, we evaluate positioning accuracy, signal quality, and clock stability under normal operation and under spoofed inputs—where false positions and delays are injected into the data. The results highlight the vulnerability of smartphone GNSS to software spoofing and underline diagnostic features that may aid in anomaly detection and robustness improvement.

## 1 INTRODUCTION

Global Navigation Satellite Systems (GNSS) are fundamental to modern positioning services, widely embedded in smartphones and critical infrastructure. Despite their ubiquity, GNSS signals are inherently vulnerable to spoofing—an attack technique where counterfeit signals deceive the receiver into computing false location or timing information. To develop robust mitigation strategies, it is essential to understand how smartphone GNSS systems behave under both normal and manipulated conditions.

This work investigates the GNSS measurement behavior of a consumer-grade Android smartphone across different operating scenarios. We analyze baseline performance by collecting and processing raw GNSS data in two real-world environments: a static session on the rooftop of Monte dei Cappuccini and a dynamic session aboard a tram in urban Turin. In addition to this baseline study, we simulate spoofing by injecting false position inputs into the processing pipeline and, separately, by introducing artificial delays. The remainder of this report is organized as follows.

- **Section 2** describes the experimental setup, including device configuration, data collection procedures, and the processing pipeline.
- **Section 3** presents results and discussion, contrasting static versus dynamic performance, examining spoofed-location impacts, and analyzing delay effects.
- **Section 4** summarizes the key findings and outlines directions for future work.

## 2 METHODS

### 2.1 Devices and Software

We used a Samsung Galaxy A51 with Android 11 for this experiment. GNSS Logger v3.1.0.4 was chosen due to its unrestricted access to raw GNSS measurements, compatibility with newer Android APIs,

and ability to record detailed GNSS data that is suitable for precise analysis. MATLAB R2024b was employed to handle data because it comes with Google’s GNSS toolbox, which accommodates robust analysis and visualization of GNSS measurements and position solutions.

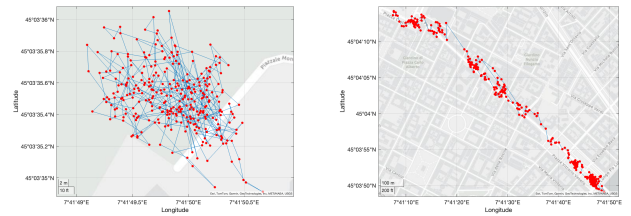
### 2.2 Data Collection Procedure

Two distinct 5-minute GNSS data logging sessions were conducted on 3 May 2025, under cloudy weather conditions, using the GNSS Logger app configured with the following settings enabled:

- **GNSS Location:** Enabled to capture location data.
- **GNSS Measurements:** Enabled to log raw GNSS measurements.
- **Navigation Messages:** Enabled to capture navigation data.
- **GnssStatus:** Enabled to log GNSS status information.
- **Sensors:** Enabled to capture sensor data.

The sessions were designed to capture both static and dynamic GNSS performance, with the following details:

- **Static Scenario:** Performed on the rooftop of Monte dei Cappuccini, Turin, starting at 10:35:20. The device was stationary throughout the entire session, providing baseline measurements.
- **Dynamic Scenario:** Conducted on tram line 15 from Piazza Castello to Piazza Vittorio Veneto, starting at 10:00:21, simulating a typical urban mobility scenario.



(a) Monte dei Cappuccini.

(b) Tram Line 15.

**Figure 1:** Comparison of GNSS data: (a) Static scenario at Monte dei Cappuccini, (b) dynamic scenario along Tram Line 15.

### 2.3 Processing Pipeline

The raw GNSS data from the GNSS Logger served as the input dataset for MATLAB. Processing involved a scripted workflow via `ProcessGnssMeasScript.m`, where the following steps were executed:

1. **Filtering:** Data points with a carrier-to-noise ratio below 25 dB-Hz or satellite elevations below 15° were excluded to improve accuracy.

\*The authors collaborated closely in developing this project.

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60 **2. Measurement Extraction:** Pseudorange and Doppler mea-  
61 surements were computed from GNSS timestamps and satellite  
62 transmission data.  
63 **3. Weighted Least Squares (WLS) Positioning:** Applied to de-  
64 rive precise positioning and clock bias estimates.  
65 **4. Visualization and Comparison:** Output plots from MATLAB,  
66 including pseudorange, pseudorange rates, and position solu-  
67 tions, were generated to facilitate comparative analysis of the  
68 static and dynamic scenarios.

69 Results from this processing pipeline provided insights into the  
70 differences in GNSS performance under static and dynamic condi-  
71 tions.

## 72 2.4 Spoofed-Input Configuration

73 Spoofing scenarios were emulated by introducing artificial vari-  
74 ations to the recorded GNSS data through MATLAB processing.  
75 Specifically, mock positions were assigned by adjusting the param-  
76 eter `spoof.position`, which represents modified latitude, longi-  
77 tude, and altitude coordinates. Additionally, artificial time delays  
78 were tested by adjusting the `spoof.delay` parameter, typically  
79 in milliseconds, to mimic delayed GNSS signal arrival. Such config-  
80 urations facilitated evaluation of the impact of spoofing scenarios  
81 on position estimation reliability and accuracy.

## 82 2.5 Optional Interference Scenario

83 Even a worst-case interference situation was simulated, replicating  
84 conditions in the vicinity of potential interference sources such  
85 as broadcasting antennas or communication areas of high density.  
86 GNSS data were collected near such sources of interference, and  
87 then processed with the same MATLAB procedure. Nominal condi-  
88 tion comparison was established to analyze the impact of external  
89 interference on GNSS observables such as variation in pseudoran-  
90 ge measurements, carrier-to-noise ratio, and positional accuracy  
91 overall.

# 3 RESULTS AND DISCUSSIONS

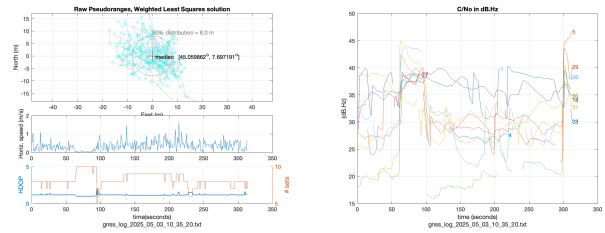
## 92 3.1 Baseline Performance: Static vs. Dynamic

### 93 3.1.1 Static Case: Monte dei Cappuccini

94 The static scenario demonstrated stable GNSS performance. Pseu-  
95 dorange values remained consistent with expected GPS satellite  
96 distances (around  $2 \times 10^7$  m). Position estimates clustered within  
97 an 8 m radius of the median, and horizontal speed remained near  
98 zero, confirming the device was stationary (Figure 2a). C/No val-  
99 ues showed minor fluctuations (Figure 2b), with a few satellites  
100 (e.g., SV 27) experiencing temporary signal degradation. HDOP re-  
101 mainned generally low, indicating good satellite geometry. Clock bias  
102 increased linearly over time (Figure 3), suggesting typical thermal-  
103 induced drift.

### 104 3.1.2 Dynamic Case: Tram Ride

105 The dynamic scenario exhibited greater measurement variability.  
106 Pseudoranges changed more rapidly (Figure 4a), reflecting motion  
107 relative to the satellites. Position estimates were more dispersed,  
108 with northward velocity peaking near 20 m/s (Figure 4b). C/No  
109 values fluctuated significantly due to urban obstructions (Figure 5a),



(a) Static scenario: Position estimates, horizontal speed, HDOP. (b) Static scenario: Carrier-to-noise ratio over time.

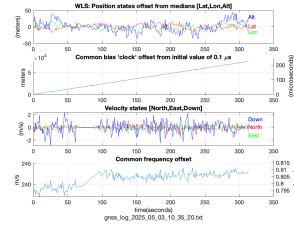
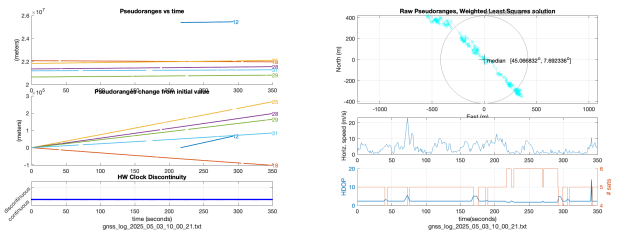
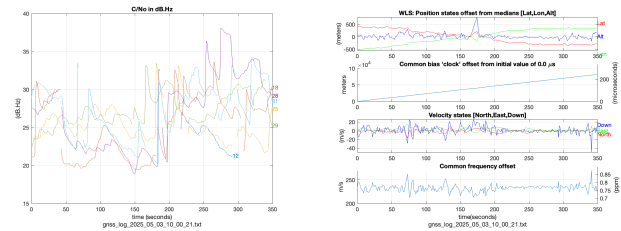


Figure 3: Static scenario: WLS state offsets and clock bias trends.

and some satellites showed weak or intermittent signals (e.g., SV 18). HDOP remained within acceptable bounds, although the number of tracked satellites varied. Clock bias again increased linearly but was more affected by motion-induced dynamics (Figure 5b).



(a) Dynamic scenario: Pseudoranges and variation over time. (b) Dynamic scenario: Position estimates, speed, and HDOP.



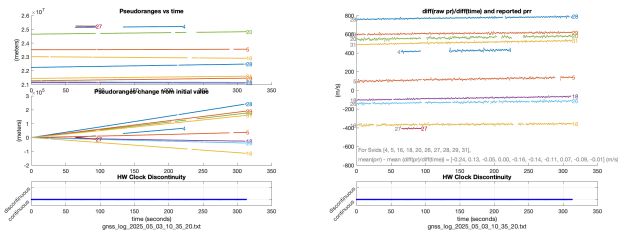
(a) Dynamic scenario: Signal quality (C/No) per satellite. (b) Dynamic scenario: WLS offset and velocity states.

### 3.1.3 Comparison of Static and Dynamic Performance

Key differences emerged when comparing static and dynamic scenarios:

- **Pseudorange Trends:** Static measurements showed flat and stable pseudorange lines with minor steps due to satellite handovers or clock corrections (Figure 6a). In contrast, dynamic

120 measurements exhibited sloped lines, reflecting relative motion  
 121 to satellites.  
 122 • **Doppler Residuals:** In the static case, calculated and reported  
 123 pseudorange rates aligned closely (Figure 6b). The dynamic case  
 124 introduced more noise and discrepancies (Figure 7), reflecting  
 125 movement complexity and potential modeling errors.  
 126 • **Signal Quality (C/No):** The static case showed relatively stable  
 127 C/No values, while the dynamic case had significant fluctuations  
 128 and lower mean values due to urban interference and multipath  
 129 effects.  
 130 • **HDOP and Satellite Geometry:** HDOP remained low and stable  
 131 in the static case, while it was slightly higher and more variable  
 132 in the dynamic scenario due to changing satellite visibility.  
 133 • **Error Statistics:** The static case achieved tighter clustering (~8  
 134 m radius), whereas the dynamic case showed larger offsets, espe-  
 135 cially in the latitude/longitude axes (up to ~500 m).  
 136 • **Velocity and Clock Behavior:** The dynamic case revealed dom-  
 137 inant northward motion and larger variation in velocity states.  
 138 Clock bias trends were similar in both scenarios but more affected  
 139 by noise in the tram ride.  
 140 These differences reflect the impact of movement, signal obstruc-  
 141 tion, and urban dynamics on GNSS performance, confirming that  
 142 dynamic environments introduce more variability and require more  
 143 robust estimation strategies.



(a) Static scenario: Pseudoranges and (b) Static scenario: Reported vs computed PRR.

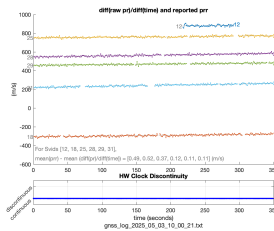


Figure 7: Dynamic scenario: Reported vs computed PRR.

144 **3.2 Impact of Spoofed Position**  
 145 **3.3 Effects of Timing Delays**  
 146 **3.4 Interference Effects**  
 4 **CONCLUSIONS**

## A APPENDIX

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