Of course. Here is a more extensive and detailed single-block input, meticulously crafted for gamma.ai's text-to-presentation feature. It incorporates the deep statistical reasoning, methodological justifications, and specific quantitative results derived from the Jupyter Notebook analysis, framed within a cohesive narrative.

# Title: An In-Depth Validation of SAOCOM L-Band InSAR for Digital Elevation Model Generation: A Land Cover-Stratified Analysis in the Verona Region

# 1. Introduction: Study Objectives, Datasets, and Area of Interest

This presentation details a comprehensive, multi-faceted validation of a Digital Elevation Model (DEM) point cloud derived from the **SAOCOM L-band Interferometric Synthetic Aperture Radar (InSAR)** system. The primary objective is to rigorously quantify the vertical accuracy and spatial coverage of the SAOCOM data, moving beyond simple bulk statistics to understand its performance characteristics under varying real-world conditions. The analysis is structured around three core pillars:

- 1. **Direct Vertical Accuracy Assessment**: Comparing the SAOCOM elevation points against two high-quality, independent reference DEMs.
- Land Cover-Stratified Performance Analysis: Investigating how accuracy and data integrity vary across different terrain and land use types, using the CORINE 2018 classification.
- 3. **Spatial Coverage and Void Analysis**: Quantifying the extent and identifying the underlying causes of data gaps (voids) within the SAOCOM dataset.

The study is geographically focused on a **52.04 km²** area in the Verona region of Italy, with the precise boundary defined by the convex hull of the available SAOCOM data points. This ensures all comparisons are made on a like-for-like basis.

### Core Datasets Utilized:

- SAOCOM: The primary dataset under investigation, an L-band InSAR point cloud in CSV format containing latitude (LAT2), longitude (LON2), relative height (HEIGHT), and a quality metric, coherence (COHER).
- **TINITALY**: The primary reference DEM, offering a high resolution of **10 meters**.
- Copernicus GLO-30: A global DEM with a 30-meter resolution, serving as a secondary reference.
- **CORINE 2018**: The definitive land cover classification dataset for Europe, used to stratify the analysis.
- **Sentinel-2**: High-resolution multispectral imagery providing a true-color visual backdrop for analysis and contextualization.

To facilitate precise spatial analysis, all datasets were geodetically standardized to a common **10-meter grid** projected in **UTM Zone 32N (EPSG:32632)**.

# 2. Methodological Framework: Data Processing and Statistical Integrity

A robust and repeatable processing workflow was established to ensure the analytical integrity of the results. The choice of parameters and methods was driven by established geospatial best practices and statistical principles.

# **SAOCOM Point Cloud Filtering:**

- Coherence Filtering (y ≥ 0.3): Coherence is a critical measure of InSAR quality, representing the stability of the radar signal's phase between the two satellite passes. A threshold of 0.3 was selected as a standard, moderately conservative value. This step is crucial for filtering out statistically unreliable points, or "decorrelated" pixels, which are often caused by temporal changes in features like dense vegetation canopies, water bodies, or agricultural fields between acquisitions. This removes significant noise from the dataset.
- **k-NN Isolated Point Removal**: A k-Nearest Neighbors statistical algorithm (k=5, distance threshold=100m) was implemented. This process identifies and removes spatial outliers—points that are geographically distant from their neighbors. Such points are statistically improbable and are assumed to be measurement or processing errors rather than true topographic features. This step ensures the point cloud represents a continuous surface.

### Raster Data Standardization:

Resampling Methodologies: The choice of resampling algorithm was critical. For the
continuous elevation data (TINITALY, Copernicus), Cubic Convolution was used. This
method considers a 4x4 neighborhood of pixels, resulting in a smoother, more
topographically realistic surface. For the categorical CORINE land cover data, Nearest
Neighbor resampling was essential. This method assigns the output pixel the value of
the single closest input pixel, critically preventing the creation of new, invalid land
cover class codes that would arise from averaging.

# 3. Height Calibration: From Relative to Absolute Elevation

The native output of the InSAR processing chain is a relative height value. To enable a direct comparison with the absolute reference DEMs, a vertical calibration was performed to correct for a systemic offset.

• Calibration Approach: A constant vertical offset was calculated and applied. This is a standard procedure to align the floating InSAR heights to a recognized geodetic datum.

- Stable Point Selection: To derive a highly reliable offset, the calculation was based exclusively on the most stable points in the scene. These were identified using a stringent coherence threshold of y ≥ 0.8, ensuring only pixels with minimal phase noise were included.
- Robust Offset Calculation: The offset was defined as the median of the differences between the reference DEM and the relative SAOCOM heights (offset = median(Reference\_Height SAOCOM\_Relative)). The median was deliberately chosen over the mean because it is a robust statistical estimator, meaning it is not unduly influenced by any remaining extreme outliers. This ensures the calculated offset represents the true central tendency of the vertical displacement.

### Calibration Results:

- TINITALY-based Offset: 4.308 meters (derived from 46,920 high-coherence stable points).
- Copernicus-based Offset: 4.761 meters (derived from 46,939 high-coherence stable points).

# 4. Results: Overall Accuracy and Spatial Coverage Analysis

A baseline comparison between the two reference DEMs (TINITALY vs. Copernicus) established a high level of agreement, with an **RMSE of 4.68 m** and a robust **NMAD of 2.18 m**. This provides a benchmark for evaluating the SAOCOM data's performance.

# **SAOCOM Vertical Accuracy:**

The calibrated SAOCOM point cloud, when compared across the entire study area, yielded the following key accuracy metrics:

- vs. TINITALY: RMSE = 15.25 m; NMAD = 5.24 m
- vs. Copernicus: RMSE = 15.40 m; NMAD = 4.79 m

While the RMSE indicates significant error magnitudes, the more robust NMAD suggests that the bulk of the points have a more reasonable error profile, with the RMSE being inflated by outliers.

# Spatial Coverage Analysis:

The most striking result from the initial analysis was the sparse spatial coverage of the SAOCOM data. A grid-based analysis revealed:

Total 10m Cells in Study Area: 520,380

SAOCOM Occupied Cells: 67,613

Void Cells (No Data): 452,767

• This results in a **void percentage of 87.0**%, indicating that the SAOCOM DEM failed to generate elevation data for the vast majority of the study area. A boolean void mask was generated as a GeoTIFF to facilitate further analysis.

# 5. Diagnostic Deep Dive: Land Cover Stratified Analysis

To understand the root causes behind both the vertical errors and the extensive data voids, the analysis was stratified by the 10 unique CORINE land cover classes present in the study area.

Height Residuals by Land Cover:

Violin plot analysis, which visualizes the probability distribution of errors, revealed dramatic performance differences between land cover types.

- **Stable Surfaces**: Classes like 'Discontinuous urban fabric' (Code 112) showed tight, narrow error distributions with a low NMAD, indicating reliable performance.
- Problematic Surfaces: In contrast, classes such as 'Vineyards' (Code 221) and 'Fruit trees and berry plantations' (Code 222) exhibited much wider, more dispersed error distributions and significantly higher NMAD values. This provides clear statistical evidence that the SAOCOM DEM's accuracy degrades substantially over vegetated and complex agricultural areas.

Void Analysis by Land Cover:

This analysis definitively answered why the coverage was so poor. By calculating the percentage of voids within each land cover class, a clear pattern emerged.

- Highest Void Percentages: Land cover classes with complex vertical structures, primarily vegetation, were almost entirely data voids. For example, 'Vineyards' (99.7% void), 'Non-irrigated arable land' (98.6% void), and 'Complex cultivation patterns' (98.4% void) had almost no SAOCOM data coverage.
- Largest Void Contributors: The same classes were also responsible for the largest absolute void areas, confirming that the failure of the L-band InSAR to penetrate or receive a stable signal return from these specific land cover types is the primary driver of the dataset's poor spatial coverage.

# 6. Comprehensive Visualization Suite & Project Outputs

An extensive suite of visualizations was created to explore and communicate these findings effectively.

• **Multi-Panel Maps**: An 8-panel figure was generated to cross-compare the reference DEMs, showing elevation, difference maps, and directional comparisons. A similar 6-

panel figure was created to visualize the spatial distribution of SAOCOM errors against both TINITALY and Copernicus, highlighting areas of positive and negative bias.

- Statistical Plots: Scatter plots with 1:1 reference lines provided a clear view of bias and correlation. A series of violin plots were crucial for visualizing the distribution of errors, both overall, binned by coherence, and, most importantly, stratified by land cover class.
- Void and Coverage Maps: Custom maps were generated to visualize the relationship between data voids and the underlying landscape. These included maps showing land cover only within void zones and detailed, per-class maps illustrating the exact boundaries of coverage versus void areas, overlaid on Sentinel-2 imagery.

### Final Deliverables:

The analysis produced a rich set of data and visual outputs, saved to a dedicated results/directory. These include:

- **Processed GeoTIFFs**: Resampled and masked versions of all reference DEMs, the remapped CORINE grid, and the final SAOCOM void mask.
- **High-Resolution Figures**: All generated plots and maps were saved as PNG files, including comparison figures, void analysis charts, and 44 individual land cover and coverage/void maps.