



# OPTIMIZING PLANETSCOPE-BASED SATELLITE-DERIVED BATHYMETRY: A SINGLE-BAND APPROACH WITH 3D SPATIAL AND STATISTICAL FILTERING



Keywords: Satellite-Derived Bathymetry (SDB), PlanetScope, Statistical Filtering, 3D Geospatial Filtering, Bathymetry Accuracy

## Authors

Maharani R.<sup>1</sup>, Setyawan F.O.<sup>1</sup>, Sarono<sup>2\*</sup>



## Affiliations

<sup>1</sup>Marine Science Study Program, Faculty of Fisheries and Marine Science, Brawijaya University, Malang

<sup>2</sup>Doctoral Program in Geography, Faculty of Geography, Gadjah Mada University, Yogyakarta

## Introduction

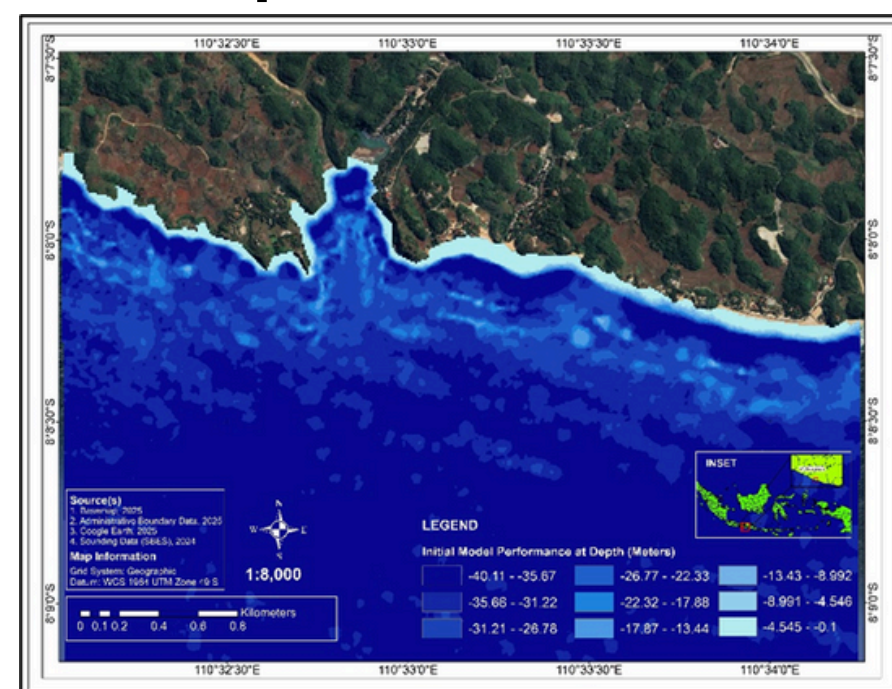


Bathymetric information is crucial for navigation safety, habitat mapping, and coastal management (Calder & Mayer, 2003; IHO, 2022). Traditional hydrographic surveys such as SBES and MBES provide high accuracy but are costly, time-consuming, and limited in spatial coverage. Satellite-Derived Bathymetry (SDB) offers a cheaper and more efficient alternative. However, the single-band approach often shows low accuracy (Stumpf et al., 2003; Thomas et al., 2021), which makes optimization necessary.

## Study Area

The study was conducted in the coastal waters of Baron Beach, Gunungkidul, Yogyakarta (110°32'–110°33' E; 8°9'–8°10' S). Baron is an active tourism and fisheries zone, which increases the demand for accurate and up-to-date bathymetric data.

## Map Initial Model



## Objective



This study aims to:

- Optimize PlanetScope-based single-band SDB using Cloth Simulation Filter (CSF).
- Compare different interpolation methods (IDW, Kriging, Natural Neighbour, Spline).
- Validate the accuracy using SBES reference data reduced to MSL.
- Assess the model against the IHO (2022) Order 1 standard.

## Methodology



01

Pre-processing: radiometric correction  
→ TOA reflectance → NDWI masking  
→ tidal correction

02

Regression: single-band test on red, green, and blue → green chosen ( $R^2 = 0.211$ ;  $RMSE = 2.080$  m)

03

Filtering: CSF applied to reduce noise in depth point cloud.

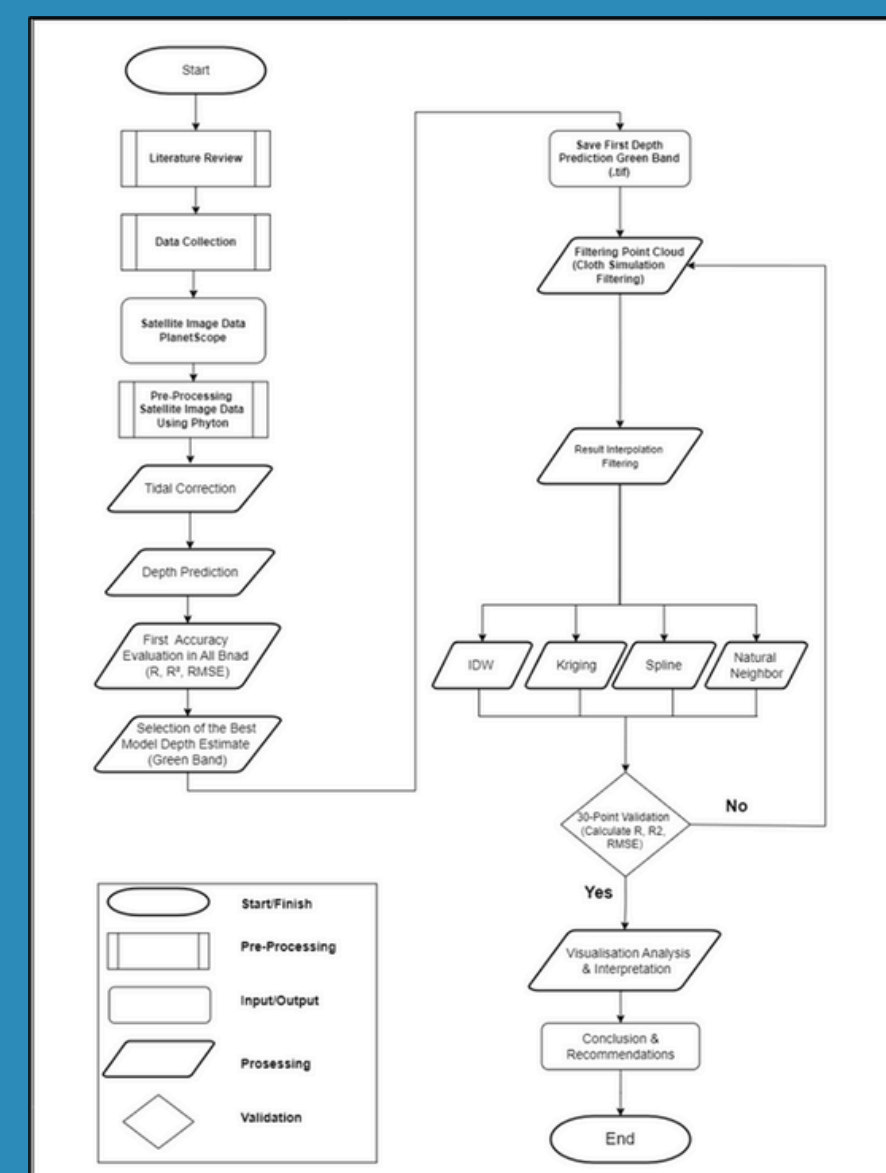
04

Interpolation: IDW, Kriging, Natural Neighbour, and Spline.

05

Validation: 30 independent SBES points using R,  $R^2$ , RMSE.

$$RMSE = \sqrt{\frac{\sum_{i=1}^n (A_i - F_i)^2}{n}}$$



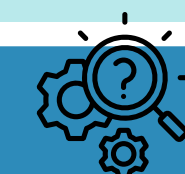
## Results

The best result was achieved by CSF + IDW ( $R^2 = 0.453$ ;  $RMSE = 1.619$  m), showing a significant improvement compared to the initial green band model ( $R^2 = 0.211$ ;  $RMSE = 2.080$  m)

Table 1: Statistical Performance of Initial and Optimised Models

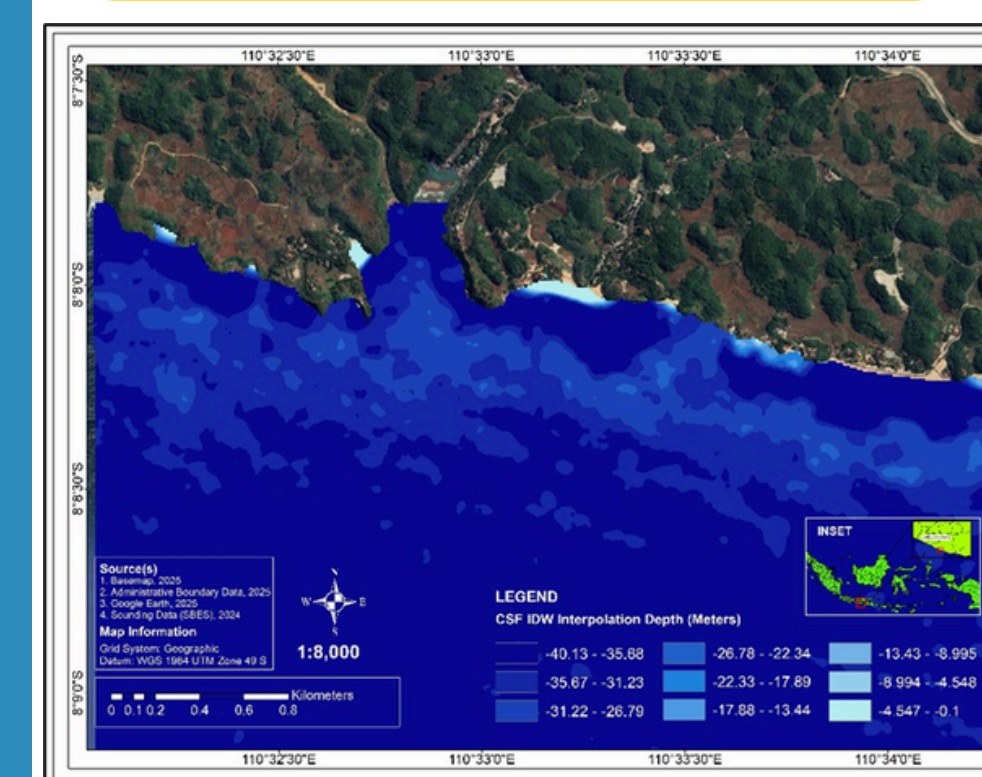
Model	R	R <sup>2</sup>	RMSE (m)
Initial Green Band	0.46	0.211	2.08
CSF + IDW	0.673	0.453	1.619
CSF + Kriging	0.67	0.449	1.626
CSF + Natural	0.642	0.412	1.674
CSF + Spline	0.669	0.448	1.625

## Analysis

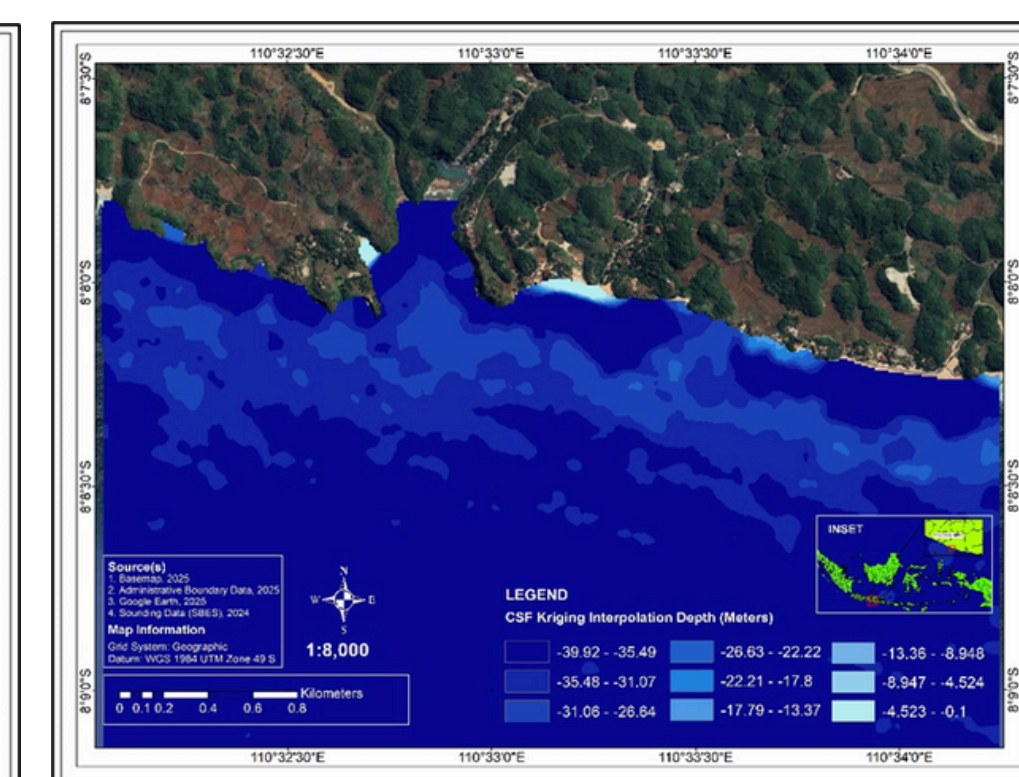


The application of CSF was effective in filtering out noise caused by turbidity, surface ripples, and seabed heterogeneity, while preserving key morphological features (Zhang et al., 2016). IDW performed best due to its suitability for dense point distributions, such as the SBES data at Baron Beach. Kriging provided detailed results but was highly sensitive to uneven data spacing. Spline tended to oversmooth, and Natural Neighbour lacked detail. The best result (CSF + IDW) reached IHO Order 1, making it reliable for coastal mapping, though not yet sufficient for high-precision navigation (IHO, 2022).

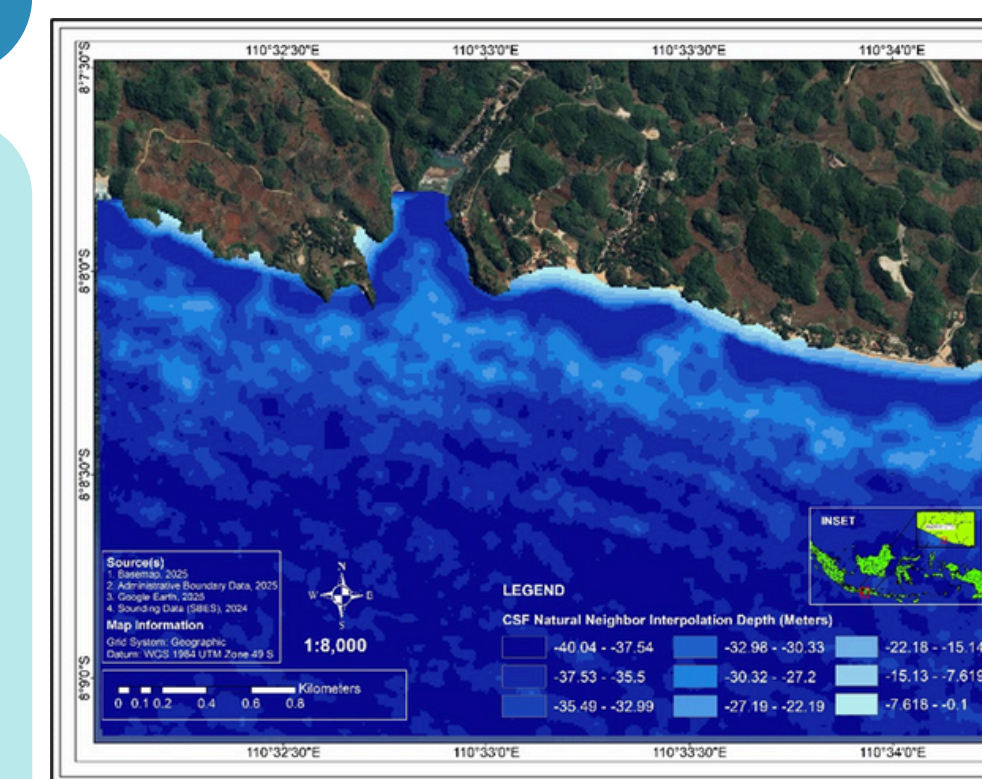
## CSF + IDW Interpolation Map



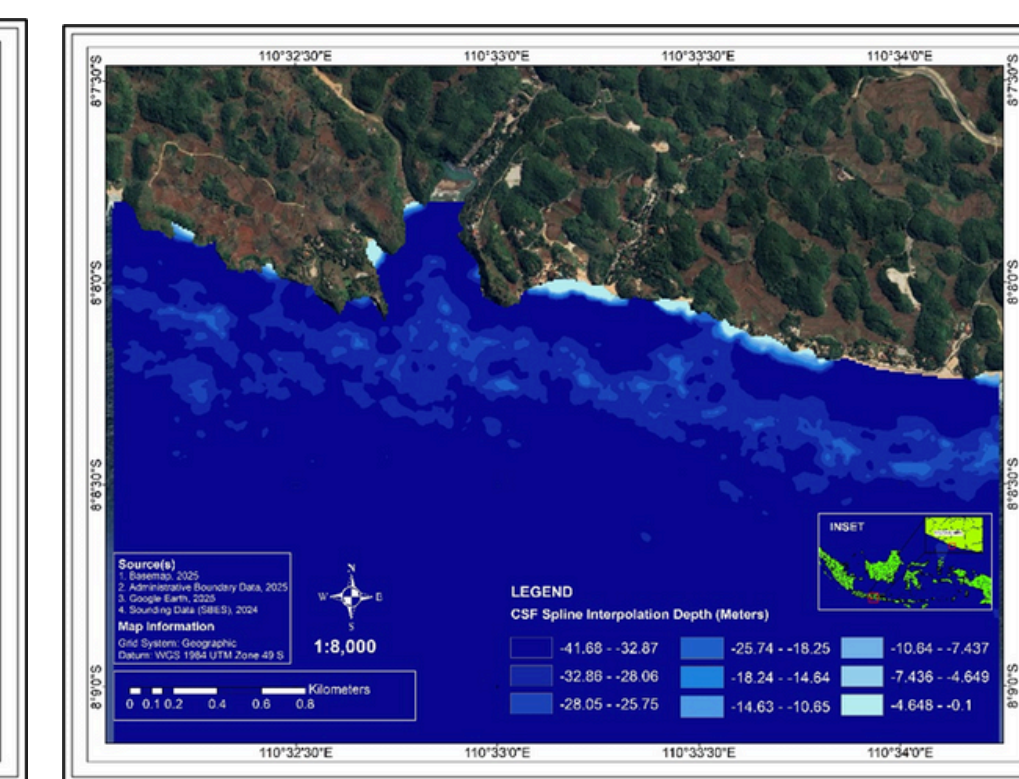
## CSF + Kriging Interpolation Map



## CSF + NN Interpolation Map



## CSF + Spline Interpolation Map



## Conclusion



This study demonstrates that integrating single-band regression, CSF filtering, and spatial interpolation can significantly improve the accuracy of PlanetScope-based SDB. The CSF + IDW model reduced RMSE by 0.461 m and increased  $R^2$  by 0.242, showing substantial improvement over the initial regression. This method is efficient and cost-effective for shallow tropical waters. Future research should explore multi-band approaches, machine learning algorithms, and applications in other coastal regions

## Key Sources

- Calder, B. R., & Mayer, L. A. (2003). Automatic processing of high-rate, high-density multibeam echosounder data. Deep-Sea Research Part II: Topical Studies in Oceanography, 50(6–7), 1047–1074. <https://doi.org/doi:10.1029/2002GC000486>
- IHO. (2022). IHO Standards for Hydrographic Surveys (Edition 6.1.0).
- Thomas, N., Pertiwi, A. P., Tragano, D., Lagomasino, D., Poursanidis, D., Moreno, S., & Fatoyinbo, L. (2021). Space-borne, cloud-native Satellite-Derived Bathymetry (SDB) models using ICESat-2 and Sentinel-2. Geophysical Research Letters, 48(11). <https://doi.org/10.1029/2020GL092170>
- Zhang, W., Qi, J., Wan, P., Wang, H., Xie, D., Wang, X., & Yan, G. (2016). An Easy-to-Use Airborne LiDAR Data Filtering Method Based on Cloth Simulation. Remote Sensing, 8(6), 501. <https://doi.org/10.3390/rs8060501>

## Acknowledgements

We would like to thank the Faculty of Marine Science, Universitas Brawijaya and the Faculty of Geography, Universitas Gadjah Mada for their support during this research. Special thanks to our supervisors for their guidance and valuable input, and to the ACRS 2025 committee for providing the opportunity to present this research.