

## **I. The Remote Sensing Imperative for Marine Plastic Monitoring**

### **1.1. Quantifying the Crisis: The Global Scale of Marine Plastic Pollution**

The contamination of marine ecosystems by plastic waste represents one of the most pressing environmental challenges of the modern era.[1] Global plastic production has grown exponentially, and with it, the leakage of waste into the oceans. In 2016 alone, an estimated 11 million metric tons of plastic entered the marine environment, a figure that is projected to double or even triple by 2040 without significant intervention.[1] These materials are exceptionally persistent, with some common polymers like Low-Density Polyethylene (LDPE) and High-Density Polyethylene (HDPE) estimated to endure for 3 to 450 years in the marine environment.[2] This durability, combined with continuous input, has led to the ubiquitous presence of plastic debris across all oceanic compartments, from surface waters to deep-sea sediments.[3]

This pollution is not monolithic. It is broadly categorized by size into macroplastics (particles larger than 5 mm) and microplastics (particles smaller than 5 mm).[2] While microplastics pose a pervasive threat through bioaccumulation in the food chain, they are largely derived from the physical, chemical, and photodegradation of larger macroplastic items.[2] Consequently, the focus of current spaceborne remote sensing technologies is necessarily on the detection of macroplastics, as individual microplastic particles are far too small to be resolved by any existing or planned satellite sensor.[3]

A fundamental constraint that defines the entire field of satellite-based plastic monitoring is the vertical distribution of plastic in the ocean. It is estimated that only about 1% of the total plastic mass in the oceans is actively floating on or near the surface.[5] The remaining 99% is suspended within the water column or has sunk to the seabed.[5] This critical fact reframes the role of satellite remote sensing. It is not a tool for creating a complete inventory of all marine plastic. Rather, it is a specialized instrument for monitoring the surface layer—the crucial interface where plastic enters the ocean, is transported by currents, and interacts with marine life before eventually sinking. The primary value of satellite data, therefore, lies in identifying surface accumulation hotspots, tracking the movement of debris to understand and validate ocean transport models, and providing synoptic data to guide targeted clean-up operations.[5]

### **1.2. Limitations of Traditional Monitoring and the Satellite Advantage**

For decades, the assessment of marine plastic pollution has relied on traditional, in-situ methods such as deploying nets from research vessels (net tows), visual surveys from ships, and beach combing.[1] While essential for providing direct physical samples and detailed local data, these methods are inherently limited. They are logistically complex, expensive, and time-consuming, and they provide only sparse, localized snapshots of a problem that is global

in scale and highly dynamic.[9] The sheer vastness of the oceans makes it impossible to achieve comprehensive, synoptic coverage through these techniques alone, leaving significant gaps in our understanding of the global distribution and transport of plastic debris.[1]

The advent of Earth Observation (EO) from space offers a transformative approach to supplement and scale these efforts. Satellite remote sensing provides a non-intrusive, systematic, and cost-effective means to monitor vast oceanic regions on a regular basis.[3] This capability allows for the consistent mapping of debris patterns, the identification of persistent accumulation zones (or "hotspots"), and the potential tracking of large pollution events from source to sea.[10] By providing this "big picture" view, satellite data can guide the more efficient deployment of limited resources, such as clean-up vessels, and furnish policymakers with the large-scale evidence needed to implement and monitor the effectiveness of mitigation strategies.[3]

### **1.3. Defining the Detectable Target: From Individual Items to Proxies**

Directly detecting individual plastic items like bottles or bags from space is, with few exceptions for very high-resolution commercial satellites, currently infeasible. The spatial resolution of freely available, wide-swath satellites like the Copernicus Sentinel-2 mission (10 meters at best) is too coarse to resolve such small objects.[3] Therefore, the operational target for satellite remote sensing is not the individual piece of litter but large aggregations of floating debris.[3]

Oceanographic processes, such as winds, waves, and currents, naturally concentrate floating materials into larger patches or elongated slicks.[11] Of particular interest are features known as "windrows," which are linear accumulations of debris that form along surface convergence zones.[3] These windrows can be kilometers long and several meters wide, making them sufficiently large to be detected by medium-resolution sensors like Sentinel-2.[3] Field campaigns have demonstrated that these features often contain a high concentration of macroplastic litter, allowing them to serve as a reliable proxy for its presence.[3]

This shifts the remote sensing problem from one of pure material identification to one of feature detection. The primary challenge, however, is that these aggregations are rarely composed of pure plastic. They are typically a heterogeneous mixture of anthropogenic litter and natural materials with similar physical properties, such as sargassum, driftwood, sea foam, and other organic matter.[3] Consequently, the central technical task for any satellite-based detection algorithm is to not only detect these floating features but also to accurately discriminate the anthropogenic components from the natural ones.[3]

## **II. Optical Detection with Sentinel-2: Spectral Signatures and Indices**

### **2.1. The Sentinel-2 Multi-Spectral Instrument (MSI): A Detailed Examination**

The Copernicus Sentinel-2 mission, comprising the twin satellites Sentinel-2A and

Sentinel-2B, is the workhorse for optical remote sensing of marine debris. Its Multi-Spectral Instrument (MSI) is a passive sensor that measures reflected sunlight across 13 spectral bands, providing a unique combination of wide coverage (290 km swath), high spatial resolution (10 m, 20 m, and 60 m), and frequent revisit times (5 days at the equator).[14] This combination makes it uniquely suited for systematic monitoring of coastal and inland waters.[15] While designed primarily for land monitoring, its spectral bands in the visible (VIS), near-infrared (NIR), and short-wave infrared (SWIR) ranges contain critical information for detecting floating plastics.[5] A detailed understanding of these bands is foundational to developing effective detection algorithms.

**Table 1: Sentinel-2 MSI Spectral Bands and Their Relevance for Marine Plastic Detection**

Band	Name	Wavelength (nm)	Resolution (m)	Relevance for Marine Plastic Detection
B1	Coastal Aerosol	443	60	Primarily for atmospheric correction and bathymetric studies. Helps estimate suspended sediment, which can be a source of confusion.[16]
B2	Blue	490	10	High water penetration but also high atmospheric scattering. Water appears dark blue. Useful for distinguishing clear vs. turbid water but less effective for

				direct plastic detection.[16]
B3	Green	560	10	Reflectance peak for phytoplankton/algae. Key for differentiating plastic from floating vegetation (sargassum).[13]
B4	Red	665	10	Strong chlorophyll absorption band, making healthy vegetation appear dark. Water also absorbs strongly. A key component of NDVI and PI indices.[17]
B5	Red Edge 1	705	20	Located on the slope of vegetation's reflectance increase. Highly sensitive to chlorophyll content and health, aiding in precise vegetation discrimination. [17]

B6	Red Edge 2	740	20	Further along the red edge slope. Used in some advanced indices (e.g., the original FDI formulation) to refine the distinction from natural materials.[17]
B7	Red Edge 3	783	20	Continues to characterize the vegetation reflectance profile, enhancing the ability to separate plastics from various types of floating biomass.[17]
B8	NIR	842	10	<b>CRITICAL BAND.</b> Water strongly absorbs NIR radiation, appearing very dark. Plastics and vegetation strongly reflect it, creating high contrast. This is the fundamental principle for most optical detection

				methods.[11]
B8A	Narrow NIR	865	20	Similar to B8 but narrower. Used in various indices (like the Moisture Index) and provides a refined NIR signal for analysis.[16]
B9	Water Vapour	945	60	Used for atmospheric correction by measuring water vapor content. Not used for direct surface detection.[19]
B10	SWIR - Cirrus	1375	60	Used for cirrus cloud detection. Not used for direct surface detection.[19]
B11	SWIR 1	1610	20	<b>IMPORTANT BAND.</b> Water absorption is very high. Plastics have distinct hydrocarbon absorption features in the SWIR range. This band is key for

				material differentiation and is used in indices like FDI.[5]
B12	SWIR 2	2190	20	<b>IMPORTANT BAND.</b> Similar to B11, provides another spectral data point in the SWIR region where plastics have unique signatures, helping to distinguish them from other materials.[9]

## 2.2. The Spectral Fingerprint of Marine Plastics

The fundamental principle underpinning the optical detection of marine plastic is the stark contrast in its spectral response compared to that of seawater, particularly in the infrared portion of the spectrum.[11] Water is a strong absorber of radiation in the near-infrared (NIR) and short-wave infrared (SWIR) wavelengths. As a result, clear water bodies appear very dark in these bands. In contrast, floating plastics—being composed of hydrocarbon polymers—and other materials like vegetation tend to reflect NIR light strongly.[11] This creates a significant positive signal, or brightness, over a dark background, which is the primary feature exploited by detection algorithms.[11]

Furthermore, plastics exhibit unique absorption features in the SWIR range related to their chemical composition.[1] While Sentinel-2's SWIR bands (B11 and B12) are broad, they are positioned in regions where these hydrocarbon absorption features can be detected, providing a means to distinguish plastics from other materials that might also be bright in the NIR, such as sea foam or certain types of algae.[9]

However, this theoretical spectral signature is often compromised in the real world. Two major factors degrade the signal: biofouling and submersion. Biofouling occurs when marine organisms like algae colonize the surface of the plastic, altering its color and spectral

properties, particularly in the visible (RGB) bands.[7] Submersion, even by a few centimeters, dramatically attenuates the signal across all wavelengths, with the most significant effect in the NIR and SWIR regions where water absorption is highest.[7] These environmental factors represent a significant challenge, as they can weaken or completely mask the characteristic plastic signature, making detection more difficult.[7]

### 2.3. Leveraging Spectral Indices for Plastic Detection

To enhance the subtle spectral signals of plastic and automate detection, researchers have developed a range of spectral indices. These are simple mathematical formulas that combine the reflectance values from two or more spectral bands to highlight a specific feature of interest while suppressing background noise. For marine plastic detection, indices are a foundational analysis tool, often used as a first-pass classifier or as input features for more advanced machine learning models.[6]

- **Floating Debris Index (FDI):** This index is specifically designed to detect floating materials that are spectrally distinct from water. The FDI calculates a baseline NIR reflectance by linearly interpolating between a Red-Edge band (e.g., B6) and a SWIR band (e.g., B11). It then subtracts this baseline from the measured NIR reflectance (B8).[18] For water, the measured and interpolated values are very similar, resulting in an FDI value near zero. For floating materials like plastic or vegetation, which have a reflectance peak in the NIR, the measured value is significantly higher than the baseline, yielding a high positive FDI value.[11] The formula is expressed as:  
$$FDI = R_{NIR} - R'_{NIR}$$
where  $R'_{NIR}$  is the interpolated reflectance at the NIR wavelength.
- **Plastic Index (PI):** The PI is another index developed to specifically target plastics. Several formulations exist, but a common version utilizes the strong reflectance of plastics in the NIR band relative to the Red band, similar to the logic of vegetation indices.[2] Positive PI values can indicate the presence of floating plastics and associated debris.[6]
- **Normalized Difference Vegetation Index (NDVI):** This is the most widely used index for assessing vegetation health and density. It is calculated as:  
$$NDVI = (R_{NIR} - R_{Red}) / (R_{NIR} + R_{Red})$$
In the context of marine debris, NDVI is not used for detection but for discrimination. Since floating seaweed and sargassum are common sources of false positives, calculating the NDVI for all detected floating objects allows for the separation of vegetation (high NDVI) from potential plastics (lower or negative NDVI).[6]

In practice, these indices are rarely used in isolation. An effective workflow often involves applying a threshold to the FDI or PI to identify all potential floating material pixels, and then applying a second threshold to the NDVI to mask out pixels that are likely to be vegetation, thereby refining the final map of potential plastic debris.[6]

### 2.4. Challenges and Mitigations



Despite the power of spectral analysis, optical detection with Sentinel-2 faces several significant hurdles that must be addressed in any operational system.

- **The Mixed Pixel Problem:** The 10m x 10m area covered by a single Sentinel-2 pixel is often larger than individual plastic items. As a result, the sensor records a spectrally mixed signal, averaged from all materials within the pixel (e.g., 10% plastic, 90% water).[11] This dilution of the plastic signature makes detection extremely challenging, especially for sparse debris fields. Studies have shown that if water constitutes more than 50-70% of a pixel, the reflectance from the plastic is significantly reduced.[11] Sub-pixel detection techniques, such as the FDI, are designed to address this by looking for subtle spectral shape changes rather than absolute brightness, but the problem remains a primary limiting factor.[4]
- **Atmospheric Interference:** The Earth's atmosphere scatters and absorbs sunlight, altering the signal that reaches the satellite sensor. Cloud cover is the most obvious and insurmountable obstacle, rendering optical imagery unusable.[11] For cloud-free images, atmospheric correction is a critical preprocessing step. Algorithms like Sen2Cor are used to convert the raw Top-of-Atmosphere (TOA) radiance measured by the satellite into Bottom-of-Atmosphere (BOA) reflectance, which represents the true spectral properties of the surface. This correction is essential for accurate and consistent spectral analysis across different times and locations.[2]
- **Sun Glint:** Sun glint is the specular reflection of direct sunlight from the water surface into the sensor's field of view. This phenomenon can create extremely bright pixels that can be easily mistaken for floating debris, leading to a high rate of false positives.[20] Mitigating sun glint often involves geometric considerations (avoiding imagery where the sun-sensor geometry is conducive to glint) or using algorithms that can identify and mask glinted areas based on their spectral properties, though this remains a challenge.

### III. All-Weather Surveillance with Sentinel-1: The Role of Synthetic Aperture Radar

#### 3.1. Fundamentals of SAR for Ocean Surface Analysis

While optical sensors like Sentinel-2 provide rich spectral information, their effectiveness is fundamentally limited by daylight and clear skies. To achieve persistent, all-weather monitoring, a different technology is required: Synthetic Aperture Radar (SAR). The Copernicus Sentinel-1 mission is an active microwave remote sensing system that provides C-band SAR imagery.[21] As an active sensor, it illuminates the Earth's surface with its own microwave pulses and records the backscattered signal. This allows it to acquire high-resolution images regardless of cloud cover, rain, or time of day, making it an invaluable tool for operational marine surveillance.[12]

The core principle of SAR imaging for marine applications is its sensitivity to the physical roughness of the sea surface.[22] A calm, smooth water surface acts like a specular reflector (a mirror), reflecting the incoming radar pulse away from the satellite. This results in a very low backscatter signal, causing calm water to appear dark in a SAR image. Conversely, any feature

that increases the surface roughness—such as wind-driven waves, ship wakes, or floating objects—will scatter the radar energy in multiple directions, including back towards the sensor. These rougher areas appear as brighter features against the dark water background.[22]

### 3.2. Methodologies for Plastic Detection

The detection of marine plastic with SAR is an indirect process based on identifying anomalies in sea surface roughness. The hypothesis is that aggregations of floating debris disrupt the smooth surface of the water, increasing the local surface roughness and thus the radar backscatter, making them visible as bright targets.[22]

- **Backscatter Analysis:** The simplest method involves visually or algorithmically searching for bright pixels or clusters of pixels that stand out against the dark, low-backscatter signature of the surrounding water. This is fundamentally an anomaly detection task. This approach has been shown to be feasible for detecting large "plastic islands" in river systems, where dense accumulations create a strong and distinct backscatter signal.[8]
- **Polarization:** Sentinel-1 transmits and receives microwave signals with a specific polarization (the orientation of the electric field). It typically operates in a dual-polarization mode, providing data in VV (vertical transmit, vertical receive) and VH (vertical transmit, horizontal receive) channels.[21] The way a surface scatters these different polarizations provides additional information. For marine debris, research has yielded varied results. Some studies focusing on open-water debris found that the co-polarized VV channel was more effective at highlighting potential floating material.[6] However, other research, particularly on dense plastic accumulations trapped by dams in rivers, concluded that the cross-polarized VH channel provided superior detection performance and better discrimination from the water background.[8] This suggests that the optimal polarization may depend on the specific environmental conditions, the density of the debris, and the presence of confounding factors like wind-induced waves.
- **Data Products and Advanced Techniques:** Sentinel-1 provides two main data products. Ground Range Detected (GRD) products are pre-processed to provide the magnitude or intensity of the backscatter and are the most commonly used for simple backscatter analysis.[6] Single Look Complex (SLC) products, however, are less processed and retain both the magnitude and phase information of the radar signal.[22] This phase information is crucial for coherent techniques like interferometry and change detection. Studies have shown that change detection algorithms, which compare two SLC images of the same area taken at different times, can be highly effective at detecting the appearance of new features like plastic accumulations, with detectors utilizing coherent SLC data outperforming those that use only incoherent GRD data.[8]

### 3.3. Limitations and Indirect Detection

While SAR's all-weather capability is a major advantage, it comes with significant limitations that make it a challenging tool for plastic detection.

- **Speckle Noise:** SAR images are inherently affected by speckle, a granular

"salt-and-pepper" noise that arises from the constructive and destructive interference of the coherent radar waves.[21] This noise can obscure small or low-contrast targets and necessitates the application of specialized speckle filters (e.g., the Lee or Refined Lee filter) as a mandatory preprocessing step to improve image interpretability.[21]

- **Ambiguity of the Signal:** The most fundamental limitation of SAR is that it does not measure material composition. It measures physical surface roughness. A bright spot in a SAR image is simply an area of high surface roughness; it is not definitively plastic.[3] Numerous other phenomena can create similar signatures, including localized wind gusts (wind slicks), rain cells, ship wakes, dense patches of sargassum, and natural biogenic slicks (biofilms) that dampen capillary waves.[3] This ambiguity makes it extremely difficult to confidently classify a SAR detection as plastic without ancillary information.
- **Microplastics and Low Concentrations:** The SAR signal is generated by the interaction of microwaves with surface roughness at the scale of the radar's wavelength (a few centimeters for C-band). While dense aggregations of macroplastics can create sufficient roughness, sparse debris fields may not. Recent studies have demonstrated that only very high concentrations of microplastics lead to a sufficient dampening of capillary waves to be detectable, making SAR generally unsuitable for monitoring typical environmental concentrations of microplastics.[3]

The distinct strengths and weaknesses of Sentinel-1 and Sentinel-2 position them as complementary, rather than competing, systems. SAR provides a persistent, all-weather surveillance capability to flag potential areas of interest, while optical sensors provide the spectral information needed for more confident material discrimination when conditions permit.

**Table 2: Comparative Analysis of Sentinel-1 and Sentinel-2 for Marine Plastic Detection**

Feature	Sentinel-1 (SAR)	Sentinel-2 (Optical)
Sensor Type	Active Microwave (C-band Radar)	Passive Optical (Multispectral Imager)
Key Strength	All-weather, day/night operation. Unaffected by clouds.	Rich spectral information for material discrimination.
Key Limitation	Ambiguous signal (detects roughness, not material). Prone to speckle noise.	Ineffective in cloudy conditions or at night. Signal affected by atmosphere/glint.
Primary Detection Mechanism	Anomaly detection based on increased sea surface	Spectral signature analysis, primarily high reflectance

	roughness caused by debris aggregations.	of plastic in NIR/SWIR bands where water absorbs.
<b>Spatial Resolution</b>	~10m (IW mode)	10m, 20m, 60m (depending on band)
<b>Data Products</b>	GRD (Intensity), SLC (Phase & Intensity)	Level-1C (TOA), Level-2A (BOA)
<b>Confidence in Plastic ID</b>	Low to Medium (Indirect evidence)	Medium to High (Direct spectral evidence)

Ultimately, Sentinel-1's role is best understood as that of a detection and cueing system. It can autonomously and reliably scan vast areas to flag anomalies. A bright spot on a SAR image should be treated as a high-priority "Area of Interest" (AOI), triggering a request for follow-up analysis. This could involve tasking a high-resolution commercial satellite or prioritizing the processing of the next available cloud-free Sentinel-2 image over that specific location to confirm or deny the presence of plastic debris using spectral methods. This synergistic approach leverages the strengths of both sensor types to build a more robust and reliable monitoring system.

## IV. Data Fusion: Combining Sentinel-1 and Sentinel-2

Data fusion aims to create a single, more informative dataset by combining the complementary strengths of different sensors. For marine debris, the most promising fusion approach involves integrating the all-weather, day/night capabilities of Sentinel-1 SAR with the rich spectral information of Sentinel-2 optical imagery.[23] This creates the potential for a more robust and reliable detection system than either sensor could provide alone.

Fusion can be implemented at several levels:

- **Decision-Level Fusion:** This is the simplest approach, where each sensor's data is processed independently to produce a classification map, and the final decision is made based on a rule, such as flagging only those pixels identified as potential debris by both sensors.
- **Feature-Level Fusion:** A more sophisticated approach involves creating a stacked dataset where features from both sensors are combined before being fed into a single classifier. For example, a study demonstrated the effectiveness of an Artificial Neural Network (ANN) trained on a 17-layer data stack. This stack consisted of Sentinel-2's spectral bands, various calculated optical indices, and the Sentinel-1 VV backscatter data.[23] This allows the model to learn the complex interplay between the spectral and structural (roughness) properties of floating debris.

The primary challenge for fusing Sentinel-1 and Sentinel-2 data for this application is the temporal mismatch between their acquisitions.[23] Floating plastic is a dynamic target that can drift several kilometers per day.[14] Since the two satellites do not acquire data over the same location at the same time, a feature detected in a Sentinel-1 image may have moved by the time the next Sentinel-2 image is available. Consequently, data fusion is most effective for monitoring relatively stationary or slow-moving accumulations, such as those trapped in coastal embayments, river mouths, or behind dams.[23]

## V. Future Improvements and Advanced AI

The next leap in satellite-based plastic detection will be driven by advancements in artificial intelligence, particularly the application of large-scale machine learning models. The current approach often relies on models trained on limited, manually labeled datasets. The future lies in leveraging **geospatial foundation models**, a concept pioneered by major technology organizations like Google and others in the AI research community.

These foundation models are trained on vast, diverse archives of unlabeled satellite imagery, encompassing petabytes of data from various sensors (optical, SAR, etc.) across the globe. Through self-supervised learning, they develop a deep, intrinsic understanding of what constitutes "normal" for the Earth's surface. This allows them to become exceptionally adept at **anomaly detection**. A patch of plastic debris, with its unique spectral and textural properties, represents a significant anomaly against the learned baseline of water, waves, foam, and natural slicks.

The key advantages of this machine learning approach are:

- **Reduced Reliance on Labeled Data:** By pre-training on enormous unlabeled datasets, these models require far less specific, hand-labeled data for "fine-tuning" on the plastic detection task, overcoming a major bottleneck in current research.
- **Enhanced Generalization:** Because the models learn from a global dataset, they are less prone to regional bias. A model trained to detect plastic in Southeast Asia would be more likely to perform well in the Caribbean without extensive retraining.
- **Multi-Modal Integration:** Foundation models can be designed to natively process and fuse data from different sources. They can learn the complex correlations between a SAR backscatter anomaly (from Sentinel-1) and a corresponding spectral signature (from Sentinel-2), leading to more robust and confident detections than single-sensor analysis.

This paradigm shift moves away from building small, specialized models for a single task and towards creating large, adaptable AI systems that can be applied to a range of environmental monitoring challenges, with marine plastic detection being a prime use case.

## Works Cited

1. Emerging Technologies for Remote Sensing of Floating and Submerged Plastic Litter, accessed on August 16, 2025, <https://www.mdpi.com/2072-4292/16/10/1770>

2. (PDF) Remote sensing of marine plastic debris using satellite images off the coast of the Basque Country - ResearchGate, accessed on August 16, 2025, [https://www.researchgate.net/publication/377527728\\_Remote\\_sensing\\_of\\_marine\\_plastic\\_debris\\_using\\_satellite\\_images\\_off\\_the\\_coast\\_of\\_the\\_Basque\\_Country](https://www.researchgate.net/publication/377527728_Remote_sensing_of_marine_plastic_debris_using_satellite_images_off_the_coast_of_the_Basque_Country)
3. Large-scale detection of marine debris in coastal areas with Sentinel-2 - PMC, accessed on August 16, 2025, <https://pmc.ncbi.nlm.nih.gov/articles/PMC10709011/>
4. Remote detection of marine debris using satellite observations in the ..., accessed on August 16, 2025, [https://www.researchgate.net/publication/350643363\\_Remote\\_detection\\_of\\_marine\\_debris\\_using\\_satellite\\_observations\\_in\\_the\\_visible\\_and\\_near\\_infrared\\_spectral\\_range\\_Challenges\\_and\\_potentials](https://www.researchgate.net/publication/350643363_Remote_detection_of_marine_debris_using_satellite_observations_in_the_visible_and_near_infrared_spectral_range_Challenges_and_potentials)
5. Cleaning our oceans with observations and models - Copernicus keeps track of plastic pollution in oceans, accessed on August 16, 2025, <https://www.copernicus.eu/en/news/news/observer-cleaning-our-oceans-observations-and-models-copernicus-keeps-track-plastic>
6. Exploring the potential of remote sensing to detect marine plastic debris in the South African Ocean region, accessed on August 16, 2025, <https://isprs-annals.copernicus.org/articles/X-G-2025/665/2025/isprs-annals-X-G-2025-665-2025.pdf>
7. Towards Spaceborne Detection of Ocean Plastic | Updates, accessed on August 16, 2025, <https://theoceancleanup.com/updates/towards-spaceborne-detection-of-ocean-plastic/>
8. Monitoring of Plastic Islands in River Environment Using Sentinel-1 SAR Data, accessed on August 16, 2025, [https://www.researchgate.net/publication/363384976\\_Monitoring\\_of\\_Plastic\\_Islands\\_in\\_River\\_Environment\\_Using\\_Sentinel-1\\_SAR\\_Data](https://www.researchgate.net/publication/363384976_Monitoring_of_Plastic_Islands_in_River_Environment_Using_Sentinel-1_SAR_Data)
9. Measuring Marine Plastic Debris from Space: Initial Assessment of Observation Requirements - MDPI, accessed on August 16, 2025, <https://www.mdpi.com/2072-4292/11/20/2443>
10. Remote Sensing of Plastic Marine Litter - HeroX, accessed on August 16, 2025, <https://www.herox.com/remotesensing>
11. On advances, challenges and potentials of remote sensing image analysis in marine debris and suspected plastics monitoring - Frontiers, accessed on August 16, 2025, <https://www.frontiersin.org/journals/remote-sensing/articles/10.3389/frsen.2023.1302384/full>
12. Detecting ocean plastic pollution with remote sensing technologies - Deltares, accessed on August 16, 2025, <https://www.deltares.nl/en/news/detecting-plastic-pollution-remote-sensing>
13. The satellite detection of marine plastic debris - SiNCEM, accessed on August 16, 2025, [https://www.sincem.unibo.it/images/tesi/tesi\\_Pelusi.pdf](https://www.sincem.unibo.it/images/tesi/tesi_Pelusi.pdf)
14. Review of Methods for Automatic Plastic Detection in Water Areas Using Satellite Images and Machine Learning - PubMed Central, accessed on August 16, 2025, <https://pmc.ncbi.nlm.nih.gov/articles/PMC11359068/>
15. S2 Mission - SentiWiki, accessed on August 16, 2025,

- <https://sentiwiki.copernicus.eu/web/s2-mission>
16. Sentinel 2 Bands and Combinations - GIS Geography, accessed on August 16, 2025, <https://gisgeography.com/sentinel-2-bands-combinations/>
  17. Assessment of Sentinel-2 MSI Spectral Band Reflectances for Estimating Fractional Vegetation Cover - MDPI, accessed on August 16, 2025, <https://www.mdpi.com/2072-4292/10/12/1927>
  18. Review of Methods for Automatic Plastic Detection in Water Areas Using Satellite Images and Machine Learning - MDPI, accessed on August 16, 2025, <https://www.mdpi.com/1424-8220/24/16/5089>
  19. Sentinel-2 - Wikipedia, accessed on August 16, 2025, <https://en.wikipedia.org/wiki/Sentinel-2>
  20. (PDF) Comparative Review of Remote Sensing Methods for Ocean ..., accessed on August 16, 2025, [https://www.researchgate.net/publication/385904825\\_Comparative\\_Review\\_of\\_Remote\\_Sensing\\_Methods\\_for\\_Ocean\\_Plastic\\_Litter\\_Detection](https://www.researchgate.net/publication/385904825_Comparative_Review_of_Remote_Sensing_Methods_for_Ocean_Plastic_Litter_Detection)
  21. Microplastics Detection in Freshwater Lakes Using Sentinel-1 Synthetic Aperture Radar Images - Asian Association on Remote Sensing | ACRS | AARS, accessed on August 16, 2025, <https://acrs-aars.org/proceeding/ACRS2024/AB0081.pdf>
  22. Monitoring of Plastic Islands in River Environment Using Sentinel-1 ..., accessed on August 16, 2025, <https://www.mdpi.com/2072-4292/14/18/4473>
  23. Detection of Waste Plastics in the Environment ... - Preprints.org, accessed on August 16, 2025, [https://www.preprints.org/frontend/manuscript/fOd29b4787a0832ab2e8338e4f79081b/download\\_pub](https://www.preprints.org/frontend/manuscript/fOd29b4787a0832ab2e8338e4f79081b/download_pub)