

Integrating Sentinel-1 and Sentinel-2 with AI for Enhanced Flood Monitoring and Disaster Response

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Flood Facts (1990–2022)



Floods are the most frequent natural disaster, affecting over 3.2 billion & 250 million people annually



Damage exceeded US\$ 1.3 trillion globally (1990–2022), with annual losses of US\$ 50–66 billion



218,000 lives are lost due to floods



Flood events are increasing in frequency and severity due to climate change, making early detection and monitoring even more critical

Why Early Flood Detection?



Provides timely warnings for evacuation and resource allocation



Helps disaster response agencies plan rescue and relief operations



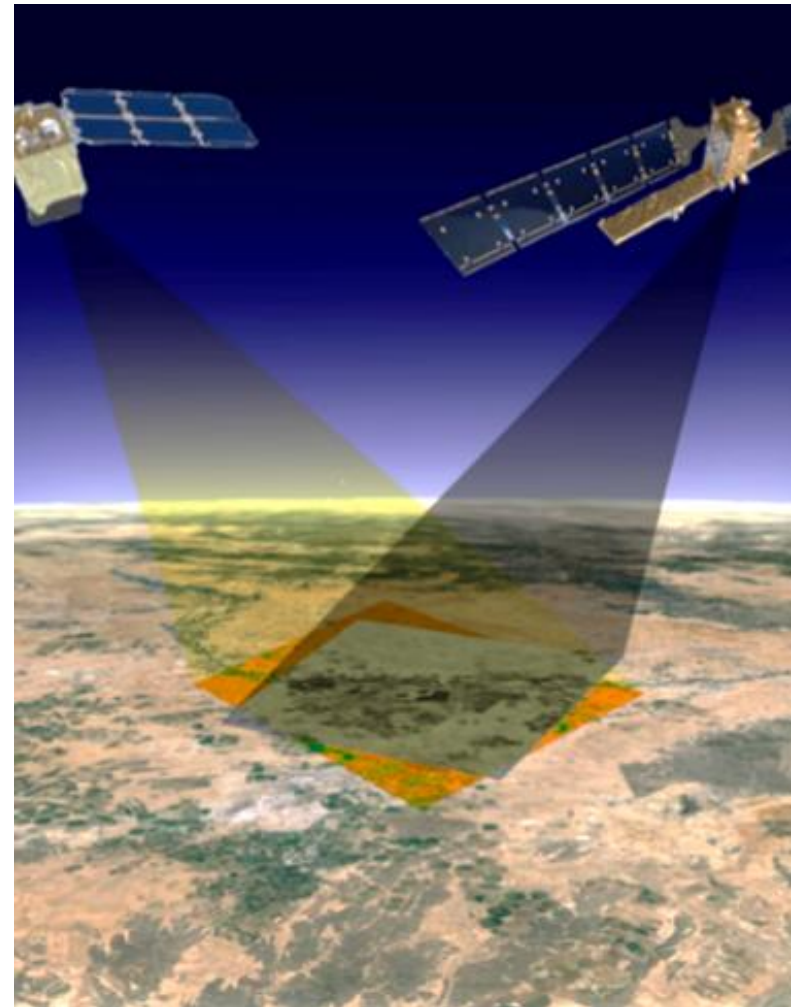
Reduces economic impact by protecting infrastructure and agriculture



Enables climate resilience planning and long-term risk assessment

Why Sentinel-1 and Sentinel-2 Are Widely Used

- Sentinel-1 (SAR): Works day/night, penetrates clouds, ideal for flood detection during storms
- Sentinel-2 (Optical): High-resolution multispectral imagery, good for visual validation & water detection
- Together: Complementary – S1 ensures continuous monitoring, S2 provides spectral information
- Free & global coverage crucial for operational use worldwide
- Frequent revisit (6–12 days), enabling near-real-time flood mapping



TerraMind

Functionality	Description	Example for Flood Detection
Any-to-Any Modality Generation	Generates one modality from another (SAR → Optical, Optical → NDVI)	Cloudy Sentinel-2 imagery can be replaced with synthetic S2-like data generated from Sentinel-1 radar, enabling clear flood mapping
Zero-Shot Capability	Performs flood detection without task-specific fine-tuning	Produces usable flood extent maps immediately for new flood events with no labeled data, reaching ~70% IoU
Few-Shot Learning	Adapts to local conditions with very few labeled examples (1-shot, 5-shot)	Fine-tunes quickly using a handful of labeled flood patches to improve accuracy for a specific region
Thinking-in-Modalities (TiM)	Generates synthetic modalities during training/inference and uses them as additional input	Generates LULC or NDVI maps during fine-tuning to enhance flood detection accuracy (+2% mIoU improvement reported)

Workflow

TerraMind

Data sources

Sentinel-1 / Sentinel-2

Check data availability

Both S1 & S2 present?

No

Yes

TerraMind any-to-any generation
(Generate missing modality (S1 or S2) from available data)

Preprocessing

-S1: Noise removal, calibration, speckle filtering, normalize
-S2: Cloud mask, resample SWIR, compute NDWI/MNDWI, normalize
-Align & stack channels + historical features

Dual-Scale Pretraining

Fusion

S1 channels + S2 bands +
Indices + Historical Features

Token

Convert image to patches
Token embedding + positional encoding

Fine Tuning

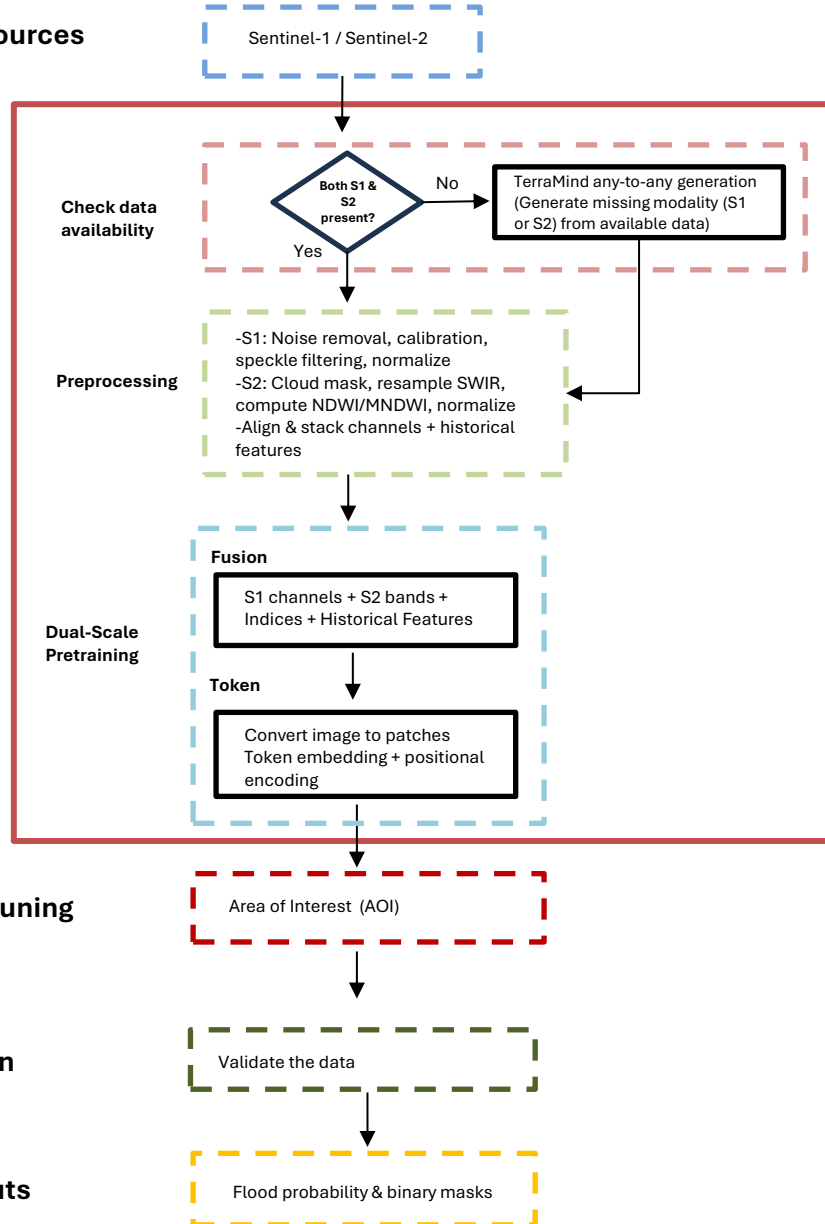
Area of Interest (AOI)

Validation

Validate the data

Outputs

Flood probability & binary masks





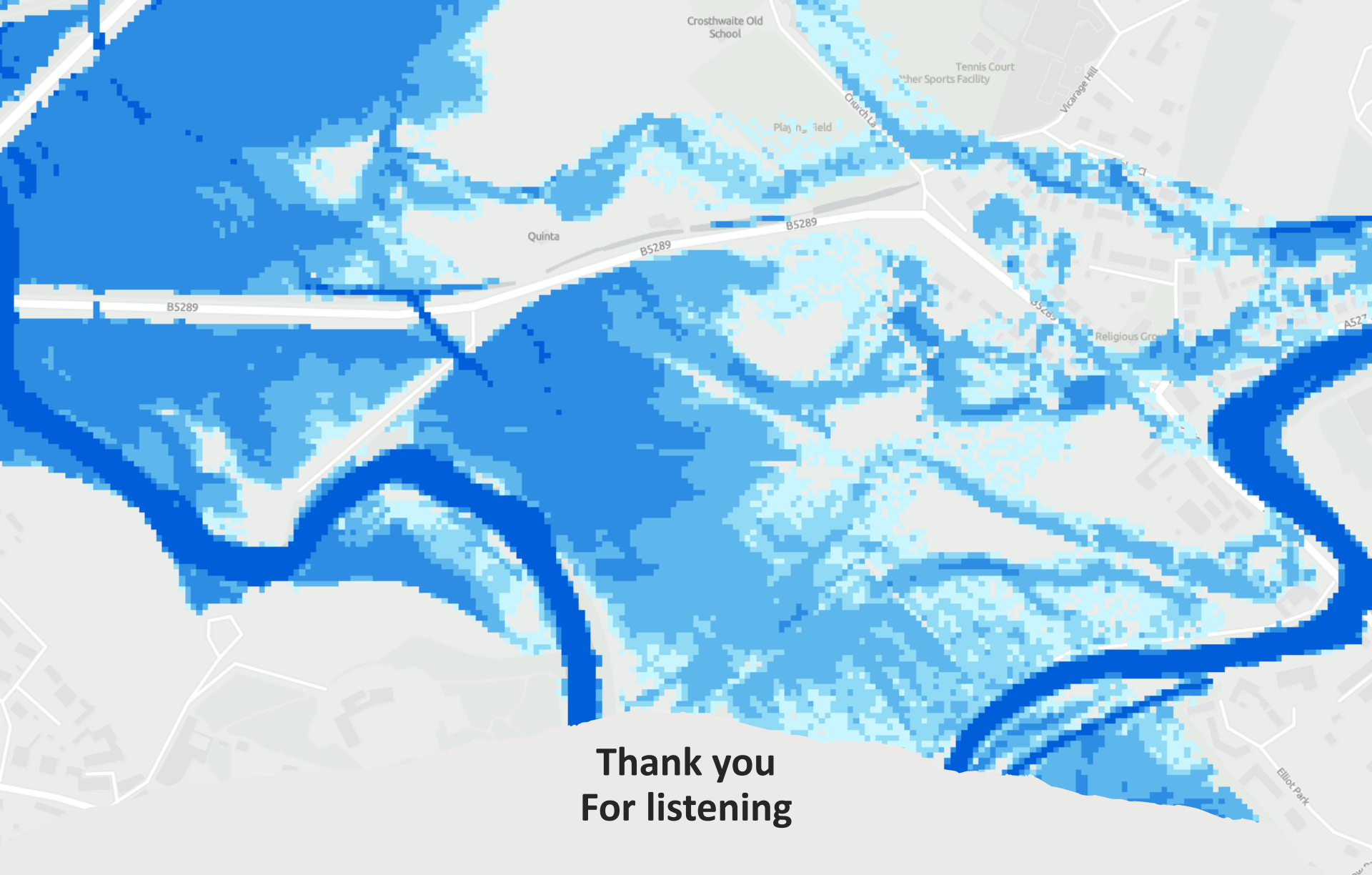
Advantages & Applications

- Works even with missing or noisy data
- Combines radar + optical for all-weather monitoring
- Zero-shot inference reduces need for large labeled datasets
- Generalizes to other disasters: wildfires, drought, deforestation
- Scalable for global monitoring and early warning systems

References

Jakubik, J., Yang, F., Blumenstiel, B., et al. (2025). TerraMind: Large-Scale Generative Multimodality for Earth Observation. ICCV 2025. arXiv:2504.11171 <https://arxiv.org/abs/2504.11171>

Bonafilia, D. et al. (2020). Sen1Floods11: A georeferenced dataset to train and test deep learning flood algorithms for Sentinel-1. CVPR Workshops. https://www.researchgate.net/publication/343275667_Sen1Floods11_a_georeferenced_dataset_to_train_and_test_deep_learning_flood_algorithms_for_Sentinel-1



**Thank you
For listening**

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