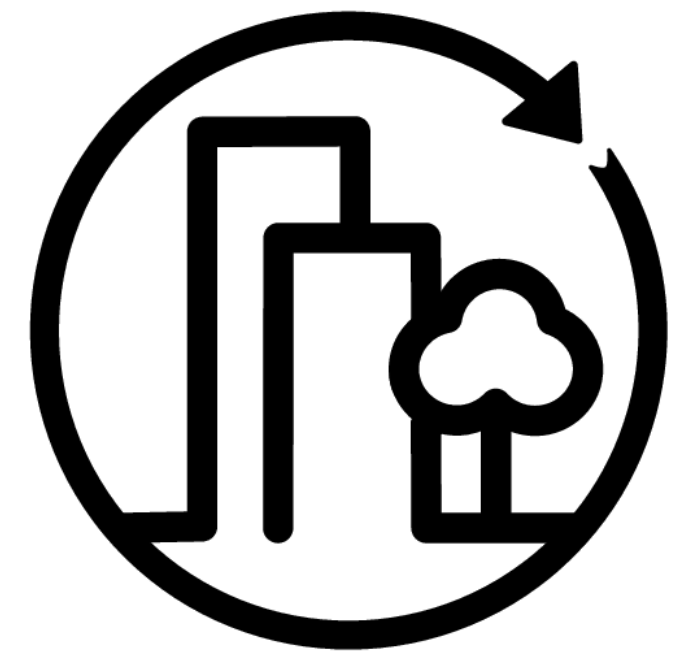




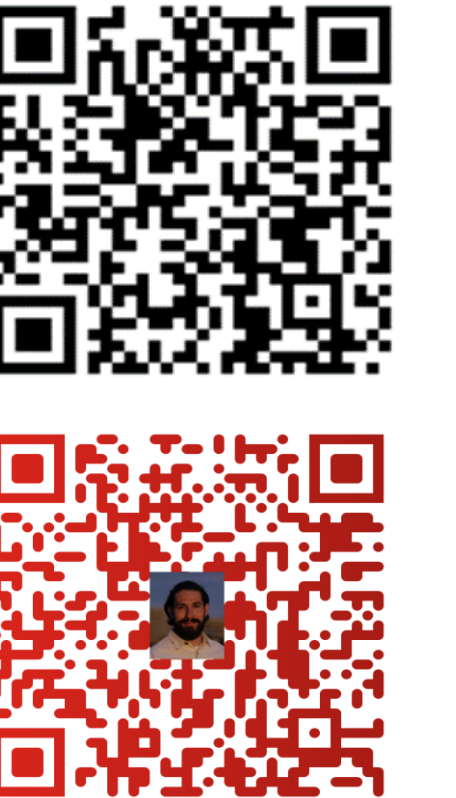
Institute of Ecology
**Urban
Ecosystem
Science**
Technische Universität Berlin

Detection of *Azolla filiculoides* in River Systems using Sentinel 2 Imagery and Random Forest Classification in the Google Earth Engine

Alexander Plakias¹, Mateusz Draga²

¹TU Berlin, Institute of Ecology, Urban Ecosystem Science, Germany

²Adam Mickiewicz University Poznan, UAM, Institute of Environmental Biology



Relevance

Invasive Aquatic Plants

- Threat to biodiversity and Water quality ^{1,2}
- EU Regulations
 - Prevention, mitigation and monitoring required ³
- High management costs ⁴

Species Distribution Data

- Improve predictive modelling (occurrence, spread, boom-bust dynamics) → mitigation and adaptation

Target species: *Azolla filiculoides*

- Small floating fern, unique pink/red coloration, forms hick mats → oxygen depletion ⁵
- Invasive in EU countries: France ⁶, Poland ⁷, Spain ⁸, ...
- Min. winter temperature limits growth → climate change amplifies spread ⁸
- Not (yet) listed as Invasive alien species of EU concern → Lack of aquatic monitoring ⁹, focus on Water Hyacinth (No.1 invasive) ¹⁰⁻¹³



Close-up of a leaf
Image: Wikimedia



A. filiculoides (pink-tinged) with *Lemna minor*



Single *Azolla filiculoides* plant showing the roots



Fig: Mass occurrence of *A. filiculoides* on Tagus, May 2020. Management: Rafting Blockade.
Source: Mateusz Draga Image: Francisco Castañares

Conclusions and Outlook

- Rapid, globally scalable framework with high accuracy for automatic water and vegetation mapping
- Refine Band selection for training: Include aquatic vegetation indices, water indices
- Apply trained classifier to new unseen region
- Peer-review method through manuscript submission

→ Investigate Spread Pattern on national scale / EU scale for last decade

Method

- Pixelwise Classification with Random Forest (RF)
- Automated Workflow: Google Earth Engine
- Water Detection:
 - Individual Classifiers for every scene = accuracy
- A. filiculoides* detection:
 - One classifier = robustness

Validation

Train/Test Area:

- Over 100 km of Tagus River Spain
- 8 years with ~ 850 Scenes

Test-Split validation: water

- Mean metrics of 30 classified scenes:
 - Overall/Test accuracy: 0.94
 - Kappa: 0.89
 - MCC:
 - Ours: 0.89
 - Singh et al., 2020: ¹⁰ 0.69

Sentinel 2 Scene



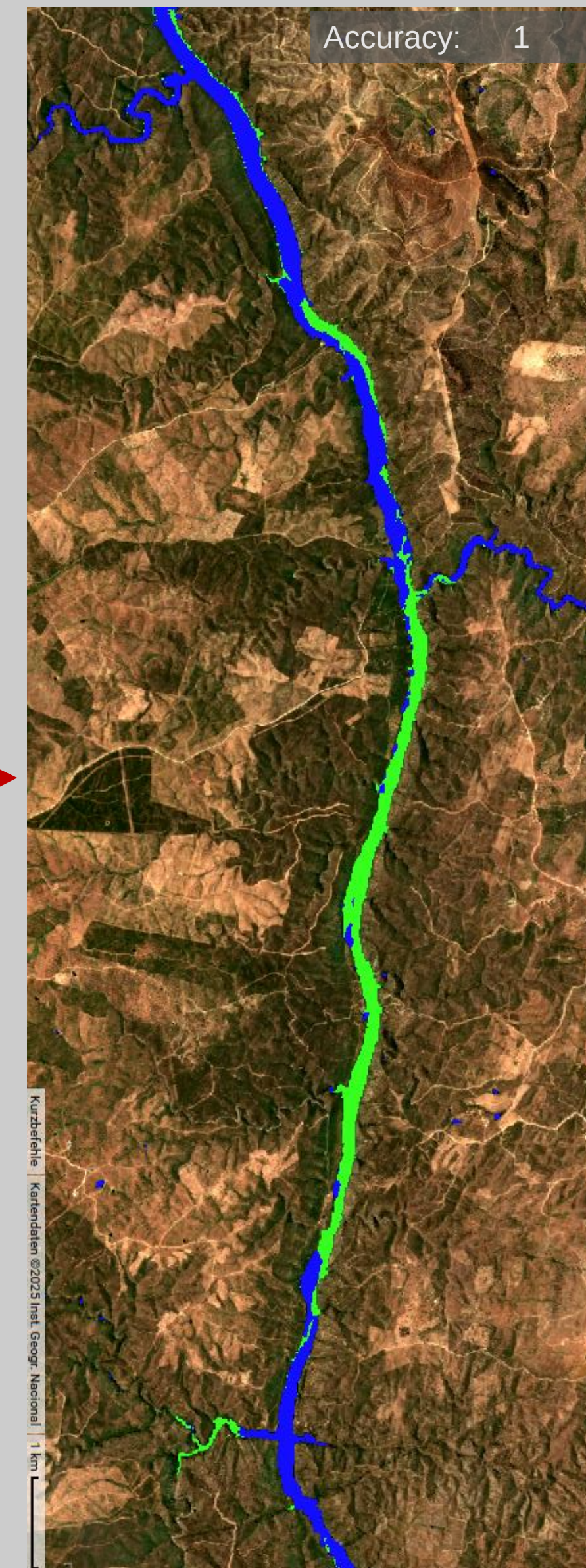
Water Body classification



Training Data:
JRC global water occurrence

1. RF

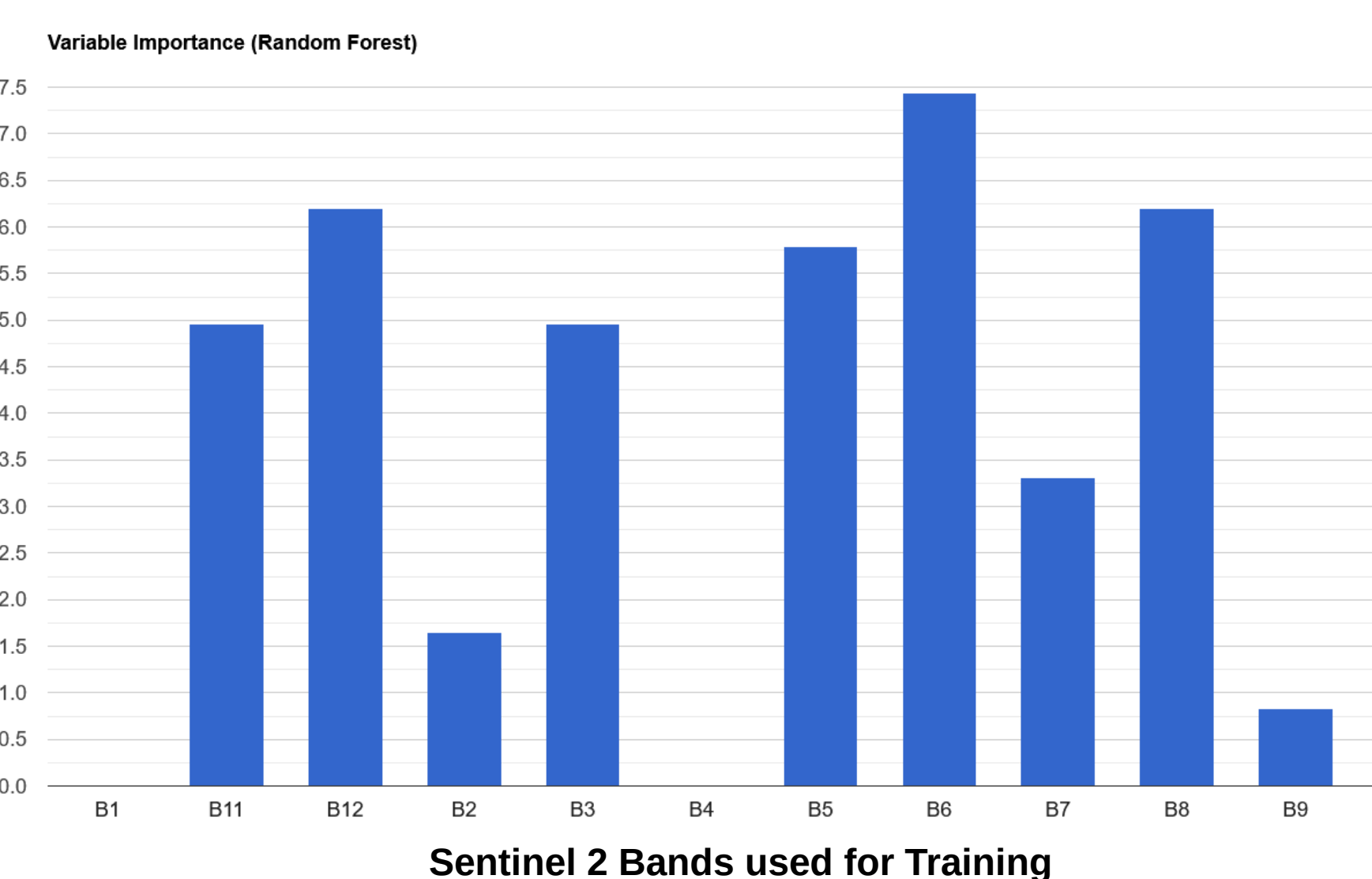
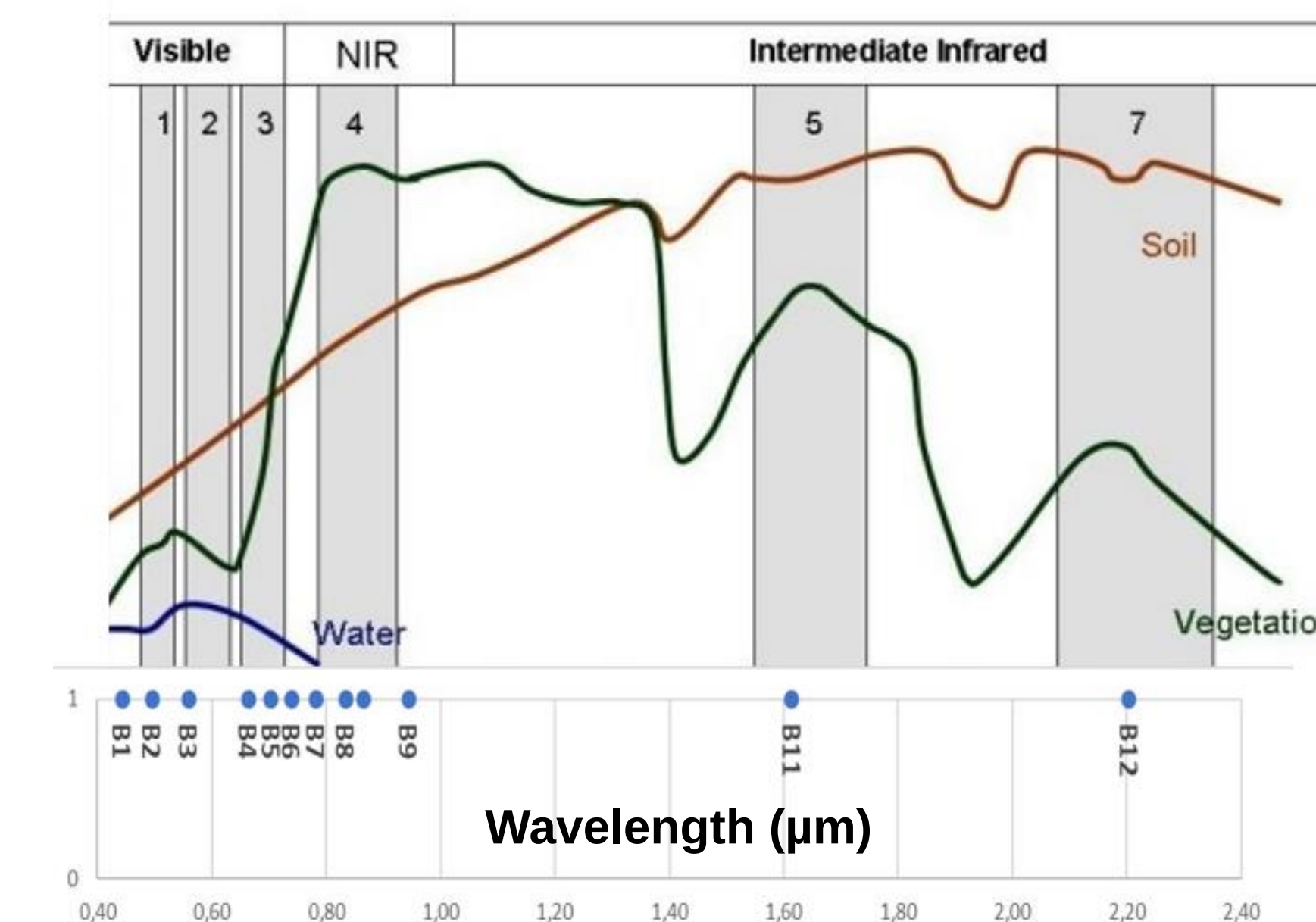
A. filiculoides classification



Training Data:
1000 expert annotated polygons of 10 Scenes

2. RF

Explainability



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