

Mapping baobab populations across savanna landscapes: evidence for an optimal intermediate spatial grain

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MOTIVATION

- Millions of people across Africa depend on baobabs for income, nutrition and traditional medicine.
- Concerns of increased threats by land-use change, climate change and overutilisation.
- Information on their spatial distribution is crucial to support sustainable management.



Image source: Sarah & ABA

MOTIVATION



- ✓ Stakeholders are interested in knowing population estimates to inform production
- **But landscape-scale information remains scarce**

- Acquiring information at scale is challenging



Field Surveys



Slow



Costly



Limited

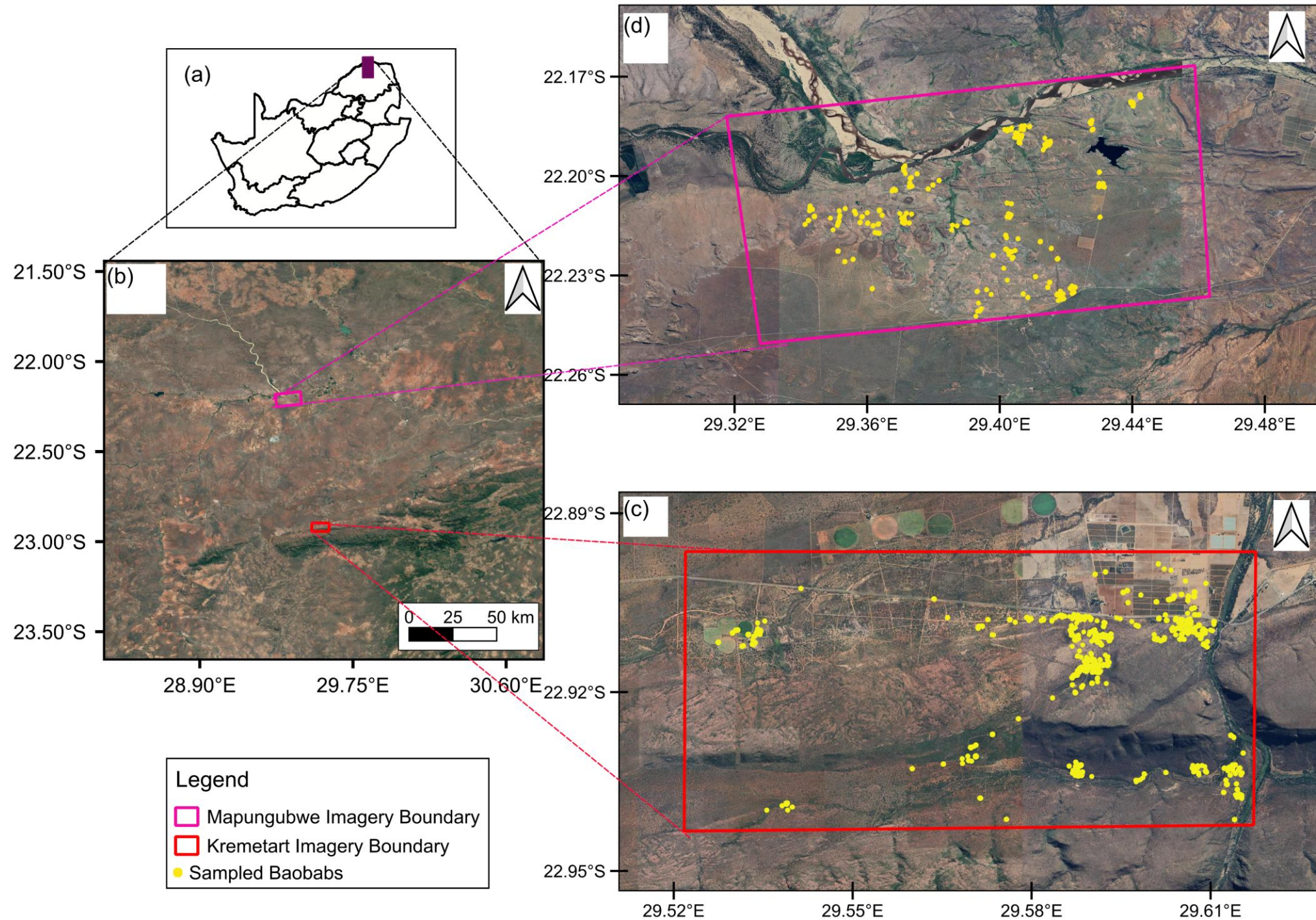
AIM

- To evaluate machine learning (U-Net) modelling approaches and satellite data to map baobab distribution

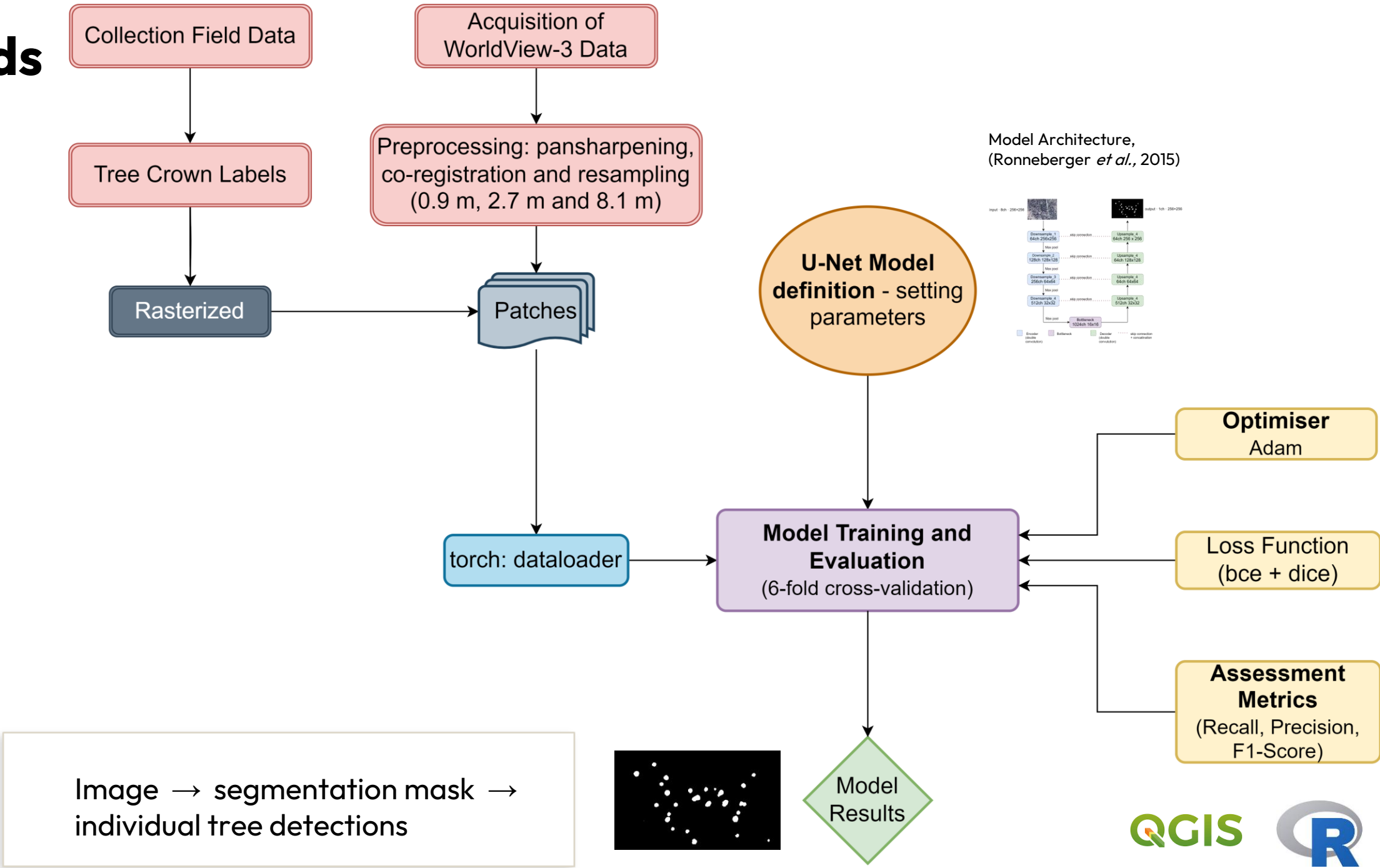
Research Questions

1. Does the U-Net model's performance differ with the spatial grain of imagery?
2. Does baobab canopy size influence U-Net model performance?
3. Does using multi-seasonal imagery improve the U-Net model performance?
4. Does the landscape complexity influence the U-Net model performance?

Study Sites



Methods



Experiments run across four factors with Worldview -3 imagery



Multiple spatial grains



Canopy-size classes



Seasonal combinations



Three land use types

- Same model — only one factor varied at a time.

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How we measured performance

Precision

Of the detected trees,
how many are real?

Recall

Of real trees, how many
did we find?

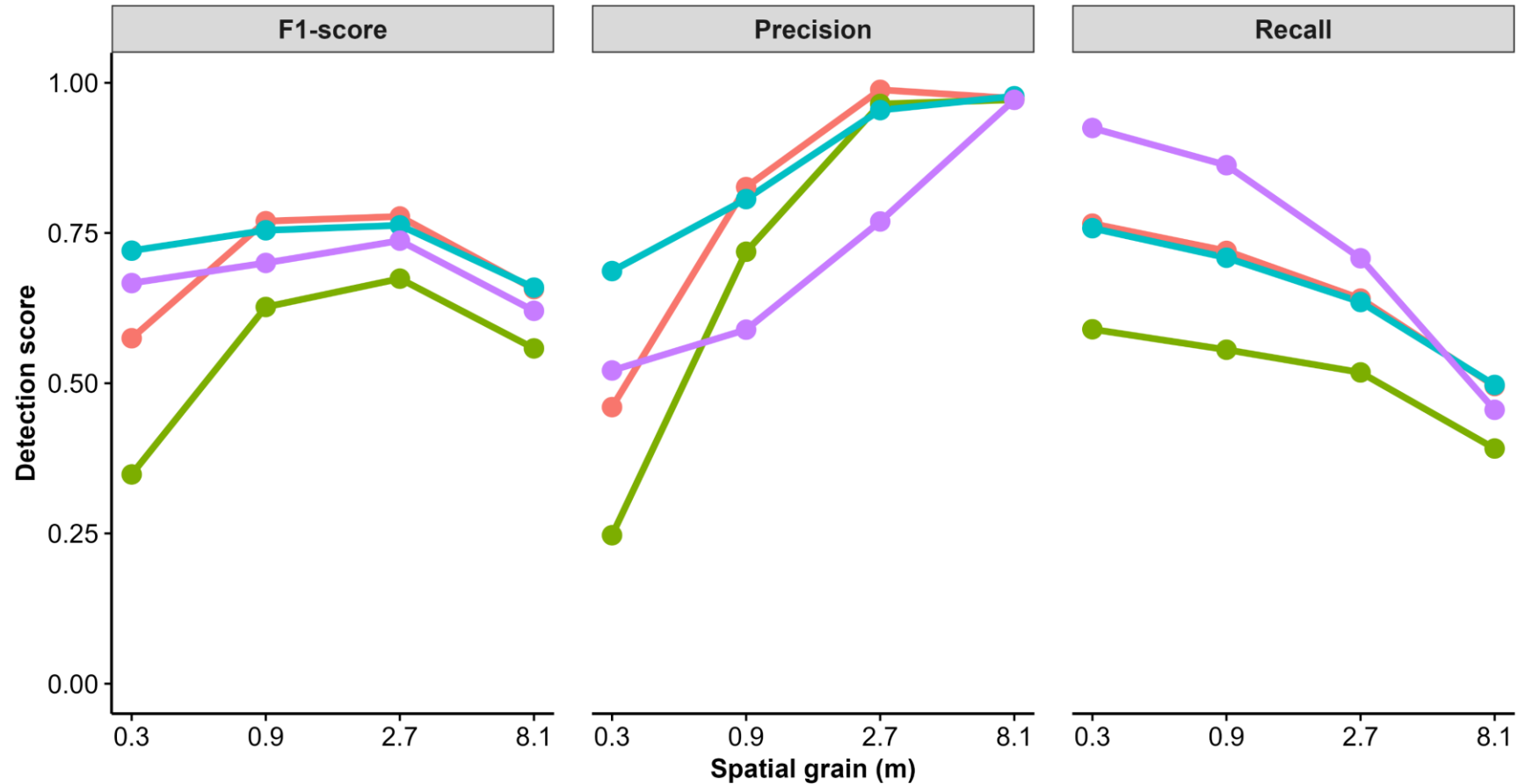
F1-score

Harmonic mean of the
two

RESULTS

Finding 1

U-Net baobab detection performance peaks at intermediate spatial grain



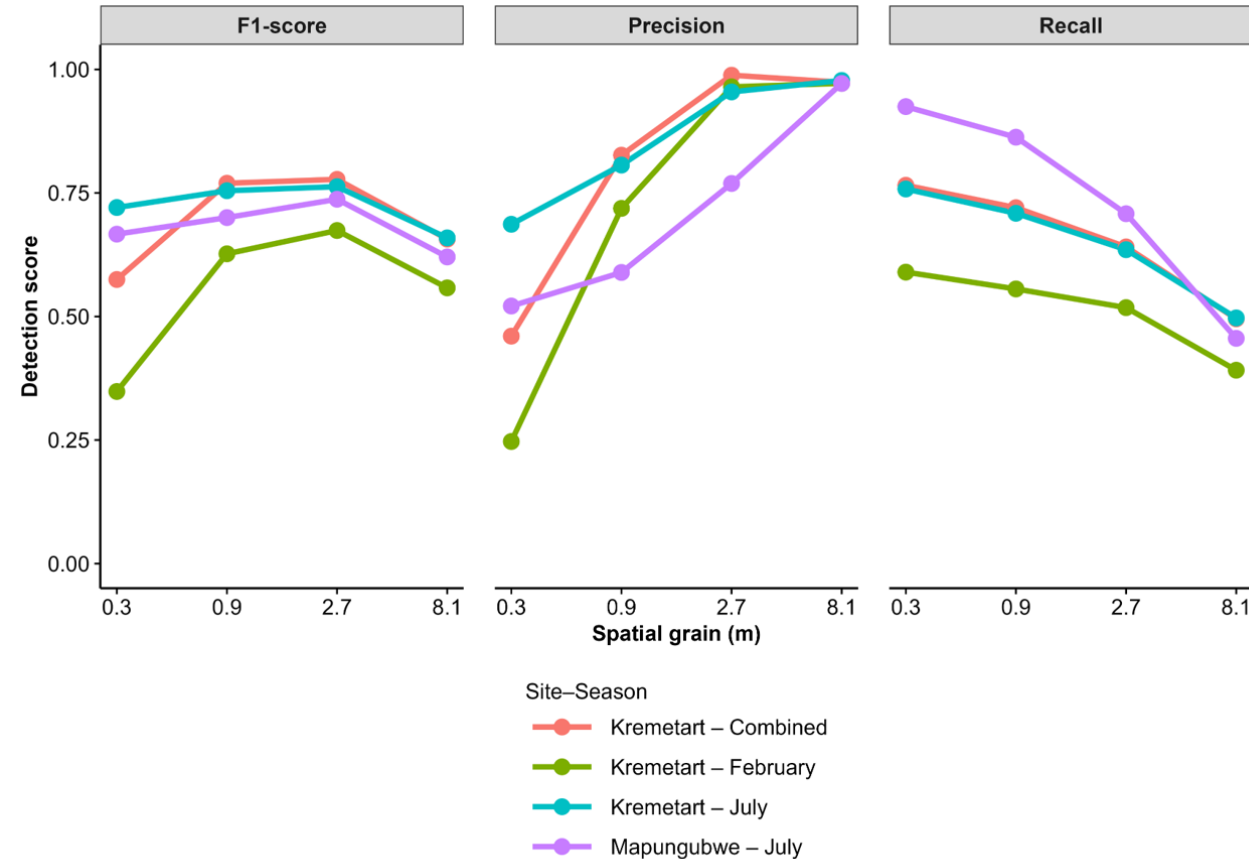
Precision = Of the detected trees, how many are real?

Recall = Of real trees, how many did we find?

F1-Score = Harmonic mean of the two

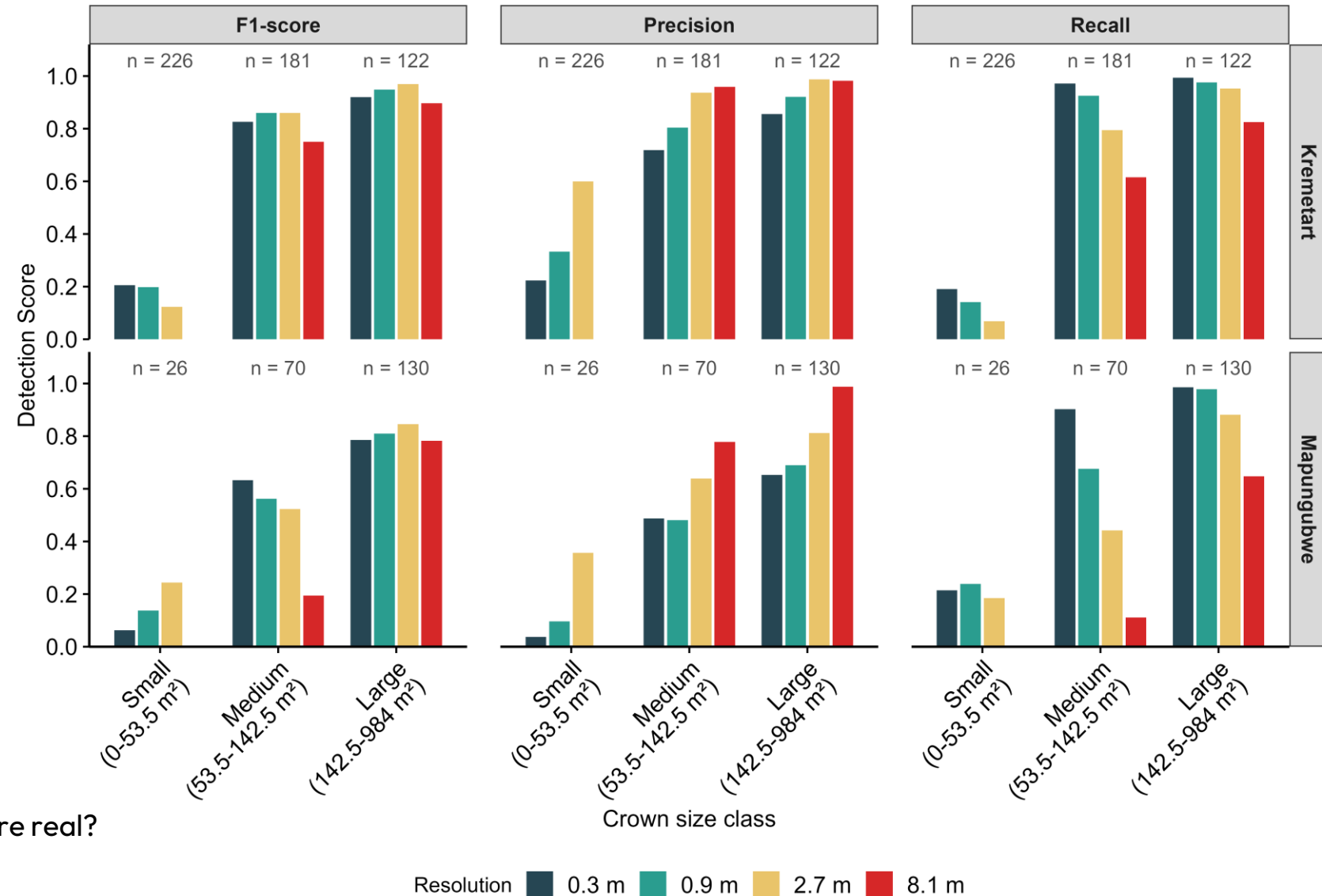
Ecological Implications

- Long-term baobab population monitoring.
 - Choosing the wrong spatial grain can lead to: Underestimation of populations, misleading ecological interpretations



Finding 2

- Large crowns are detected reliably ($F1 \geq 0.85$).
- Small crowns are often missed ($F1 \leq 0.2$, roughly 80% go undetected)



Precision = Of the detected trees, how many are real?

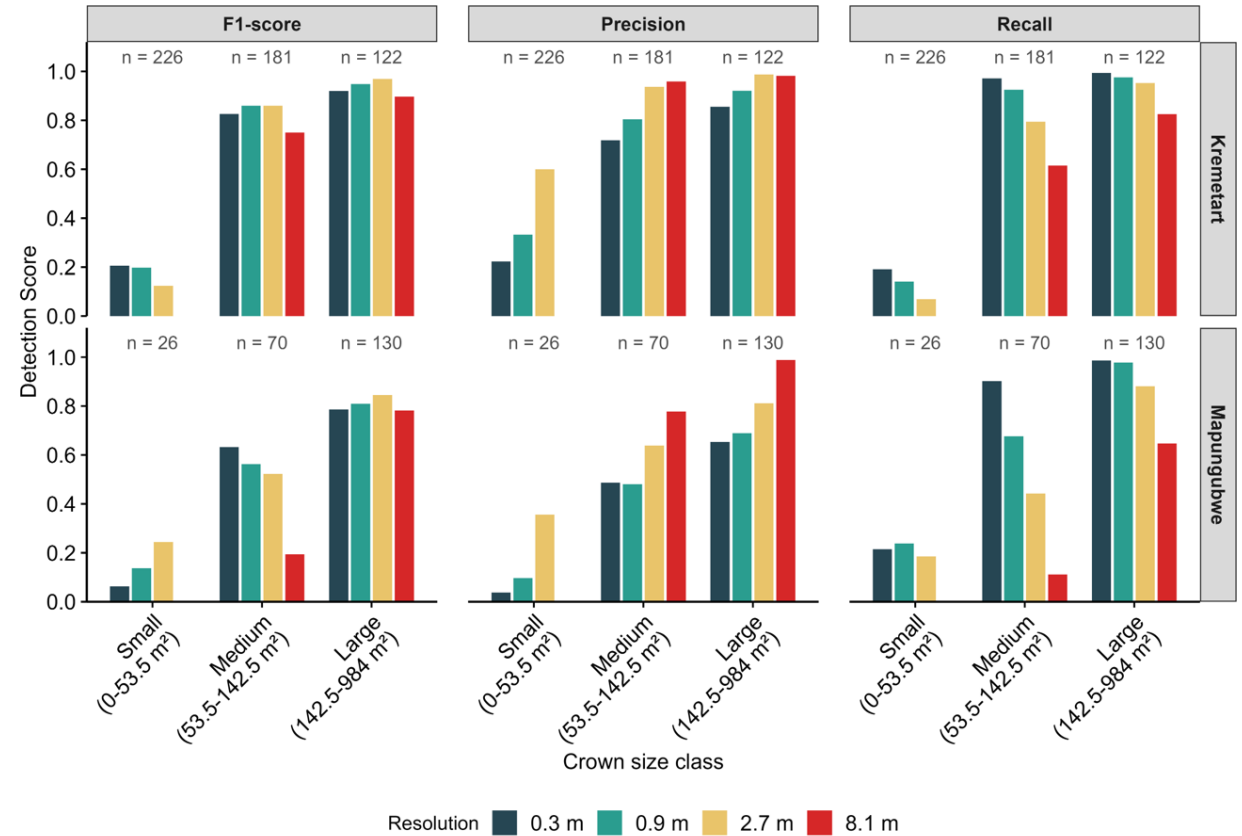
Recall = Of real trees, how many did we find?

F1-Score = Harmonic mean of the two

Resolution 0.3 m 0.9 m 2.7 m 8.1 m

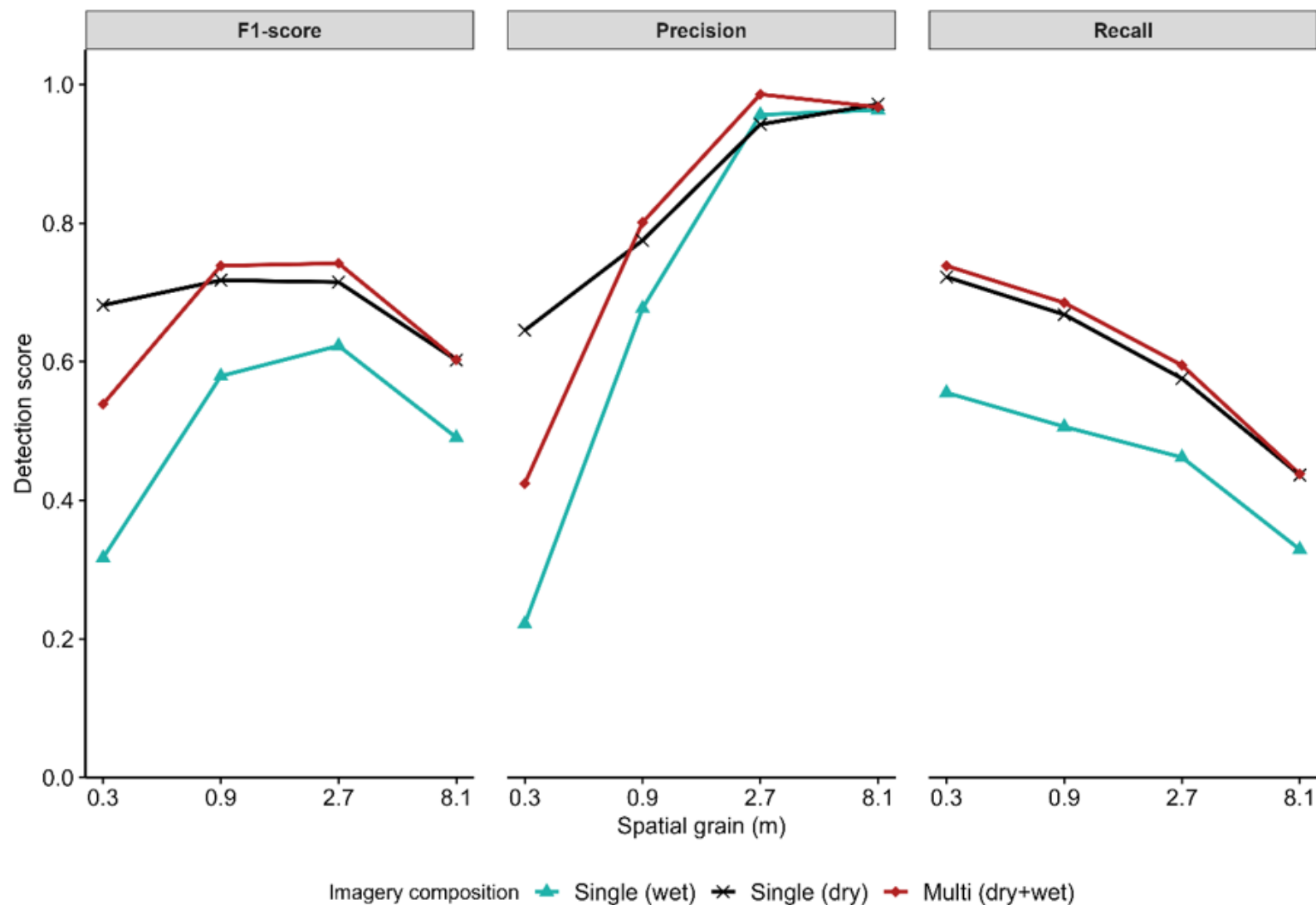
Ecological Implications

- Recruitment assessment and Population structure analysis
 - Juvenile/smaller trees missed
→ regeneration may appear weaker than it is



Finding 3

Multi-season imagery provides the highest baobab detection performance at moderate spatial grain



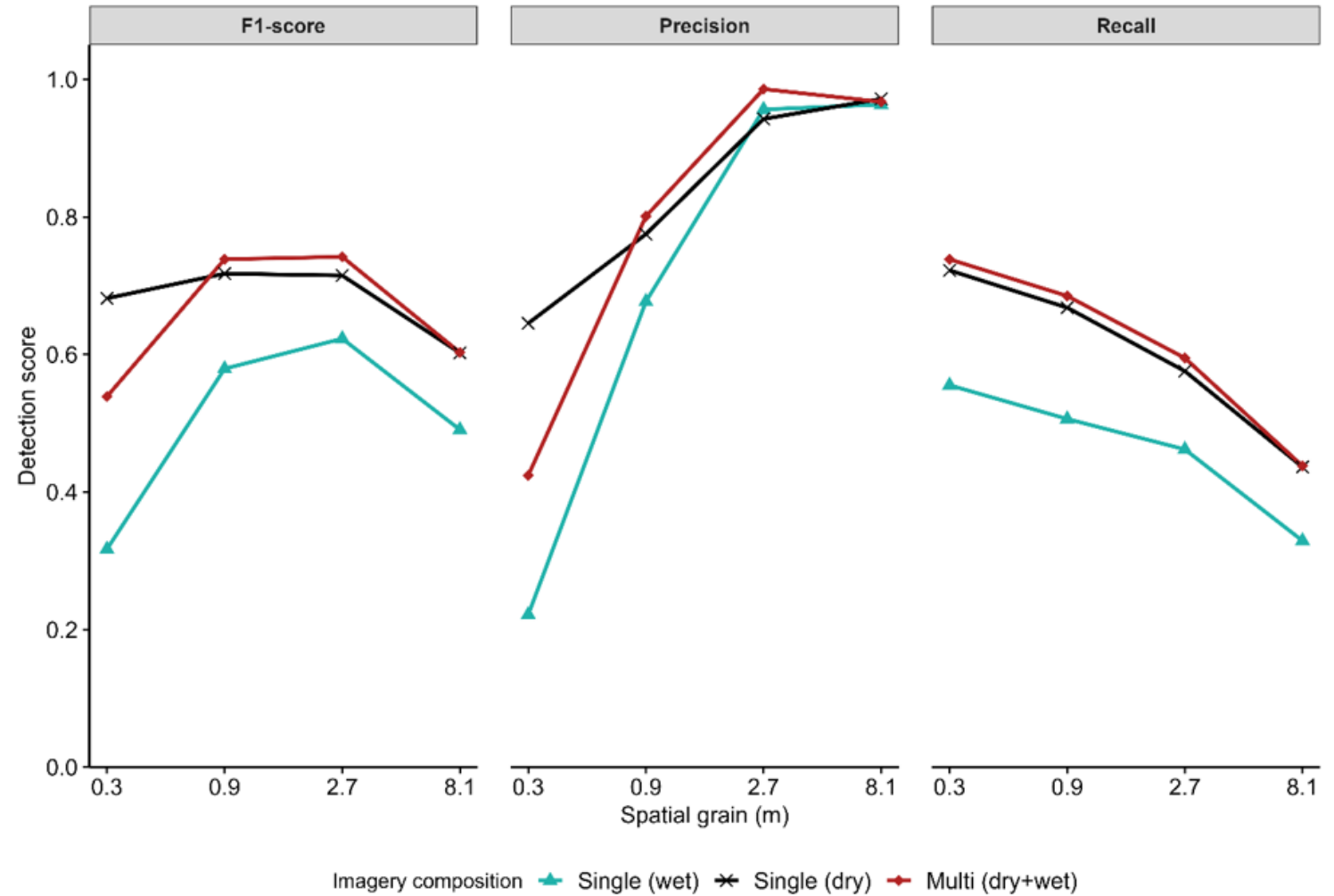
Precision = Of the detected trees, how many are real?

Recall = Of real trees, how many did we find?

F1-Score = Harmonic mean of the two

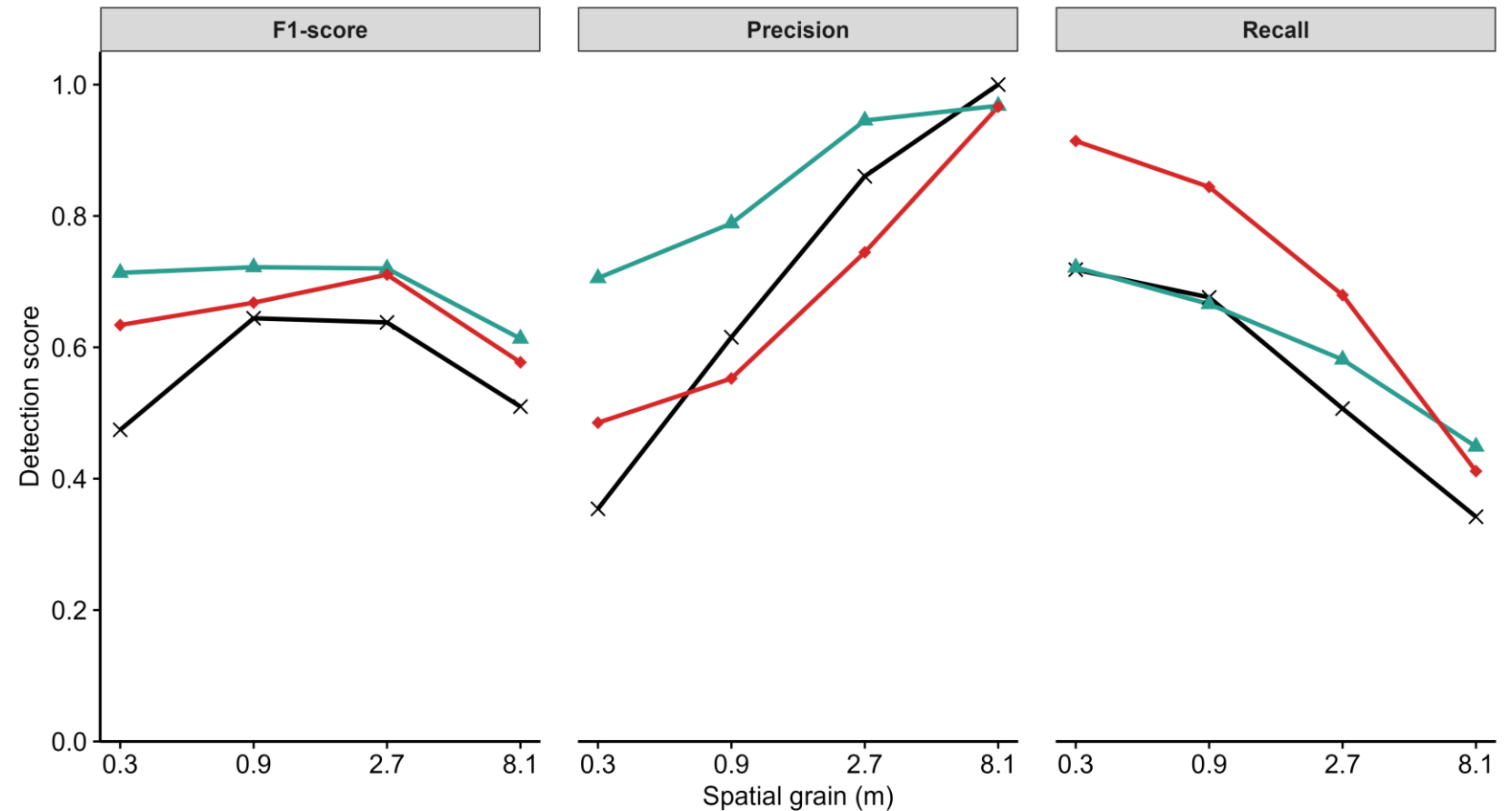
Ecological Implications

- Baobab phenological stages.
 - ✓ Pheno-phases aid detection.



Finding 4

Intermediate spatial grains
maximise baobab detection
performance across
landscapes differing in
complexity



Land-use type

✕ Agriculture Land (high complexity, n = 78)

▲ Game Reserve (moderate complexity, n = 456)

◆ National Park (low complexity, n = 226)

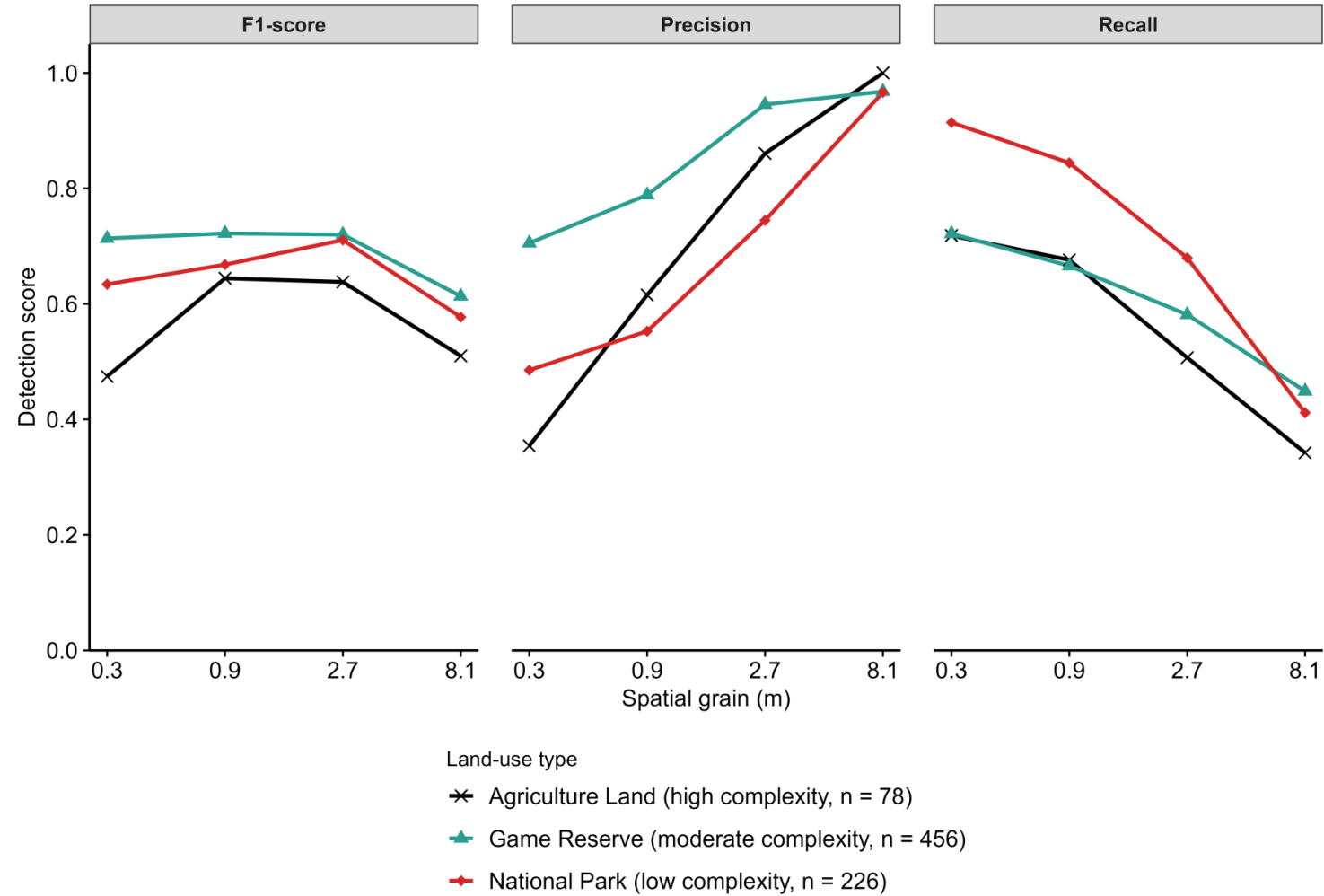
Precision = Of the detected trees, how many are real?

Recall = Of real trees, how many did we find?

F1-Score = Harmonic mean of the two

Ecological Implications

Threats in human-dominated areas can be underestimated.



Precision = Of the detected trees, how many are real?

Recall = Of real trees, how many did we find?

F1-Score = Harmonic mean of the two

Summary

- ✓ Intermediate spatial grain optimises model performance
- ✓ U-net model performance depends on crown size
- ✓ Combined-season imagery improves baobab detection.