

Scan Abstract

## 1. PROBLEM IDENTIFICATION

**PROBLEM STATEMENT**

- Landslide mapping in arid regions is highly uncertain.
- Multiple factors interact in complex ways.
- Most models act as black boxes.
- Irrelevant variables reduce model reliability

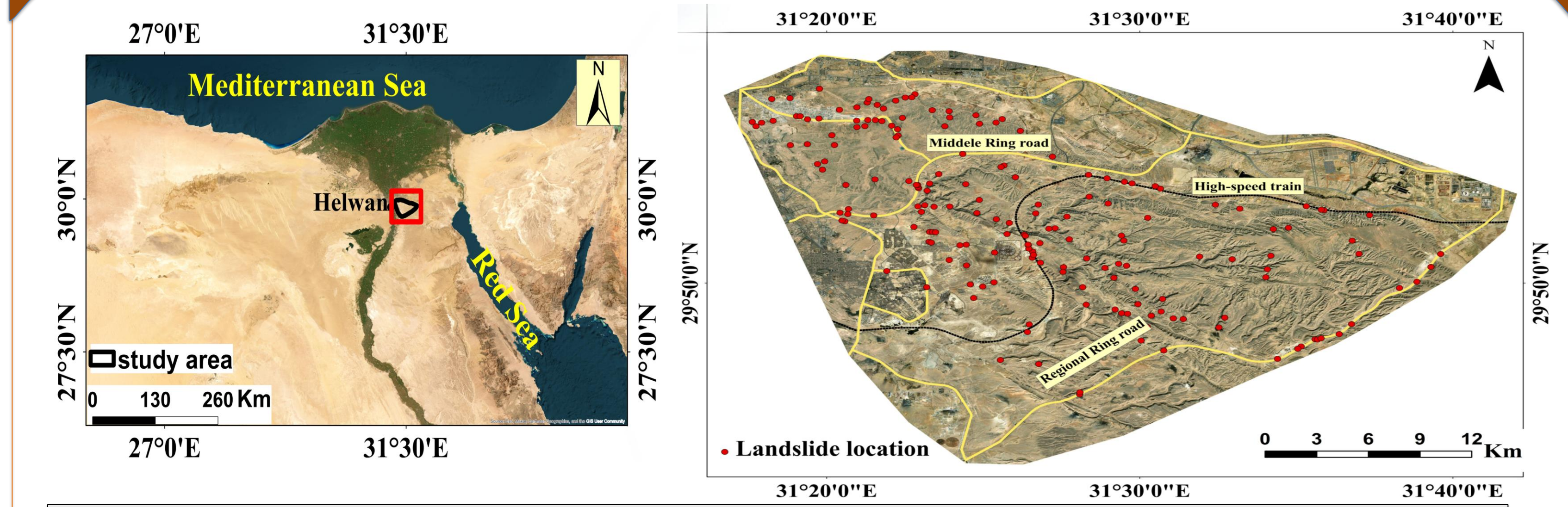
**RESEARCH GAP**

- High accuracy ≠ clear understanding.
- Key drivers are often unknown.
- Models lack physical interpretability.
- Limited support for decision-making.

**OBJECTIVE**

- Apply SHAP-based explainable AI.
- Integrate with recursive feature elimination (RFE). Select the most relevant predictors.
- Improve transparency and performance.

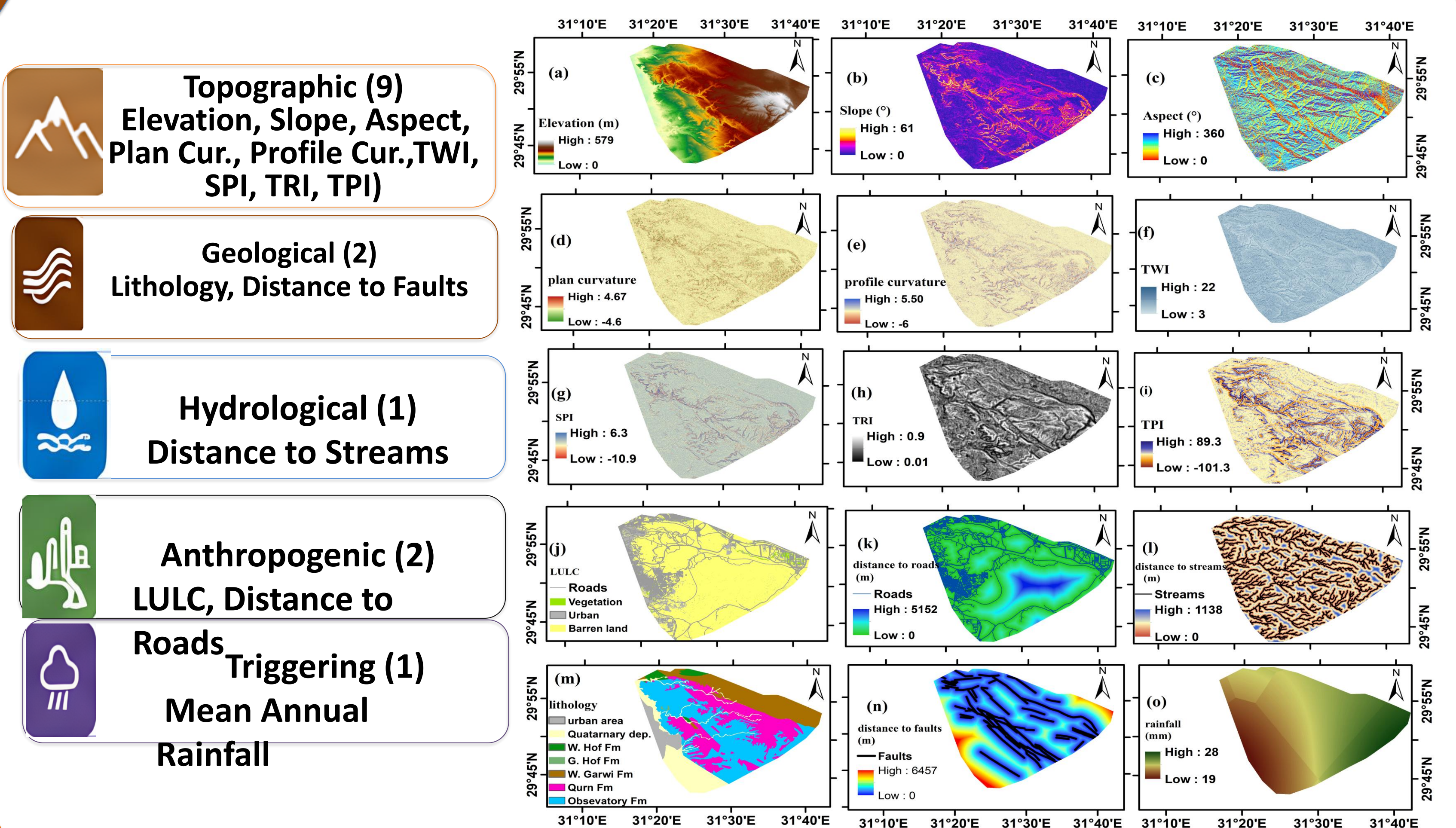
## 2. Study area: East Cairo, Egypt



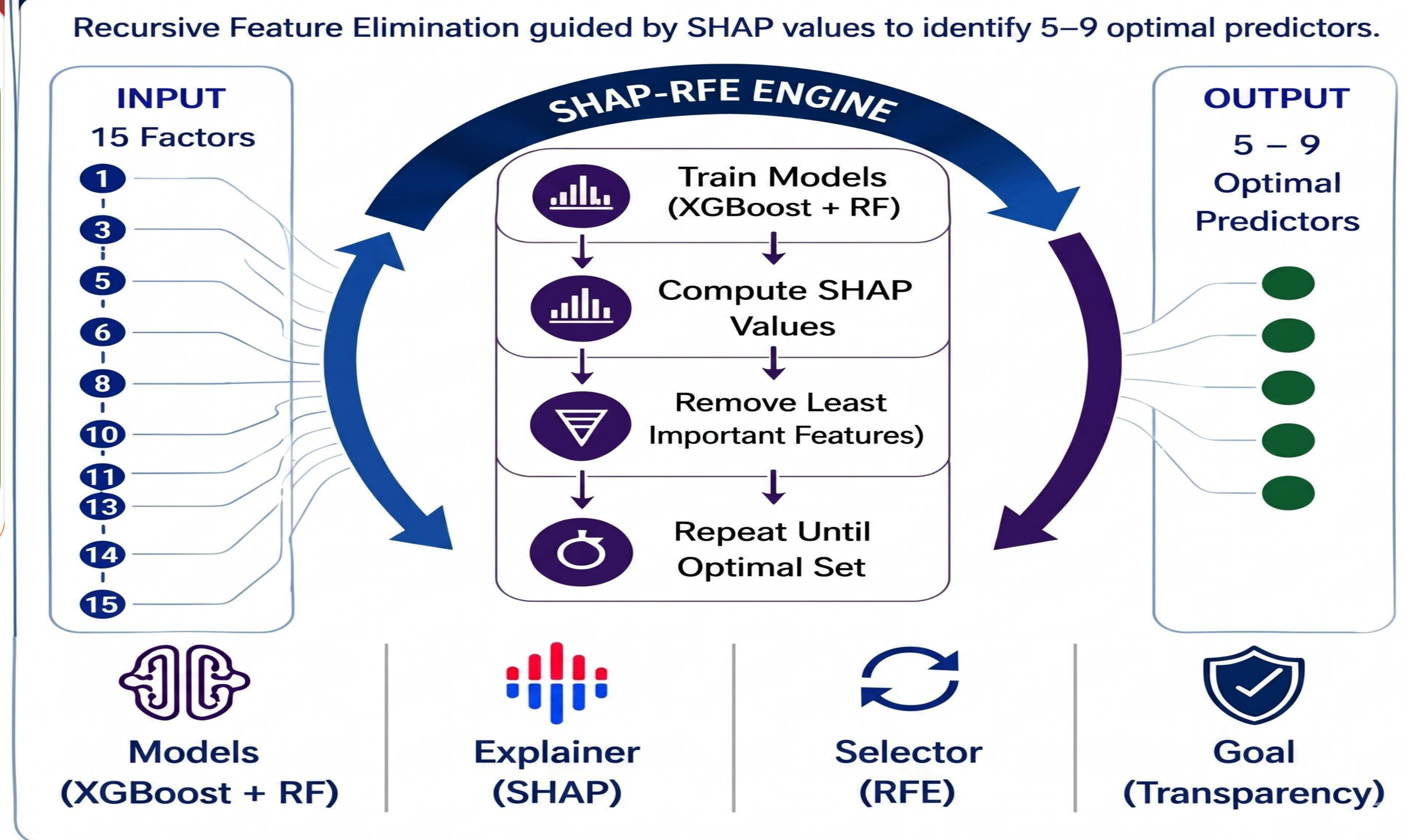
Field evidence of landslide occurrences in the study area



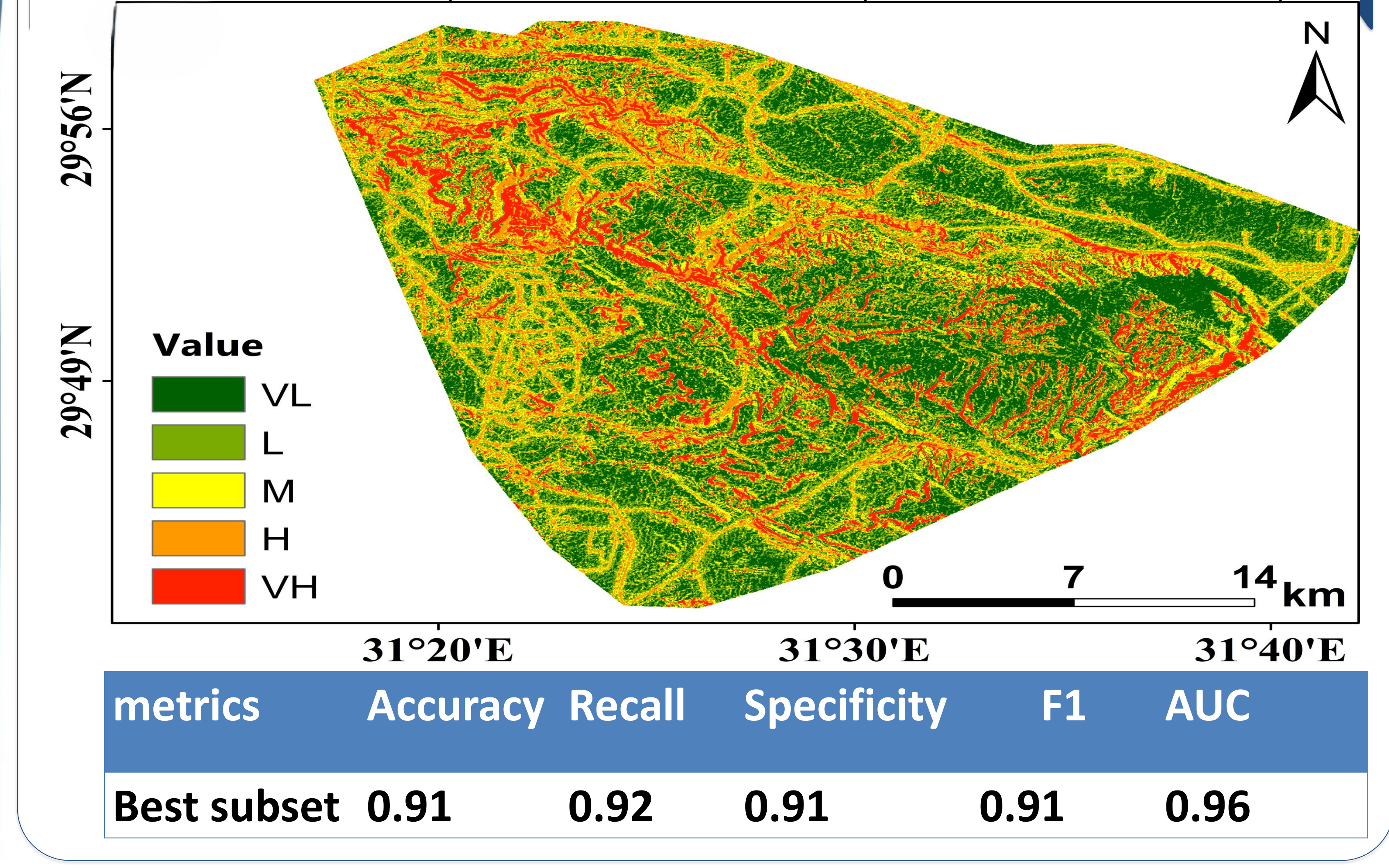
## 2. Conditioning Factors (15)



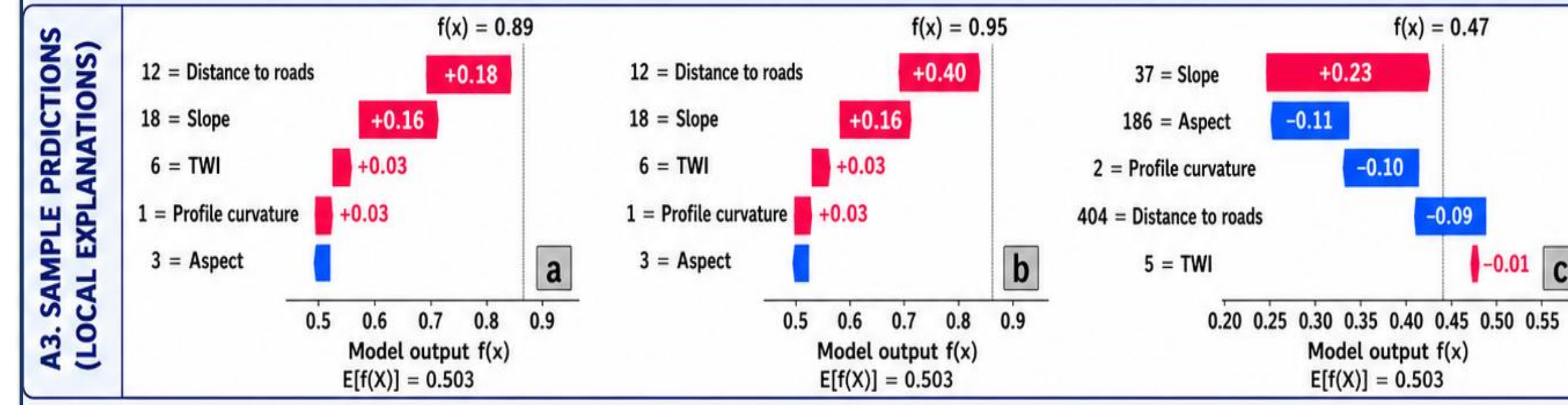
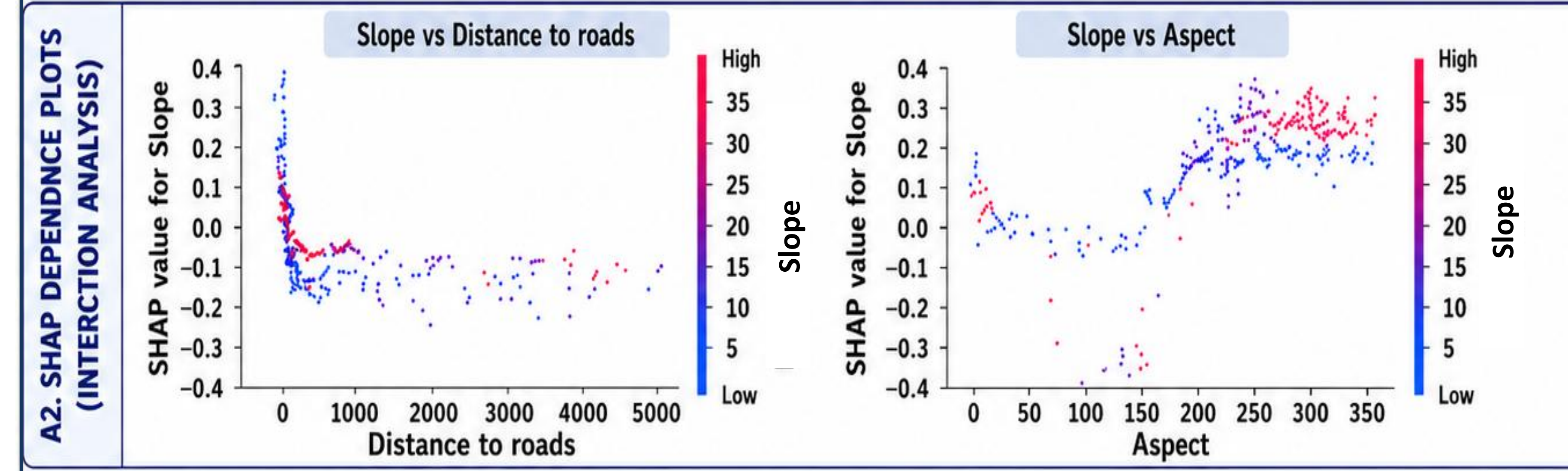
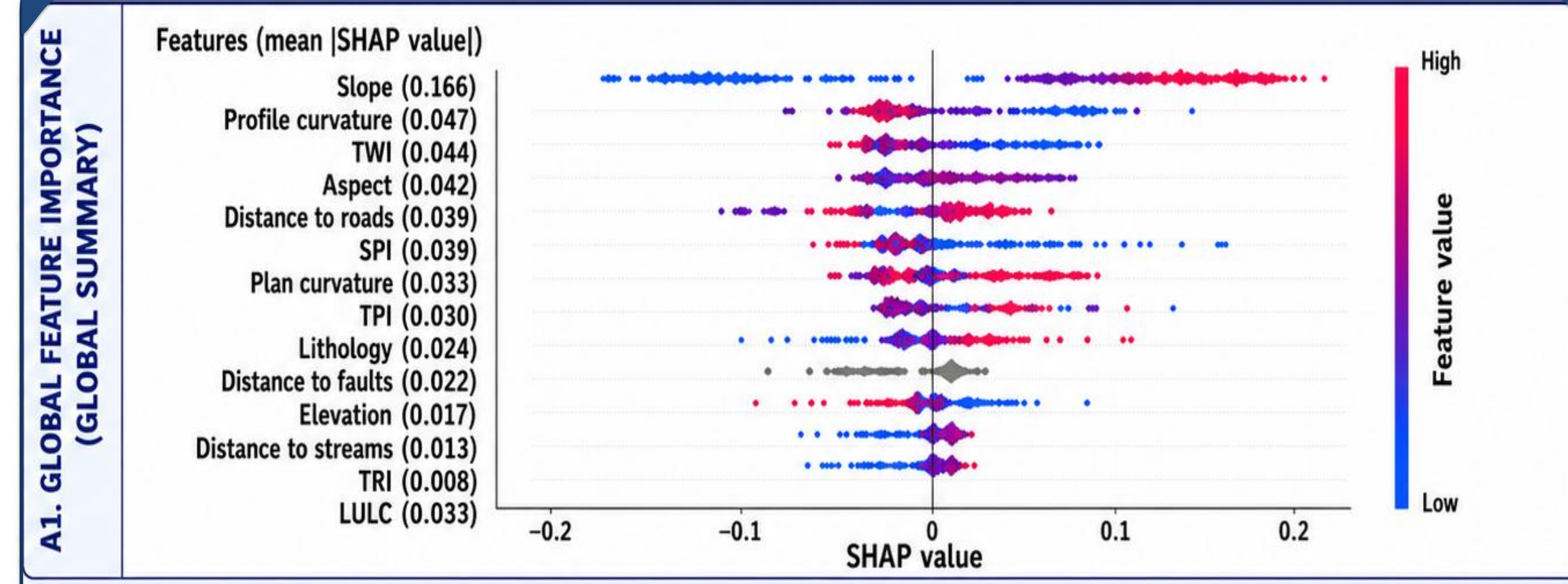
## 4. METHODOLOGY: SHAP-RFE ENGINE



## 6. LANDSLIDE SUSCEPTIBILITY MAP



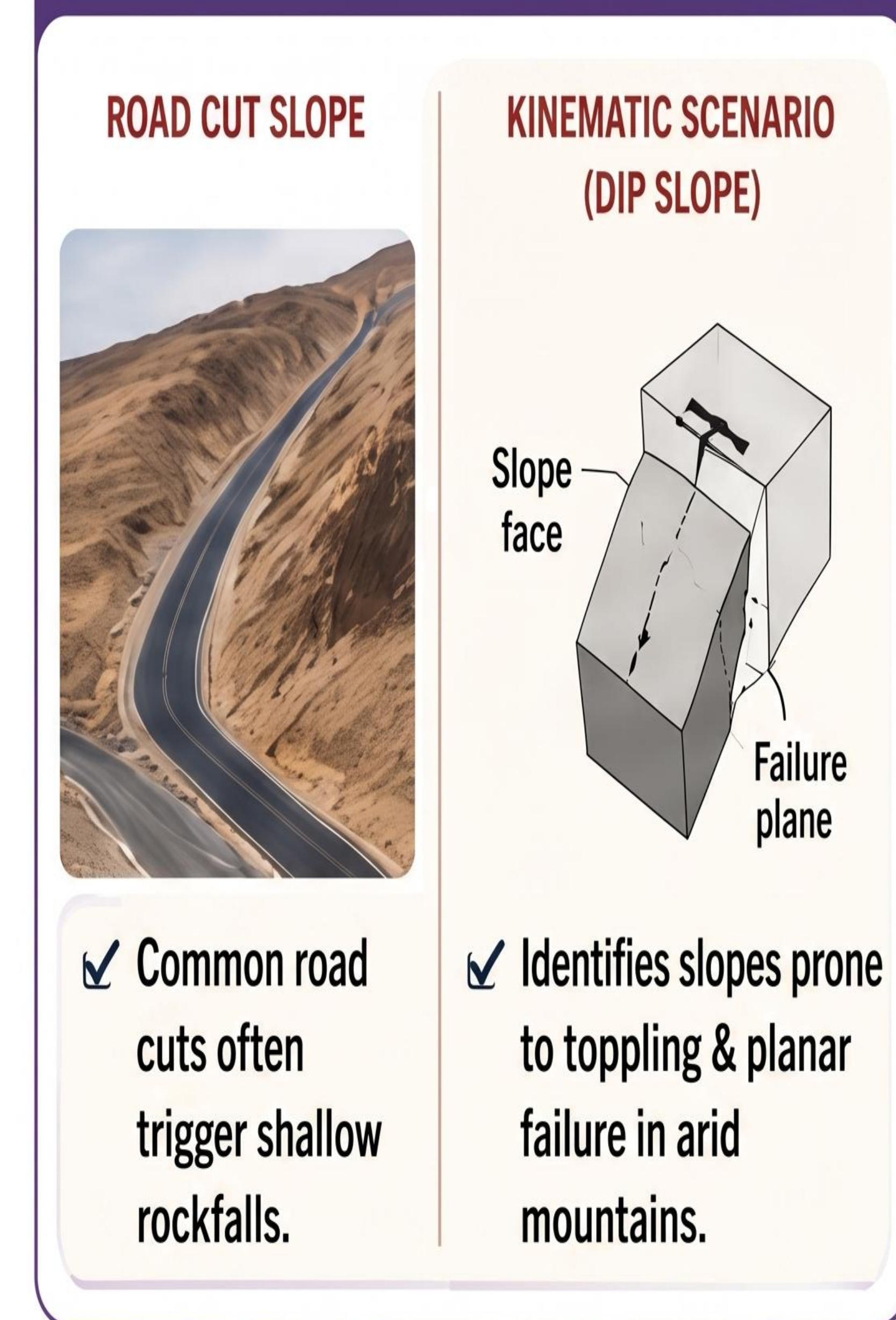
## 5. EXPLAINABILITY WITH SHAP



SHAP value (impact on model output): Positive (increases susceptibility) / Negative (decreases susceptibility)

Color scale (feature value): Low (blue) to High (red)

## 7. TYPICAL PROFILES



## 8. KEY RESULTS / SUMMARY

- Distance to roads is the most dominant anthropogenic driver.
- Aspect (20°–30°) controls kinematic daylighting failures.
- SHAP reveals model transparency and interpretability.
- SHAP-RFE selects the most relevant predictors.
- Optimal results achieved with 5–9 key predictors.

## 9. APPLICATIONS



## KEY MESSAGE

Explainable AI transforms black-box models into trustworthy decisions for landslide risk management

