

## Terrain Analysis via Supervised Machine Learning

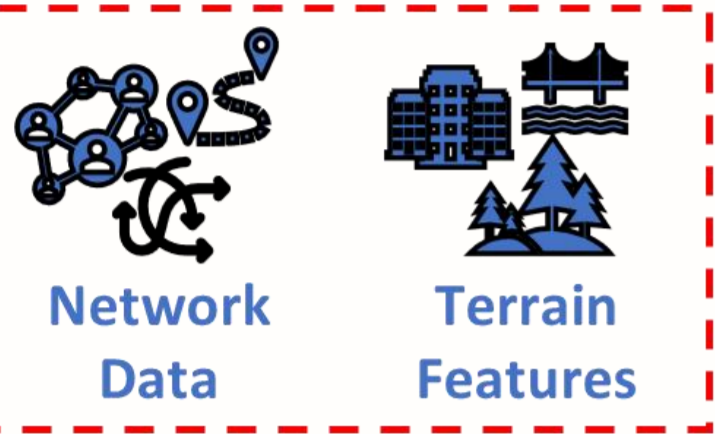
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### Bottom Line Up Front

#### Phase 1. Build & Optimize



Monte Carlo, Pyomo, Data Farm



Supervised Machine Learning Model

#### Phase 2. Input & Predict



Actual Data from RUW



Validate Predictions with actual RUW Strategies

**Motivation:** Traditional terrain analysis depends on manual overlays and heuristics, which are slow, subjective, and hard to scale. Modern conflicts with drones, improvised munitions, and multi-axis threats exploit terrain in ways that make axis bound logic inadequate.

**Objectives:** *This thesis builds a supervised ML framework that transforms GEOINT into terrain intelligence.* The model evaluated network analysis and terrain data to train a model to predict Ground of Tactical Importance (GTI). Validation against the 2022 Russia–Ukraine War (RUW) shows close alignment with Ukrainian defenses, demonstrating a scalable AI-enabled tool for decision support.

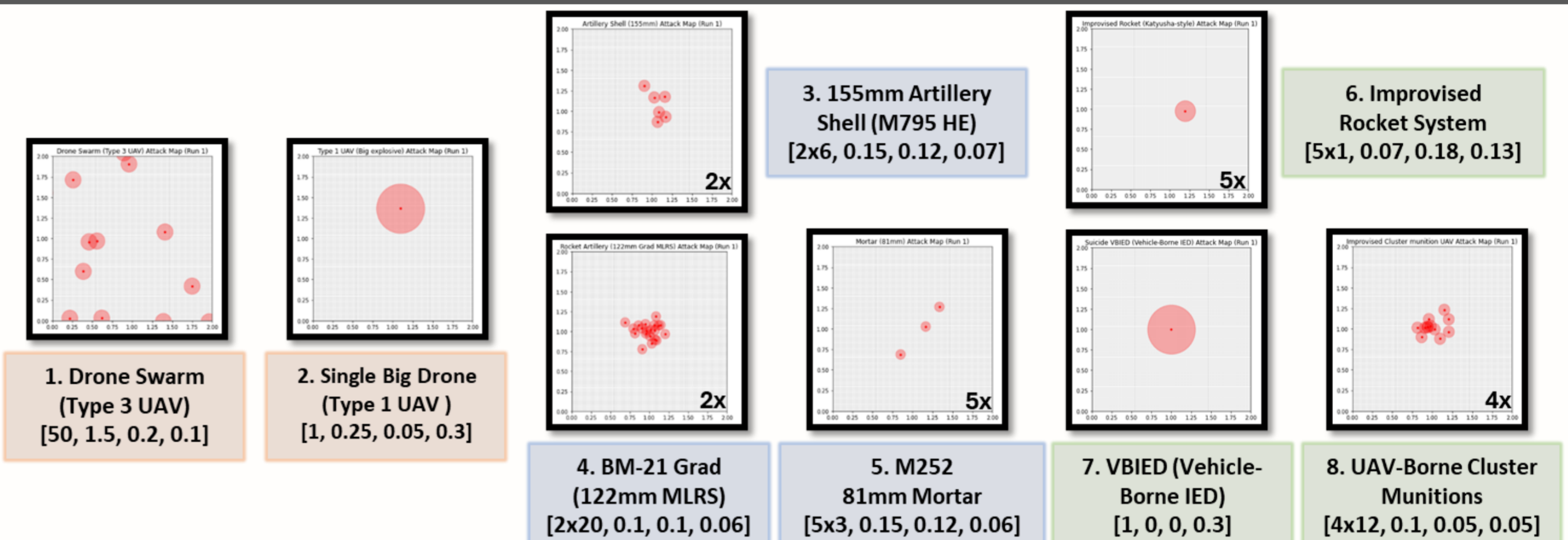
### Methodologies

### Results

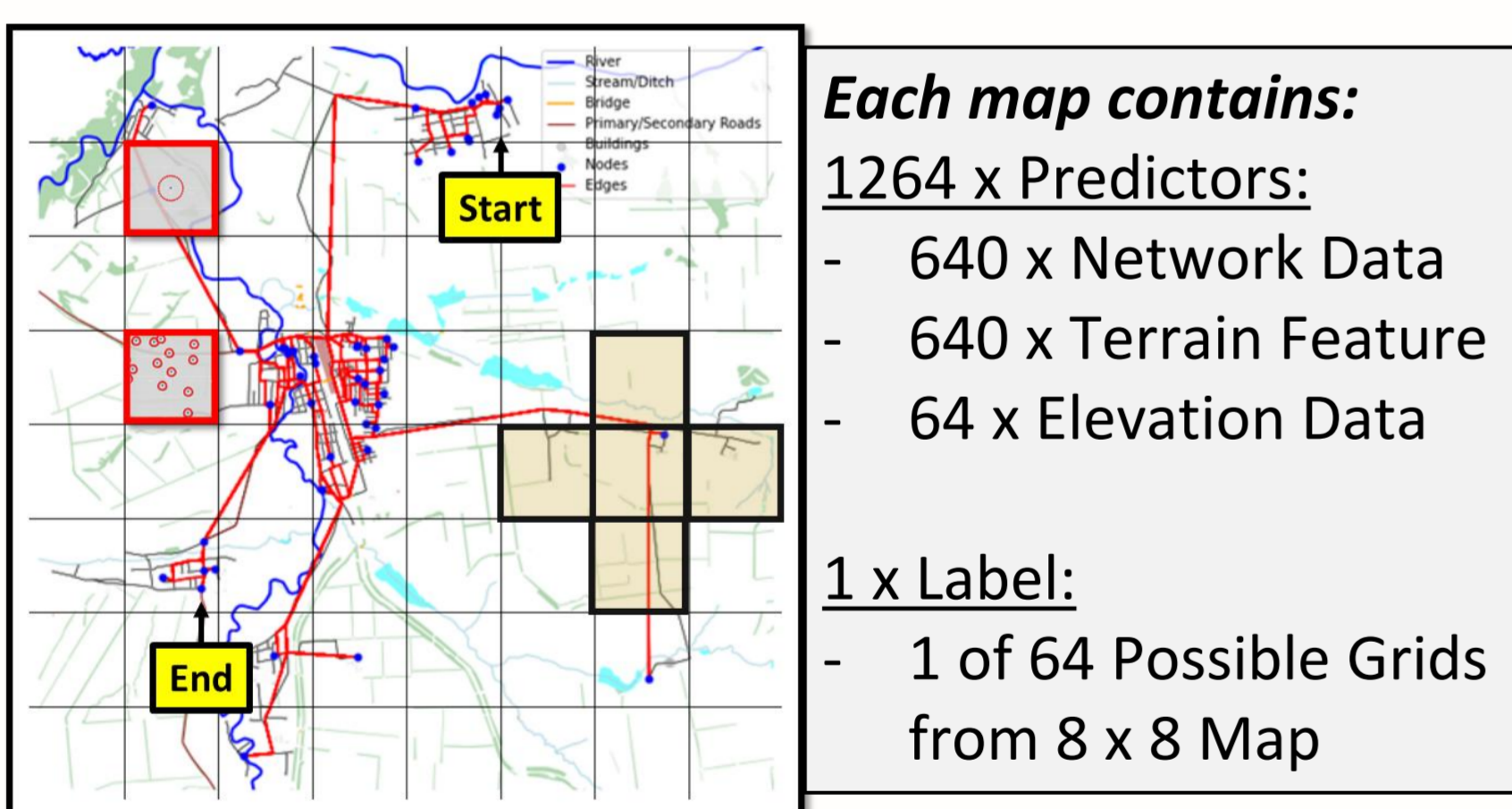
### Weapon Effects Simulation

Weapons are defined by four core attack parameters and modeled with Monte Carlo simulation to capture probabilistic effects. Nelson–Swan Group Selection (NSGS) technique is applied to evaluate and rank weapon effectiveness for interdiction.

Number of Attacks	Fire count per strike
Dispersion Radius	Mean spread of impact
Targeting Error	Randomness from inaccuracy
Blast Radius	Damaged radius



### Networks Interdiction & Supervised Machine Learning



#### Each map contains:

1264 x Predictors:

- 640 x Network Data
- 640 x Terrain Feature
- 64 x Elevation Data

1 x Label:

- 1 of 64 Possible Grids from 8 x 8 Map

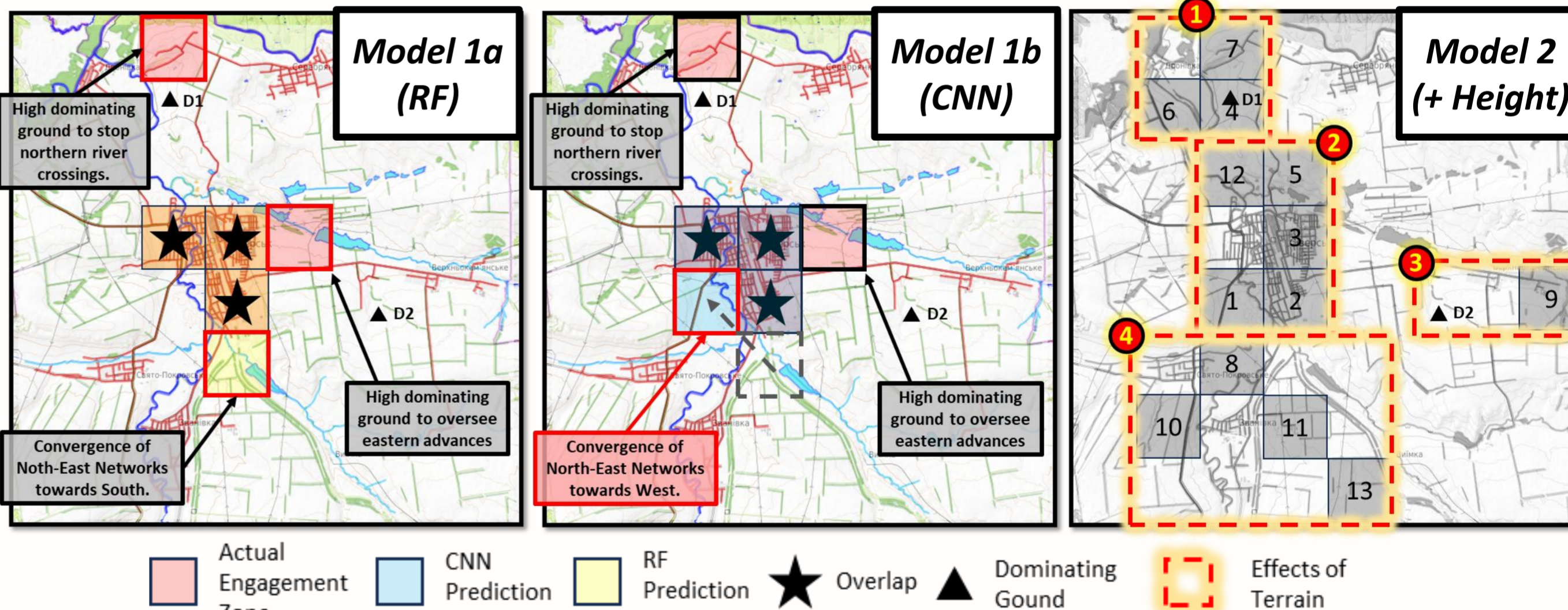
Predictors were extracted from OSMnx (road/terrain features) and USGS DEM (elevation).

Labels were generated by combining:

- Network Disruption:** Simulate weapon effects, compute shortest path via Bender's decomposition, and evaluate max-min operational cost.
- Terrain Advantage:** Weighted score of rivers, bridges, roads, urban areas, and elevation.

Both scores were normalized into an objective label.

The model was applied to Siversk in the 2022 RUW to test prediction fidelity.



The optimized model achieves **Classification Accuracy of Top-3: 61.7%; Top-6: 82.3%.**

- Model 1a/1b:** RF and CNN models predicted similar GTIs with one different grid based on alternate constriction points, confirming similar terrain-analysis logic.
- Model 2:** Adding elevation improved predictions, aligning it more closely with actual Ukrainian defenses (urban strongpoints, high ground, chokepoints).

(1) Dominating Ground D1 overlooks Siverskyi Donets River, offering observation and fields of fire, allows defenders to delay or block River Crossing Operations.	(2) Settlements in Siversk contain dense urban terrain that provides cover and concealment, allows defenders to use it as a strong defensive position.	(3) Dominating Ground D2 oversees a Type X route from East that supports speedy projection of forces, allows defenders to observe and interdict enemy maneuver.	(4) Constriction Points where the roads lead into the depth of Ukraine from the East and West, allow defenders to observe and interdict enemy maneuver.
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### Future Work

- Predictors:** Transform military GEOINT archives into actionable terrain features.
- Labels:** Derive CONOPS-based information from Operational Plans to train the Supervised Model with Military CONOPS Logic.