



Full Framework Overview (English Version)

The model is trained using **general, global oil-spill scenarios** (various oceans + numerical simulations).

Taeon is *not* used for training — instead, it is used afterward as a **hold-out test case** to evaluate how well the model generalizes.

The three-model architecture remains the same:

1 Prediction Model

Purpose

Predict how oil will spread over the next N hours anywhere in the world, given currents, wind, oil type, and initial conditions.

Training Data

- Large-scale numerical simulations and historical spill events
 - NOAA GNOME, OpenDrift, global reanalysis data
 - Diverse coastlines, depths, tides, winds, and oil properties
- Important:
 - **Taeon conditions are intentionally excluded from training**
 - Taeon acts as an unseen test region

How Taeon is used

- After training is complete:
 - Feed Taeon coordinates
 - Reanalysis wind/current fields for 2007

- Initial spill location and volume
 - The model makes predictions **without any retraining**
 - We then compare:
Real Taaan spread pattern vs. Model prediction
→ Evaluating generalization (zero-shot performance)
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2 Response / Strategy Model

Purpose

Given the predicted spread, determine *when / where / how much cleanup resources* should be deployed.

Training Data & Environment

- Again, trained on non-Taaan scenarios
- Numerous synthetic episodes:
 - “Resource allocation → resulting oil spread”
- Methods:
 - Reinforcement learning environments
 - Numerical diffusion models
 - Mixed policies (random / heuristic / optimized) to generate diverse experience

How Taaan is used

- Taaan spill is fed into the trained models:
 - Prediction model outputs future spread
 - Response model proposes:
 - Where to send vessels

- Where to place booms
 - Where to apply dispersants
- Comparison:
 - **Actual 2007 Taaen response vs AI-derived response**
- Both strategies are passed into the recovery model to compare outcomes

Note:

You may adjust Taaen-specific environmental parameters,
but **you do NOT fine-tune or retrain the model using Taaen data.**

3 Cleanup & Ecological Recovery Model

Purpose

Estimate long-term environmental differences between cleanup strategies (actual vs AI-based):

- Residual oil quantity
- TOC / DO levels
- Biological recovery time
- PAH/TPH degradation
- Formation of MOS/MOSS, etc.

Training Data

- Long-term datasets from various regions:
 - Post-cleanup DO/TOC time series
 - Residual oil decay
 - Bio-degradation rates

- Ecological stress indicators
- None of these include Taeon

How Taeon is used

- Inputs:
 - Taeon initial state + “Actual 2007 response”
 - Taeon initial state + “AI response”
- Outputs:
 - Month-1 / Month-6 / Year-1 recovery index
 - Total recovery time
- From this:
 - Statements such as
**“With AI involvement, recovery time could be reduced from X years → Y years
 and contaminated area reduced by Z%.”**

Summary: Taeon’s Role = Testbed (Not Training Data)

Training phase

- Global + multi-region numerical simulations
- Taeon intentionally excluded to avoid overfitting
- Ensures the model learns *general principles* rather than region-specific quirks

Evaluation phase

- After training, Taeon is used as an unseen example

- Zero-shot / domain-generalization test
- We compare:
 - Spread accuracy
 - Response optimization
 - Long-term recovery differences

Outcome

- Shows that
 **“Our model is not overfitted to a single region;
 it generalizes across diverse marine environments.”**