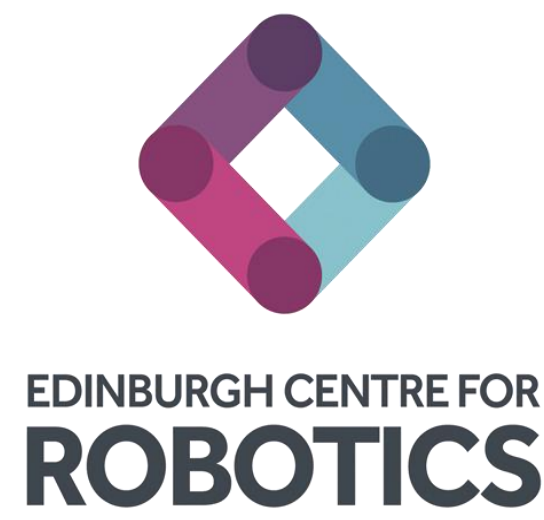


A Surrogate Model Framework for Explainable Autonomous Behaviour



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Abstract

Adoption and deployment of robotic and autonomous systems in industry are currently hindered by the lack of transparency, required for safety and accountability. Methods for providing explanations are needed that are agnostic to the underlying autonomous system and easily updated. In this work, we use surrogate models to provide transparency as to the underlying policies for behaviour activation. We show that these surrogate models can effectively break down autonomous agents' behaviour into explainable components for use in natural language explanations.

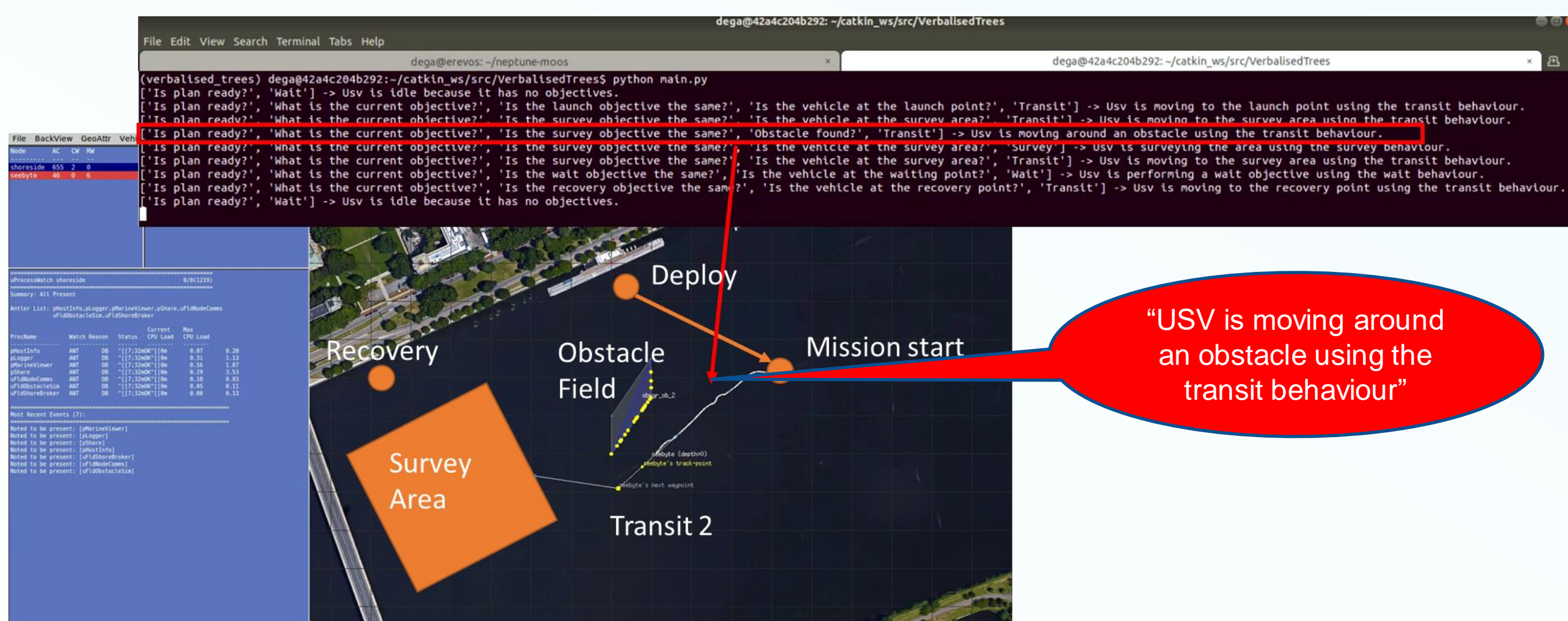
Introduction

1. There is a need to convey rationale behind the decision-making of an autonomous system to operators.
2. We explore various vehicles that can exhibit a variety of behaviours, including:

1. Autonomous Underwater Vehicles (AUVs) for pipeline inspection and Autonomous Surface Vehicles (USVs) for surveys at designated areas.

3. With this work, we attempt to answer the following **research questions**:

1. **RQ1**: How robust are surrogate models in policy approximation for behaviour activation?
2. **RQ2**: Can these surrogate models be used to effectively generate explanations?
3. **RQ3**: How is the performance affected when going from simulated data to real trials with real vehicles tested in a realistic environment?



Results

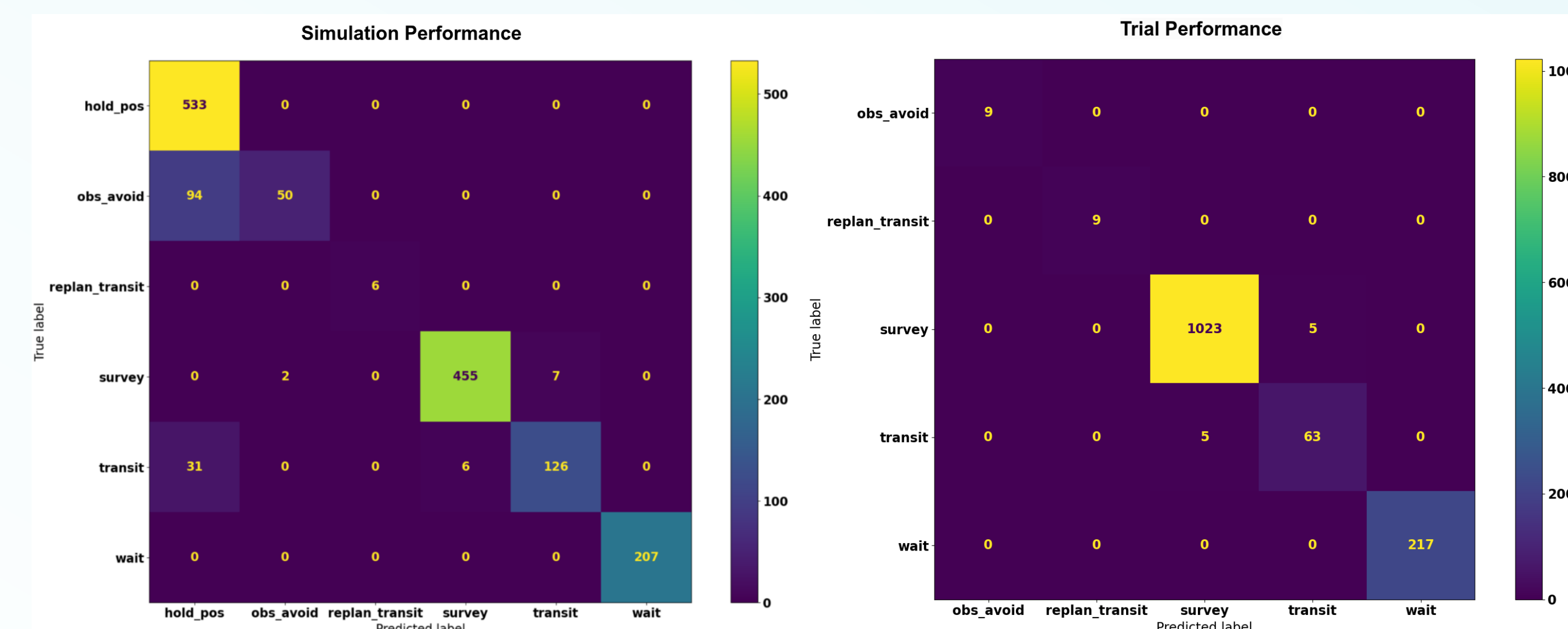
RQ1: Model Selection Metrics

Models	Accuracy	Precision	Recall	F1-Score	Fit Time	Score Time
Decision Tree	0.8981	0.9464	0.8498	0.8712	25.0936	0.0025
CategoricalNB	0.8247	0.8701	0.8616	0.8379	0.1127	0.0019
KNN	0.6655	0.7806	0.8291	0.6953	4.8554	0.0721
SVM	0.8846	0.9163	0.8378	0.8535	14.9298	0.0547
Multi-layer Perceptron (MLP)	0.8987	0.9459	0.8496	0.8707	147.8816	0.0075

RQ2: Comparison of intrinsic features and Shapley Values



RQ3: Behaviour Predictions

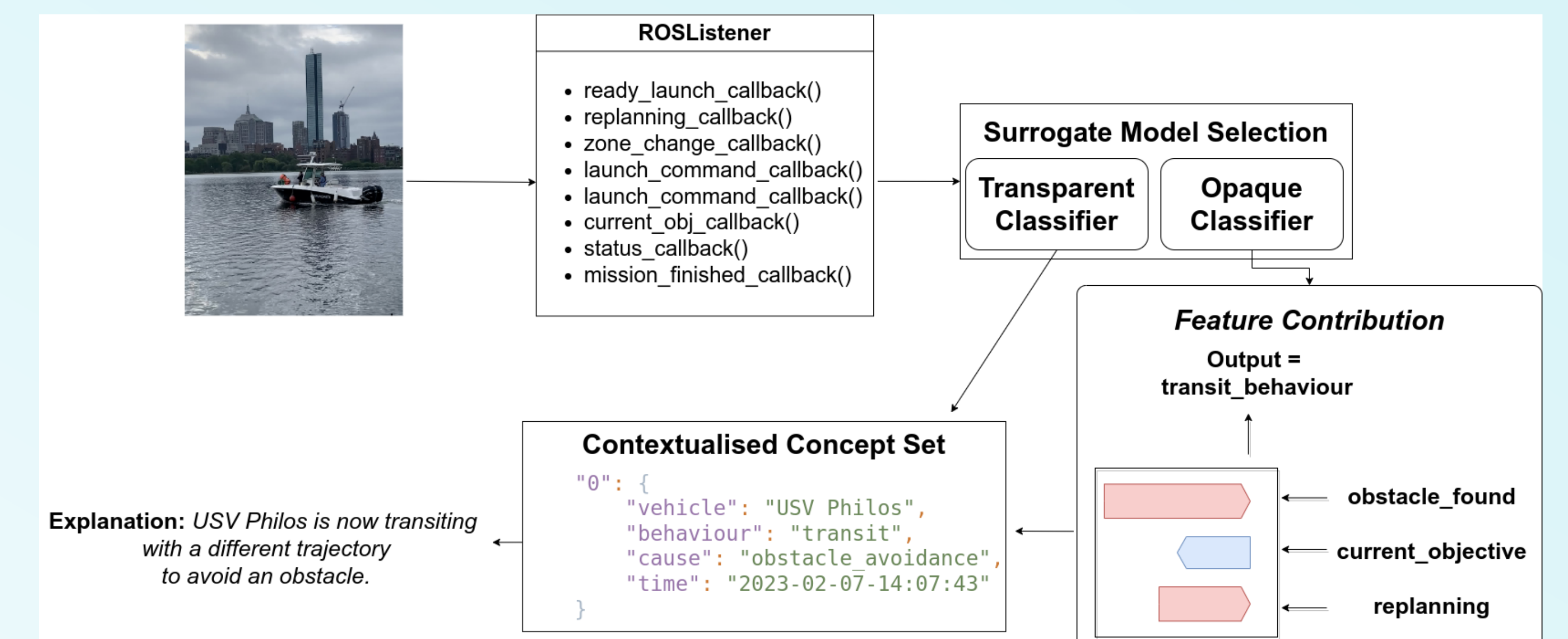


For **RQ1**, we demonstrate the classification performance across multiple models and select the best for our use case (Decision Tree). For **RQ2**, we made a comparison of intrinsic features within our surrogate model and corresponding Shapley Values to compare each estimated causality. Finally, for **RQ3** we present classification performance for both simulated and real missions.

Methodology

The proposed framework consists of the following steps:

- 1) Extraction of vehicle states for surrogate model training and real-time explanation generation.
- 2) Model Selection with Nested Cross-validation to choose optimal Surrogate Model.
 - 1) Use of intrinsic features for transparent models or feature contribution estimation for opaque models.
- 3) Representation of exhibited behaviours with Contextualised Concept Sets.
- 4) Use of concept sets to generate Natural Language Explanations.



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We would like to thank MIT's AUV Lab and Laurence Boe from SeeByte Ltd for their assistance with the simulator and the real trial. This work was also funded and supported by the EPSRC Prosperity Partnership (EP/V05676X/1), the UKRI Node on Trust (EP/V026682/1), EPSRC CDT on Robotics and Autonomous Systems (EP/S023208/1), and Scottish Research Partnership in Engineering.

Contact Info



Conclusion & Future Work

1. A domain-agnostic framework for approximating behaviour activations and replanning of an autonomous agent with classification models has been introduced.
2. Our approach is capable of discovering the causality of autonomous decisions and storing that information with Contextualised Concept Sets.
3. Moving forward, we plan on using these representations to investigate data-driven language explanations such as large language models.
4. Further evaluation of explanations is also required to examine the capacity of our approach to disambiguate robotic behaviours.