# **Literature Survey**

## **Review of Existing Systems:**

Sr. No.	System/ Paper	Authors/ Source	KeyFeatures	Limitations
1	Automating the Detection of Oil Spills from Sentinel-1 SAR Imagery Using Deep Learning [Thesis].	Al Hashmi, M. (2023)	Uses <b>deep learning</b> techniques for automating oil spill detection.	Being a thesis, might have limited peer review compared to journal articles.
2	Fast detection of oil spills and ships using SAR images	Lupidi et al. (2017)	Uses <b>SAR images</b> which are suitable in cloudy or night-time conditions.	
3	Oil spill detection from Synthetic Aperture Radar Earth observations: a meta-analysis and comprehensive review	Jafarzadeh et al. (2021)	Provides a <b>comprehensive review</b> of SAR-based oil spill detection.	Does not propose a new method— purely a review.
4	Furuno's Operator Manual for U- AIS Transponder	https://www .furuno.com/ en/support/ manuals/	Provides <b>technical specifications</b> and operational guidelines for AIS devices.	Not academic or analytical—more of a <b>technical manual</b> .
5	A new ship tracing technology from oil spills based on multi- source data	D. Luo et al.	Suggests use of multi-source data fusion for tracing ships responsible for spills.	Unclear if peer- reviewed or published.

### **Limitations of Existing Systems:**

#### • High False Positives & Limited Accuracy

Traditional SAR-based methods often misidentify natural look-alikes (e.g., low wind areas) as oil spills, leading to unreliable results (Lupidi et al., 2017; Jafarzadeh et al., 2021).

#### • Lack of Automation & Data Integration

Many systems do not fully utilize deep learning or multi-source data (e.g., AIS), resulting in poor scalability and limited ability to trace spill sources (Al Hashmi, 2023; Furuno Manual).

#### • Poor Generalization & Standardization

Models often fail to perform consistently across different regions due to environmental variability and lack standardized datasets for evaluation (Jafarzadeh et al., 2021).