

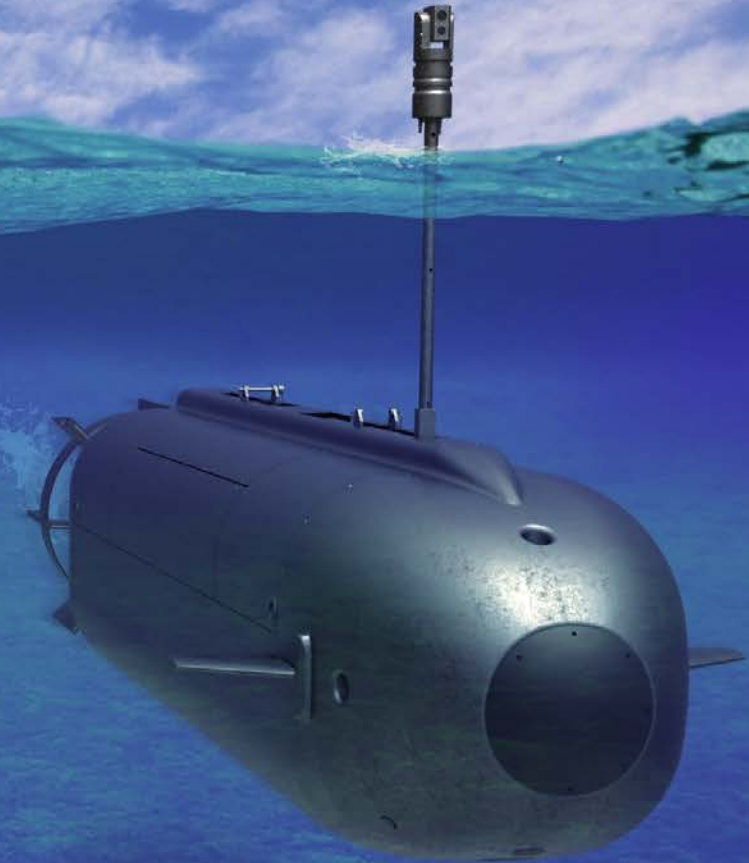
Smart Maritime Surveillance System



Valerio Imbriolo
MATLAB EXPO 2021

Table of Contents

- Introduction
- ARTS
- Process & Tools
- Results
- Conclusions & Future Developments



About Me



Role
Command & Control Systems Manager

Education
University of Pisa
Robotics and Automation Engineering

Contact
Email: valerio.imbriolo@drass.tech
Mobile: +39 366 6358872

- **MATLAB**[®] user since 2010.
- My team and I currently use **MATLAB**[®] and **Simulink**[®] as base benchmark for our ideas both in DEFENCE and INDUSTRIAL DIVING business lines:
 - Dynamic Simulation
 - Control Development
 - Artificial Intelligence

DRASS Company

DIVING & SIMULATOR

- Midget Submarine
- Compact Submarine
- Swimmer Delivery Vehicle



- Deep Diving
- Surface Diving
- Modular Systems

DEFENCE SOLUTIONS

SUBMARINE RESCUE SYSTEMS

- Spares and Components
- Repair and Maintenance
- Training Centre



- Resident Rescue System
- Deployable Rescue System
- Submarine Ventilation

GLOBAL SERVICE

ENGINEERING & TESTING

- Tunneling Hyperbaric Chambers
- Industrial Pressure Vessels



- Consultancy
- Laboratory and Experimental

INDUSTRIAL

MEDICAL HOT SERIES & SCUBA



- Medical Hyperbaric Chambers
- Scuba Diving Chambers
- Diving Boats

Takeaways

- Speed of dataset creation and management for training and validation
- Easier Project Exploration (Rapid function implementation & function workflow analysis)
- Automatic code generation to quick prototype and integration with external environment
- The power of collaboration with MathWorks

ARTS

Augmented Reality Tracking System is an optronic system software for maritime surveillance capable of:

Video Flow Stabilization



Obstacle Awareness



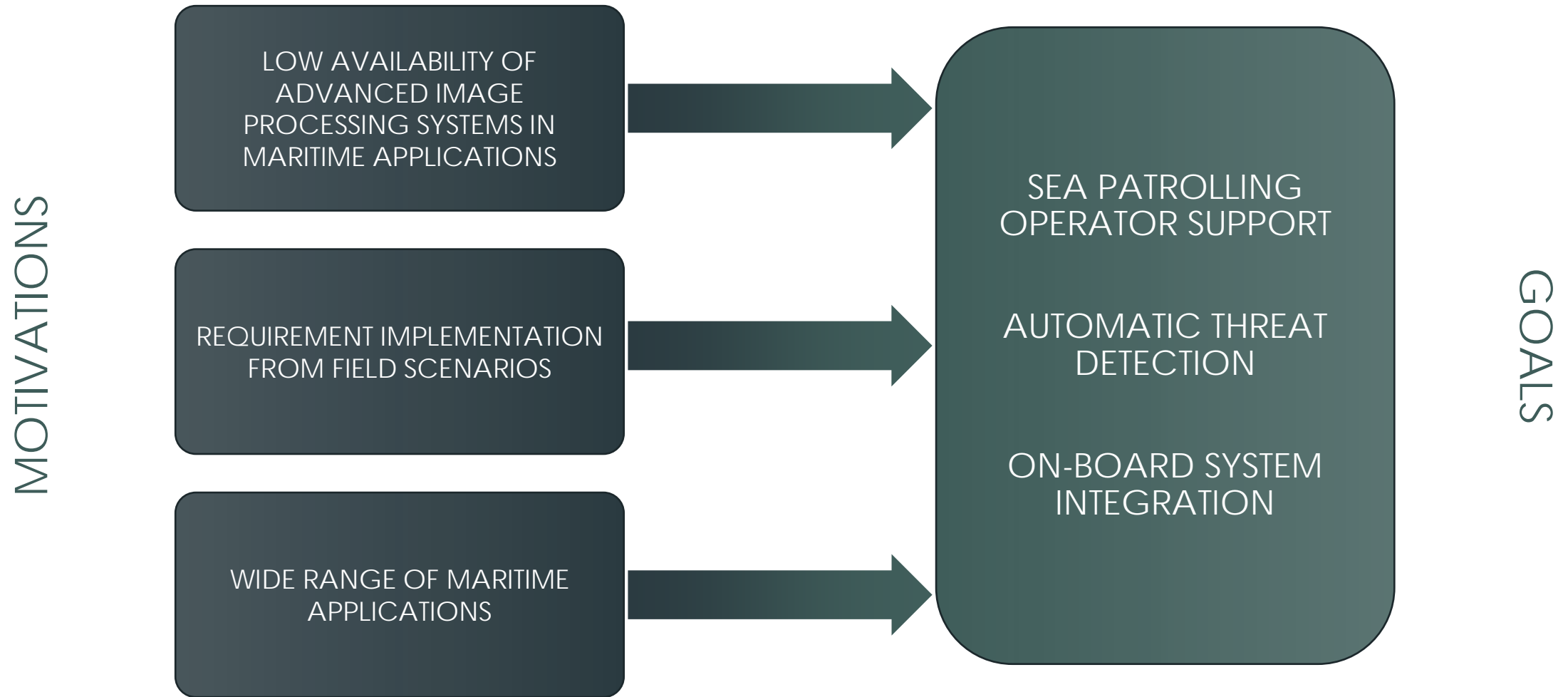
Object Detection



Multi-Spectra Video Merging



Motivation & Goals



Challenges & Solutions

OPEN SEA ENVIRONMENT

Wave Motion
Continuous background changes
Almost total lack of reference points



STATE OF THE ART ALGORITHMS

Two different stabilization algorithms
Tests of different detectors
Frames preprocessing



REAL-TIME PROCESSING ON TWO VIDEO FLOWS



ALGORITHMS OPTIMIZATION

Parallel computing toolbox
Project dependency analysis
MATLAB® Profiler
Slow dynamic speed



TIGHT DEADLINE MATLAB® CODE INTEGRATION IN EXTERNAL PROJECT

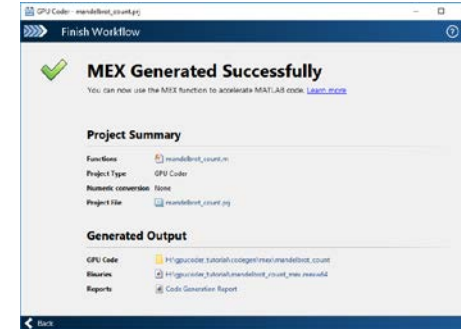


GPU Coder & MATLAB Coder MathWorks SUPPORT



Process & Tools

- Wavelet Toolbox™
- Image Processing Toolbox™
- Deep Learning Toolbox™
- Computer Vision Toolbox™
- Signal Processing Toolbox™



IMPLEMENTATION

RESULT COMPARISON
BY MEANS OF KPI

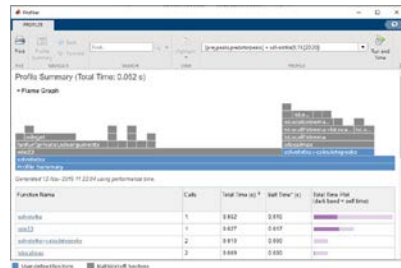
CODE GENERATION

V&V

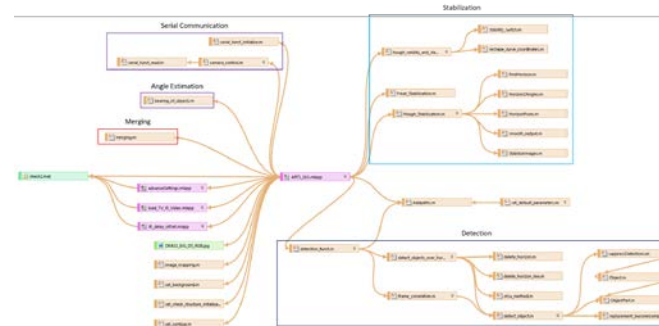
OPTIMIZATION

FUNCTION WORKFLOW
MANAGEMENT

IMPLEMENTATION IN
EXTERNAL PROJECT



Parallel Computing Toolbox



Implementation: YOLOv2 Detector

Why YOLOv2?

- Traditional detection methods are not suitable for unstructured environment
- Its competitors, e.g. R-CNN, are not adequate for real-time processing
- Gives great results in different domains

~5000 images/dataset

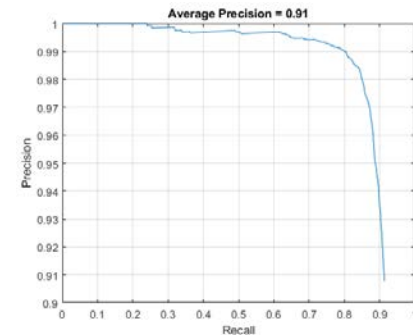


~30000 images/dataset

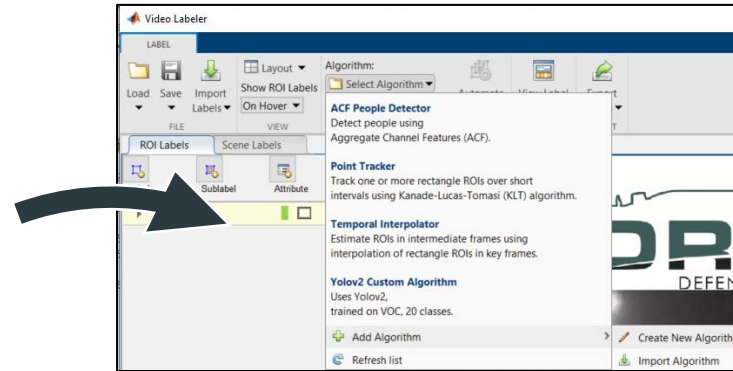
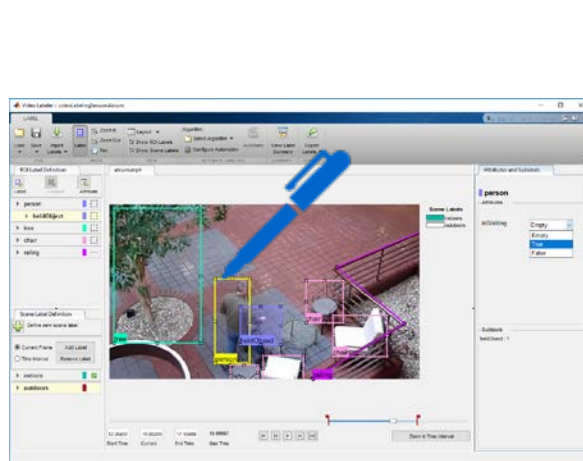


~12h/training

Training Option
Variation
(Learning rate,
Epochs,
Algorithms)



Speed Up the Process: Database Creation



Name	Size
Folder	
+vision	
+labeler	
YoloCustomAlgorithm.m	12 KB
Function	
darknet2matlab.m	2 KB
Script	
recreateTiny_yolov2.m	8 KB
MAT-file	
params_2019_11_14_15_22_05.mat	56,28 MB
tiny_yolov2Detector.mat	56,29 MB
ONNX File	
Model.onnx	60,54 MB

Manual Video Labeling

Training rough YOLOv2 detector

Use Detector in Video Labeler app

Manual incorrect frames discharging

Checking and discarding incorrect GTs

Time only manual labeling
3 min/frame x 5000 frame ~ = 249 hrs

vs.

Above process with detector obtained after 16 hrs of work
0,3 sec/frame x 5000 frame ~ = 30 min
30 sec/frame x 5000 frame ~ = 42 hrs

Speed Up the Process: Intra-Team Exchanges

Performance
Evaluation TEAM

MATLAB® and
Simulink® TEAM

Visual Studio TEAM

LIVE SCRIPTS

YOLOv2 from ResNet-50

Create a YOLOv2 network from a ResNet-50 CNN.

Load Dataset

Load the data set extracted from the Singapore clips

```
load('vessel_table.mat')
vehicleDataset = trainingTable;
clear trainingTable
```

The vehicle data is stored in a two-column table, where the first column contains the image file paths and the second column contains the vehicle bounding boxes.

```
% Display first few rows of the data set.
vehicleDataset(1:4,:)
```

```
ans = 4x2 table
      imageFilename    Vessel
   1  'TrainingData...'  5x4 double
   2  'TrainingData...'  5x4 double
   3  'TrainingData...'  5x4 double
   4  'TrainingData...'  5x4 double
```

Split the data set into a training set for training the detector, and a test set for evaluating the detector. Select 60% of the data for training. Use the rest for evaluation.

CODE GENERATIONS

- GPU coder automatically generates from the same function generic .dll or specific .cu code, NVIDIA hardware is available
- The code and the necessary libraries can be integrated with no effort in Visual Studio thanks to *packNGo* MATLAB® function

Video Stabilization

Original Source

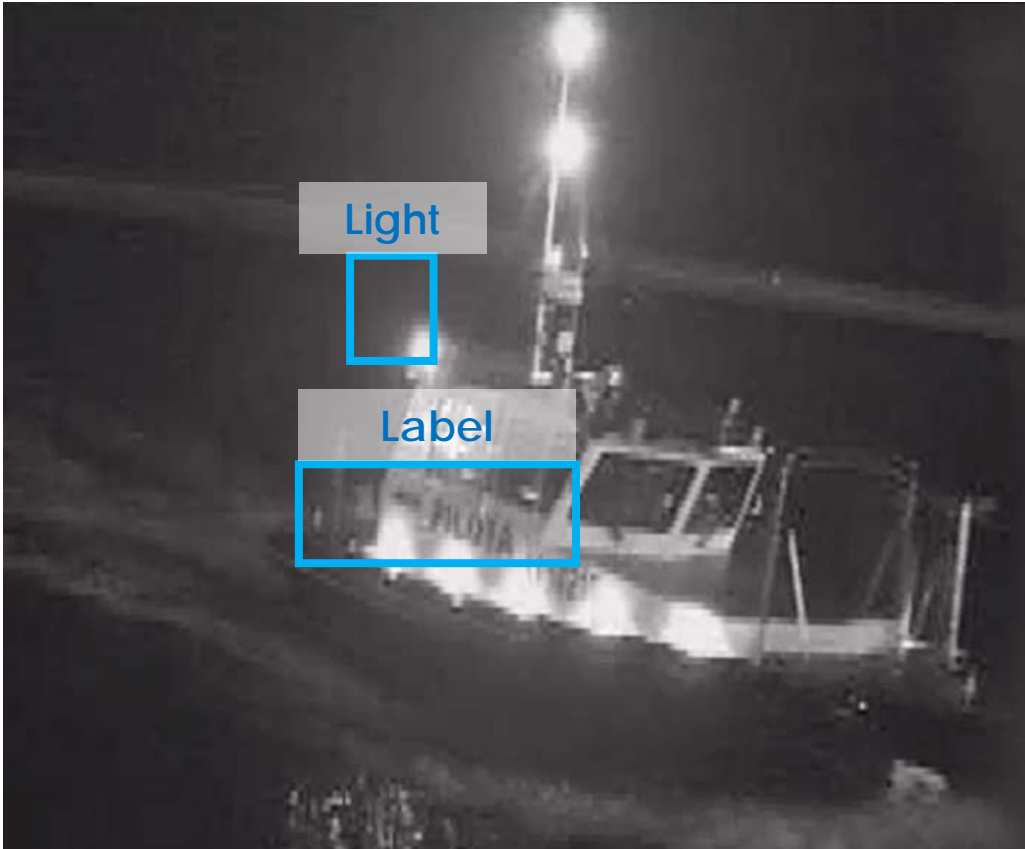


Stabilized Source

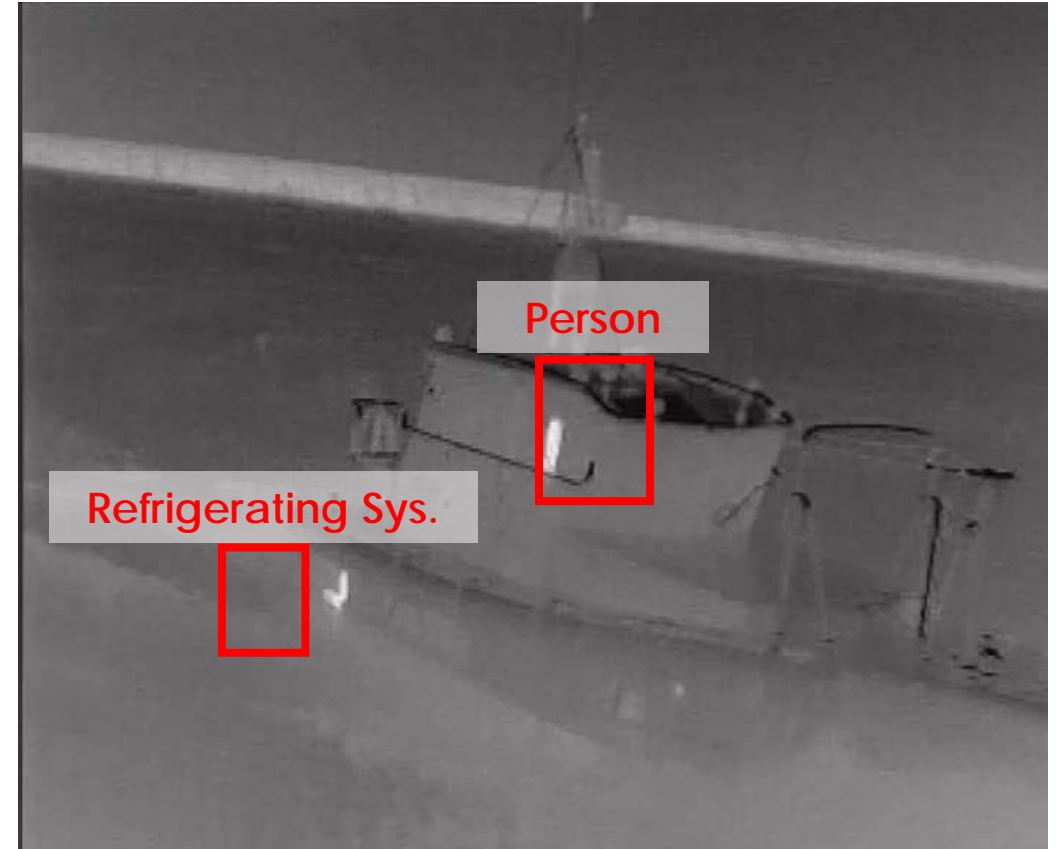


Sensors Merging

Daylight Camera

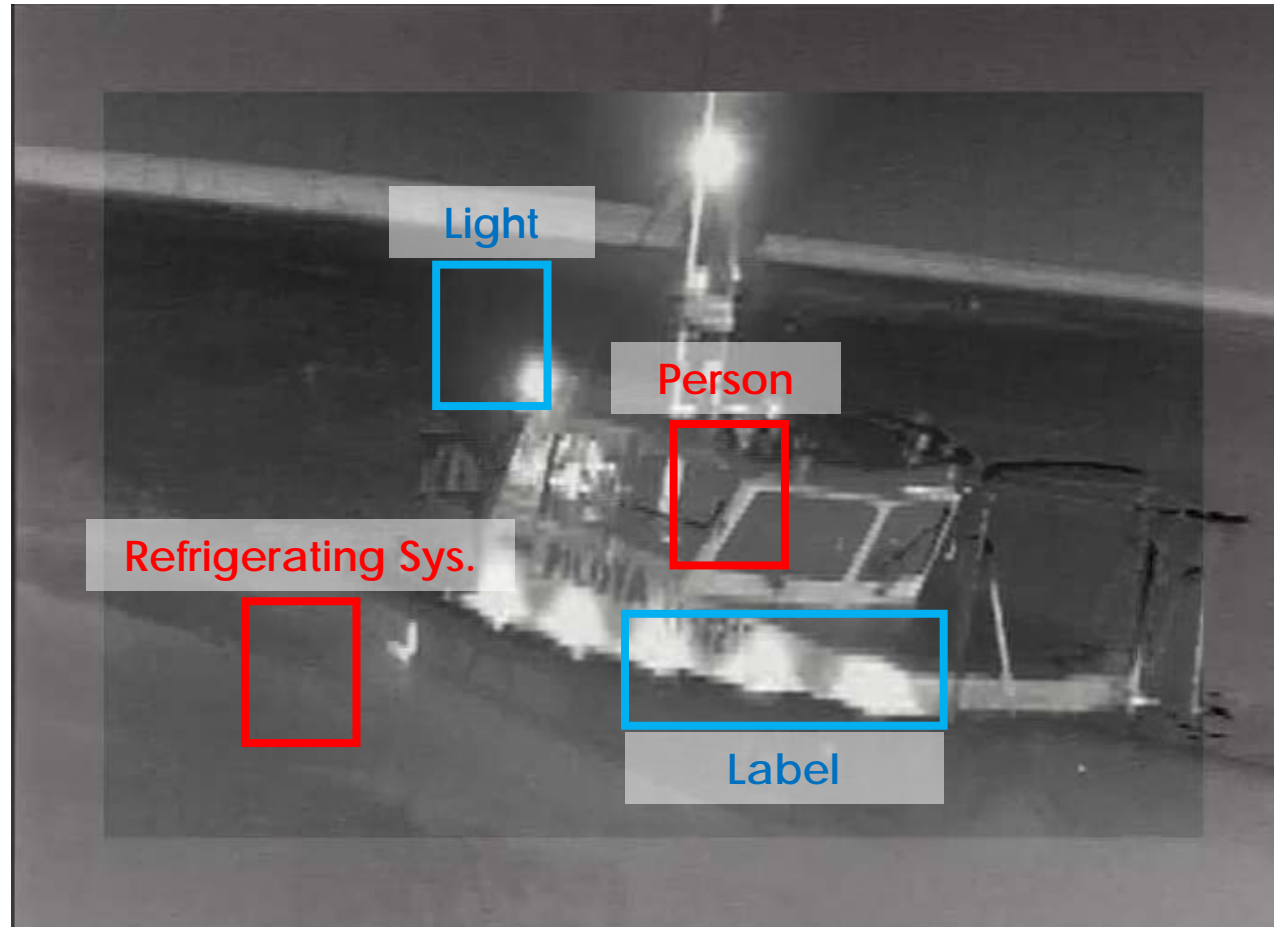


IR Camera



Sensors Merging

Merging



Tracking & Obstacle Awareness

Container Ships Detection & Tracking

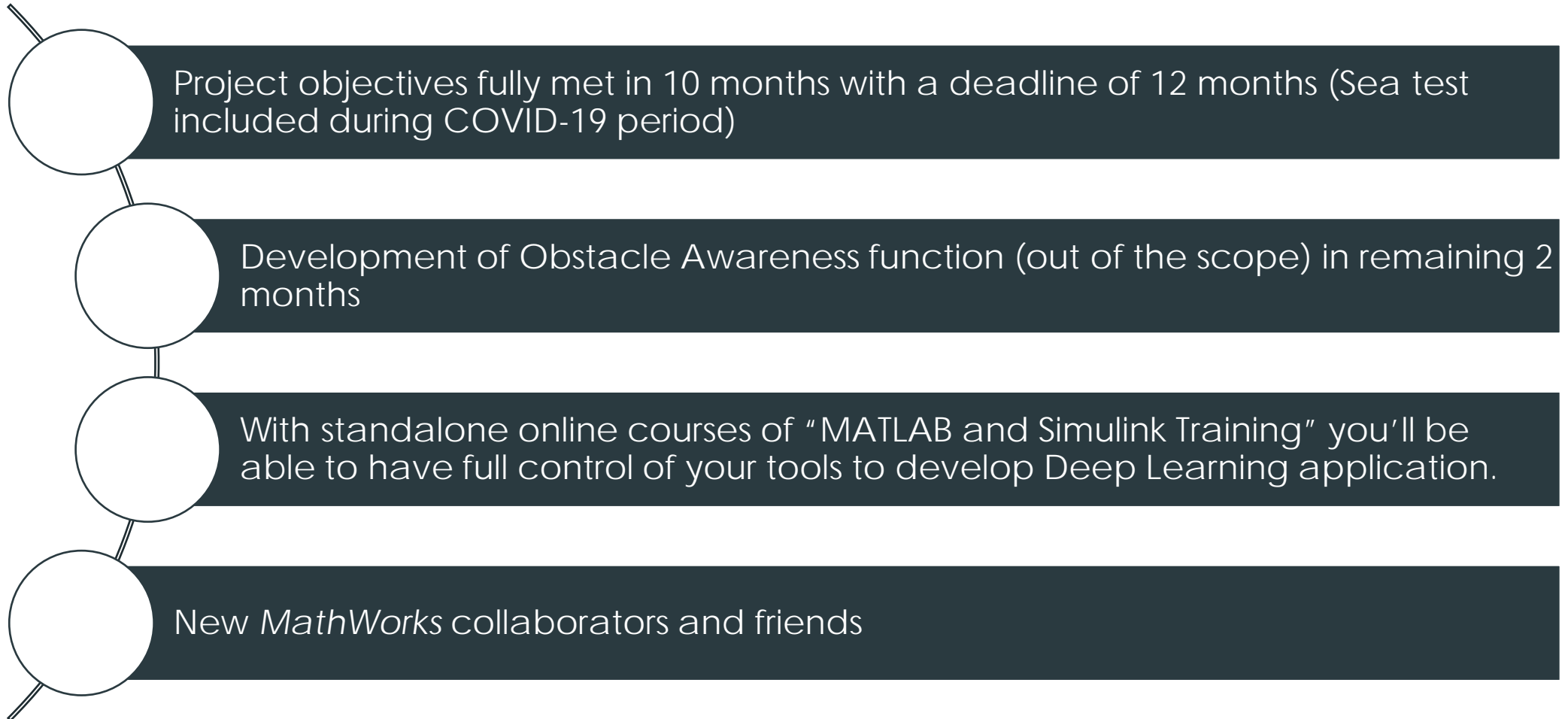


Source Video from Singapore Maritime Dataset (SMD)

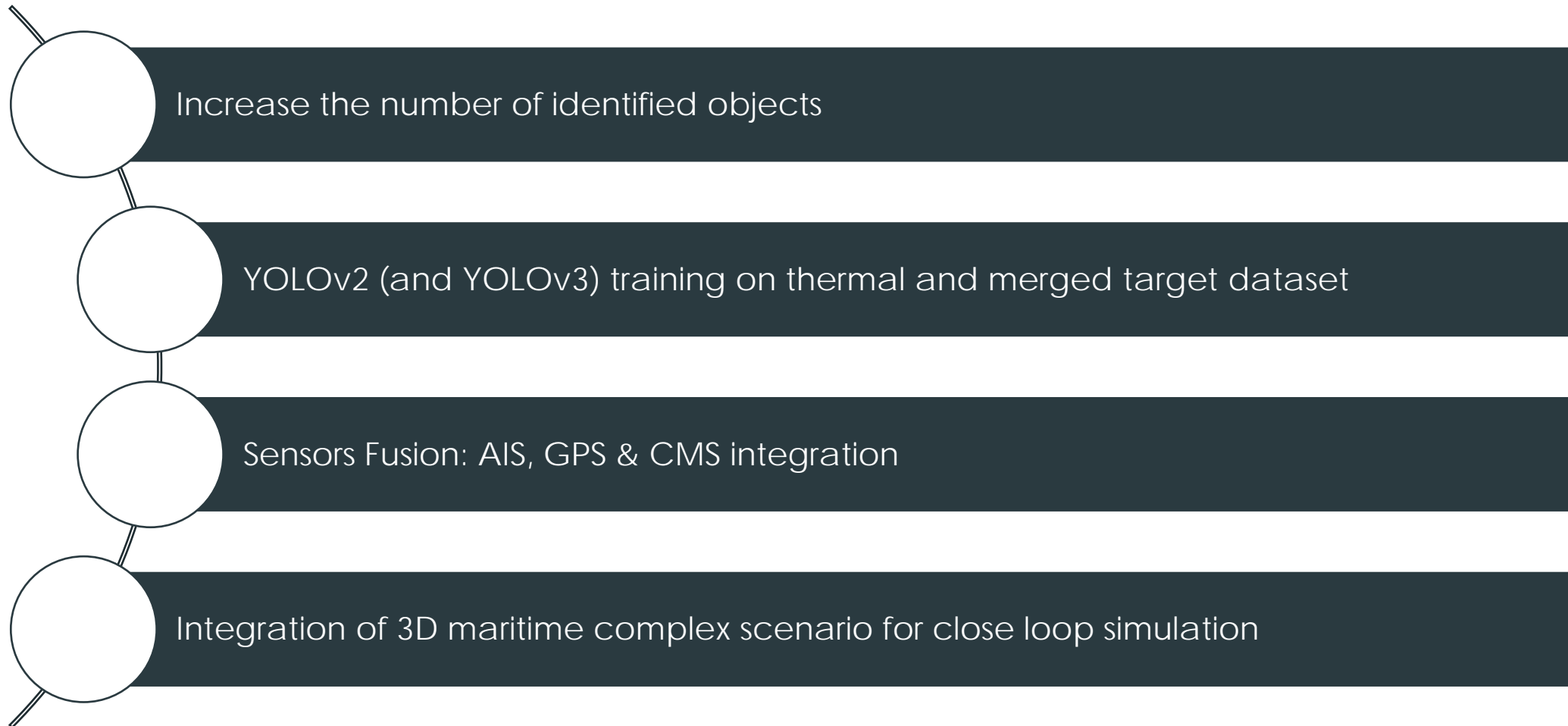
Obstacle Awareness



Conclusions



Future Developments



Thank You for your kind attention

