

CAPSTONE PROJECT

Oil Spill Detection from Satellite Imagery

WARINTORN NAWONG



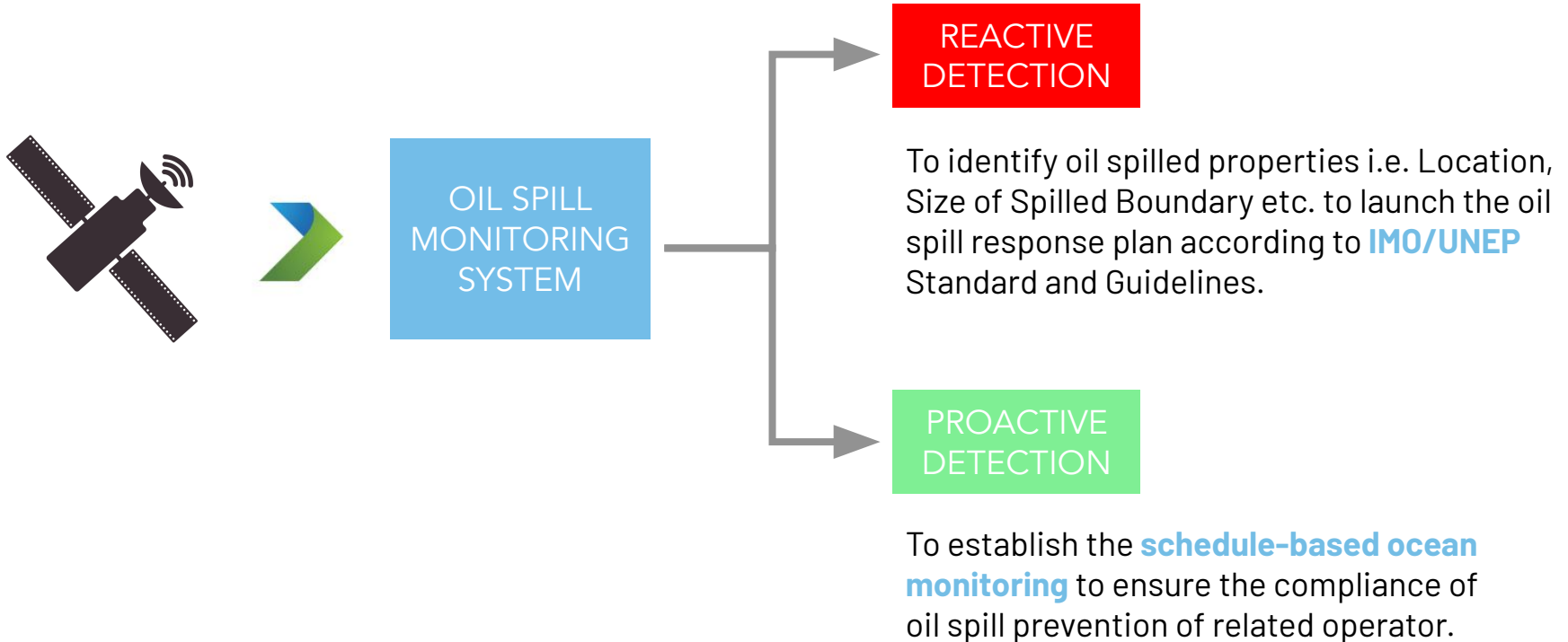


AGENDA

- **Problem Statement**
- **Introduction of Oil Spill**
- **Introduction of Satellites Imagery**
- **Model Development Journey**
- **Conclusion**

PROBLEM STATEMENT

To develop OIL Spill Detection from **Satellite Images** in order to build spilling detection system.



THE CONSEQUENCE OF OIL SPILLS



Environmental Impacts

Oil spill events caused damage to **wildlife, marine ecosystems**, and **coastal environments**.



Economical Impacts

Oil spills can lead to severe disruption of the **tourist industry**.



Human Impacts

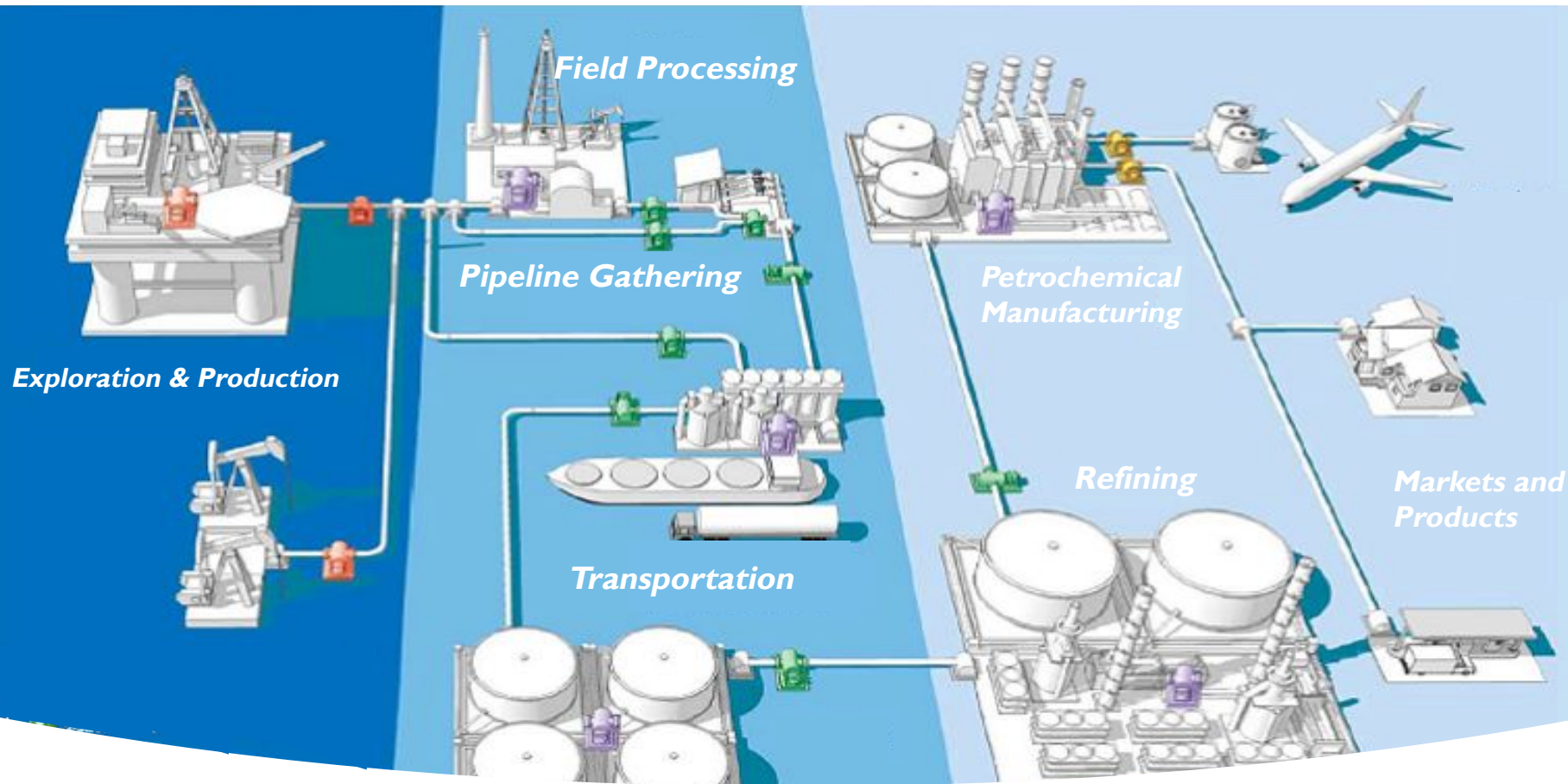
Oil spills impact on human health, including **respiratory**, and **immune system** damage.

OIL SUPPLY CHAIN

Upstream

Midstream

Downstream



CAUSE OF OIL SPILL



Oil Tanker/Transport Vessels

≈ 50.0 %

of all oil spills are directly or indirectly caused by **human error**.



Oil Rig/Platform

≈ 40.0 %

of oil spills are attributed to **equipment failure or malfunction**.



Pipeline



Storage tanks

≈ 21.0 %

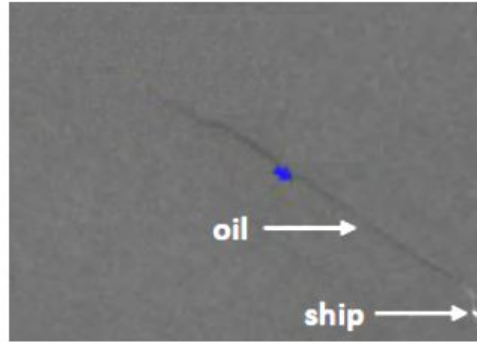
of oil releases are from **Operational discharges** from vessels.

OIL SPILL CHARACTERISTICS

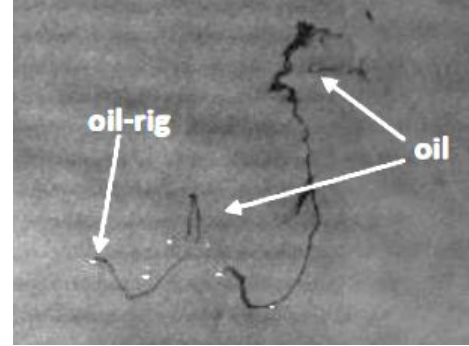
Operational Discharges

- Tank-washing Procedure
- Platform-sourced pollution

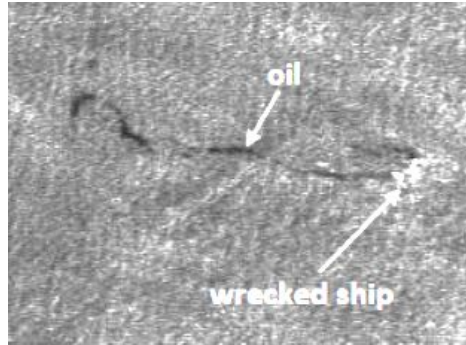
Straight Linear



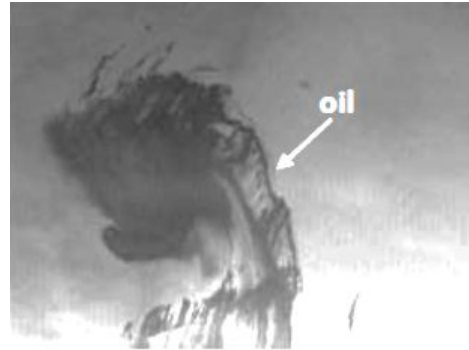
Curvilinear



Discontinuous Patches



Rounded Shape



Accidental Discharges

- Ship Accident
- Platform Accident

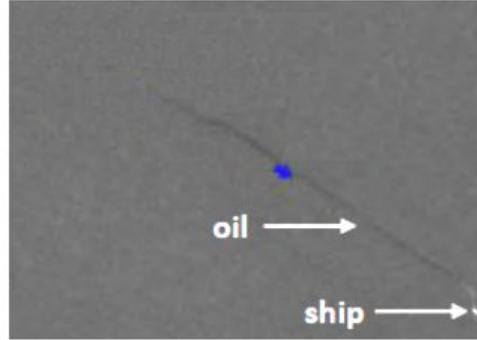
OIL SPILL CHARACTERISTICS

LINE SHAPE

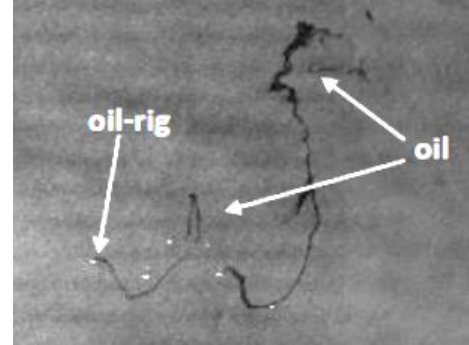
Operational Discharges

- Tank-washing Procedure
- Platform-sourced pollution

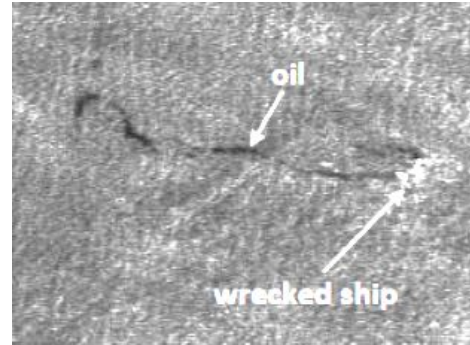
Straight Linear



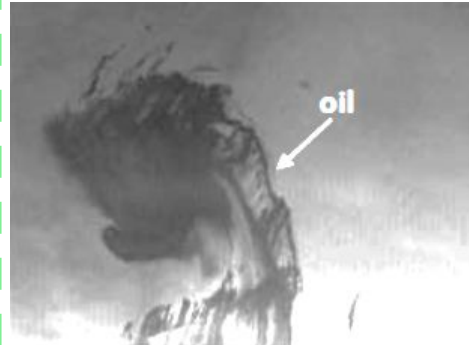
Curvilinear



Discontinuous Patches



Rounded Shape

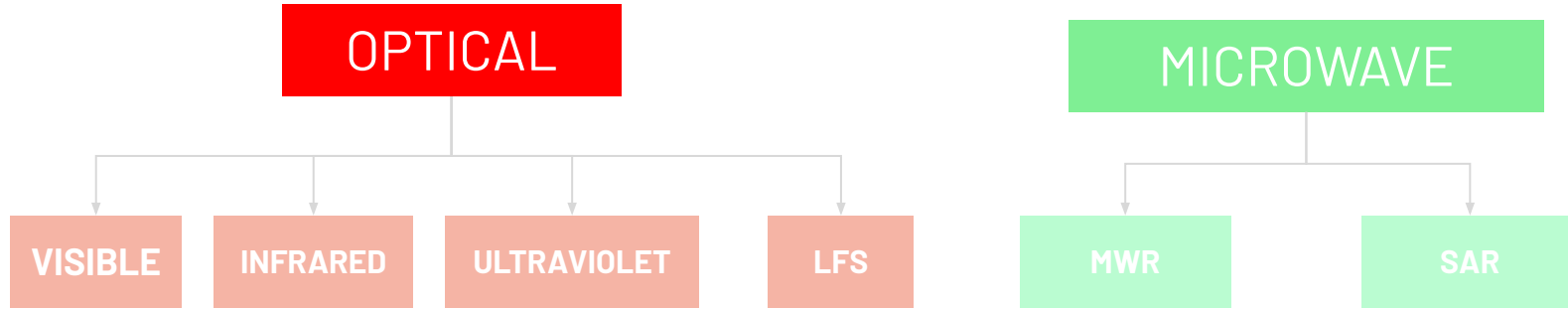


Accidental Discharges

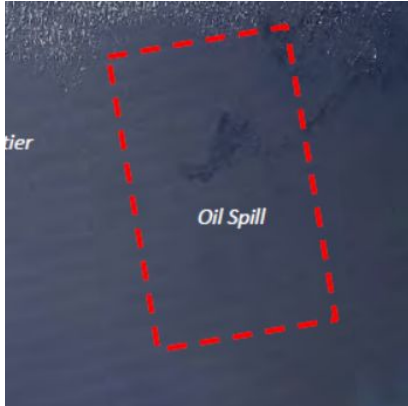
- Ship Accident
- Platform Accident

NON-LINE SHAPE

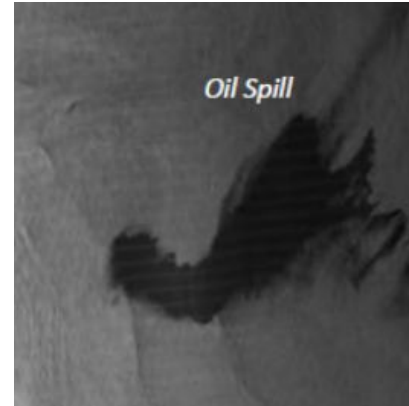
SATELLITES IMAGE TYPES OF OCEAN MONITORING



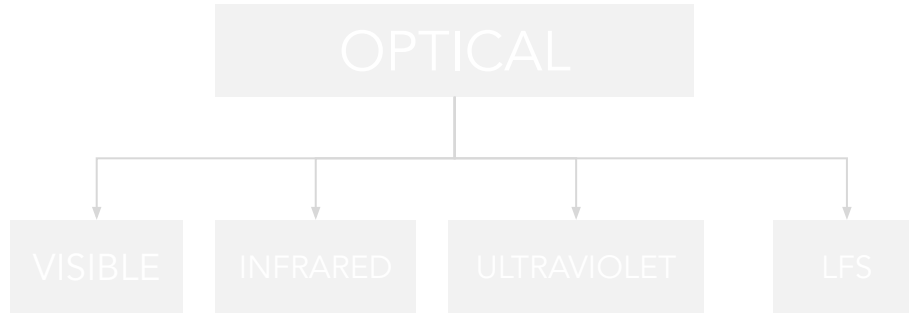
Example of **Optical Image** acquired by the NASA Satellite



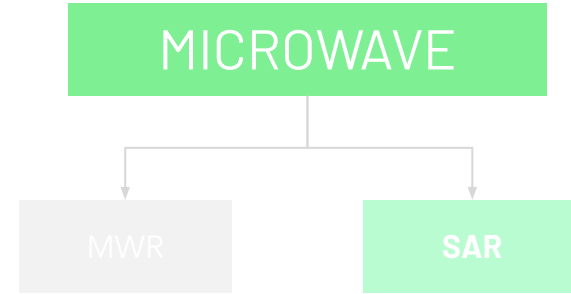
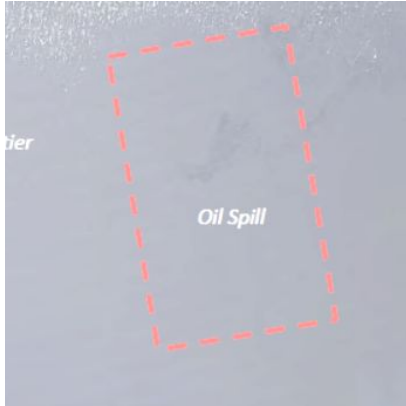
Example of **Synthetic Aperture Radar (SAR) Image** acquired by the Terra SAR-X.



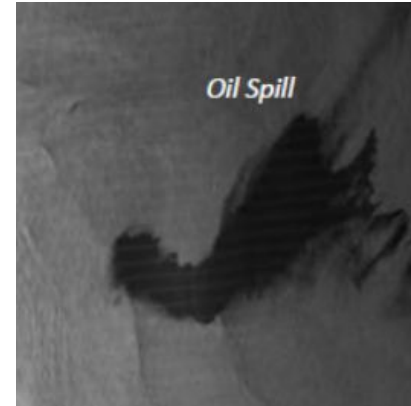
SATELLITES IMAGE TYPES OF OCEAN MONITORING



Example of **Optical Image** acquired by the NASA Satellite



Example of **Synthetic Aperture Radar (SAR) Image** acquired by the Terra SAR-X.



THE STRENGTH OF SAR IMAGES

“ Synthetic Aperture Radar (SAR) ”



ALL WEATHER CONDITION

SAR sensors are self-illuminating and can **penetrate clouds, fog, smog, darkness and smoke.**



24 HR OPERATION

SAR satellites can capture images from precisely the **same imaging geometry** every 24 hours.



HIGH SPATIAL RESOLUTION AND COVERAGE

SAR can provide a **wide range** of spatial resolution and coverage.

DATASET

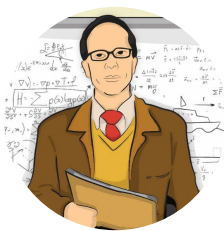


DATASOURCE

The trained data source
(SARImgV3) is stored in
Roboflow websites.

1,400

No. of TRAIN SET



DATA PROVIDER

The data is provided by
Mr. Matteo Attimonelli.
(Student of Oil Spill Detection
Professor)

200

No. of VAL SET

100

No. of TEST SET



TYPE OF DATA

The Oil spill Satellite Data is
the **Synthetic Aperture
Radar (SAR)** type.

416 x 416

Dimensions (pixels)

MODEL DEVELOPMENT JOURNEY

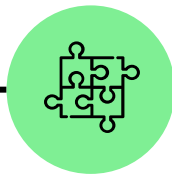
MODEL PRIMARY SELECTION

To select the most appropriate model.



OIL SPILL SHAPE AUGMENTATION

To do shape-based image augmentation.



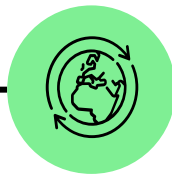
EPOCHS OPTIMIZATION

To find the most optimized epochs per performance.



OIL SPILL DIRECTION EQUALIZATION

To do direction-based image augmentation.

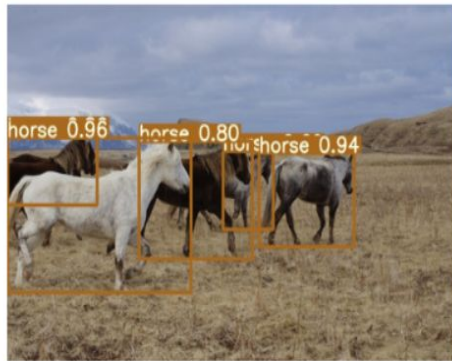


MODEL PRIMARY SELECTION

YOLO (YOU ONLY LOOK ONCE) **Version 7.**



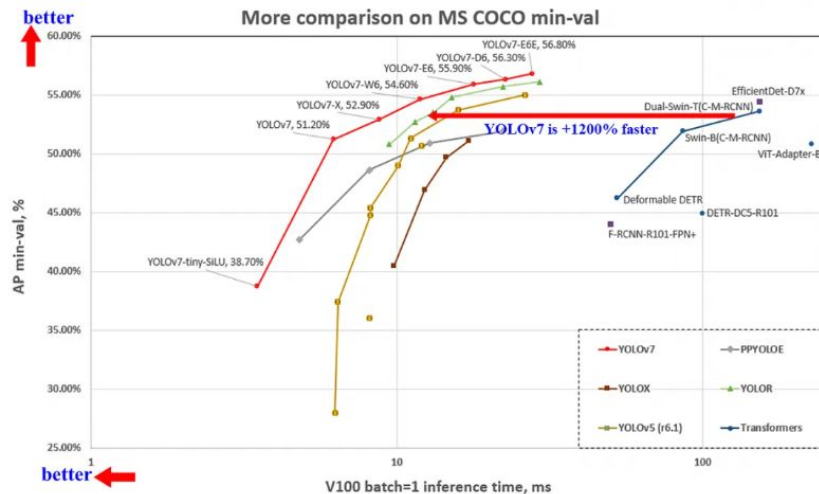
Original Image



Detected Image

YOLOv7 Architecture

- Extended Efficient Layer Aggregation Network (E-ELAN).
- Model Scaling for Concatenation based Models.



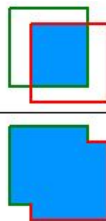
“YOLOv7 is the **fastest** and **most accurate** real-time object detection model for computer vision tasks. (2022)”

EPOCH OPTIMIZATION

mAP : Mean Average Precision

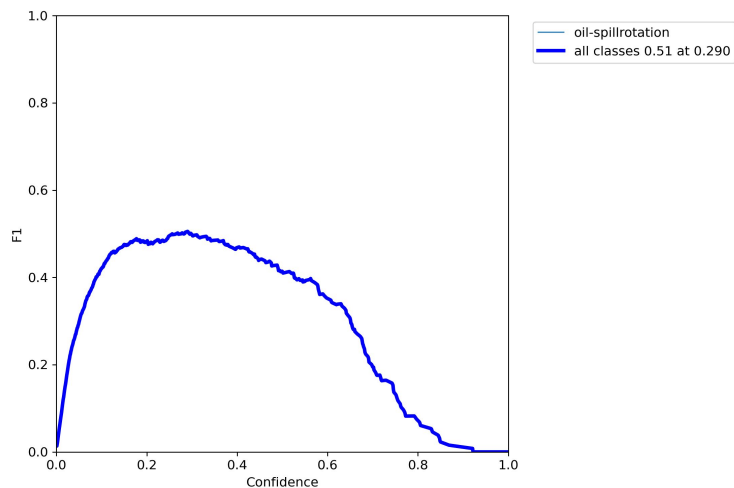
$$mAP = \frac{1}{|classes|} \sum_{c \in classes} \frac{|TP_c|}{|FP_c| + |TP_c|}$$

IoU : Intersect Over Union

$$IOU = \frac{\text{area of overlap}}{\text{area of union}} = \frac{\text{area of overlap}}{\text{area of union}}$$


Metrics	200 EPOCHS	300 EPOCHS
Precision	0.598	0.675
Recall	0.438	0.352
mAP @ IoU > 0.5	0.432	0.367

BASELINE RESULTS

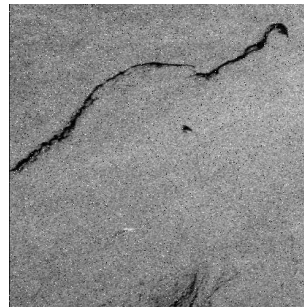


Metrics	200 EPOCHS
Precision	0.598
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ERROR ANALYSIS

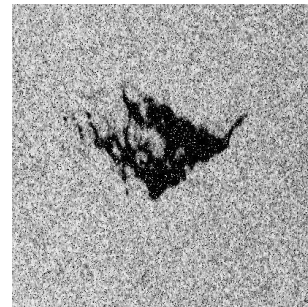
01

Imbalanced Shape Prediction



Line Shape Oil Spill

> 70 ~ 80 % of Training Set.



Non-line Shape Oil Spill

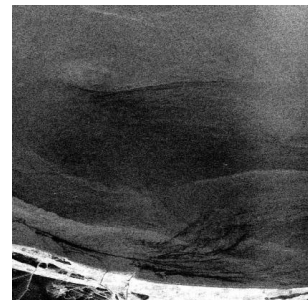
< 20 ~ 30 % of Training Set.

02

Spill Direction Bias Prediction



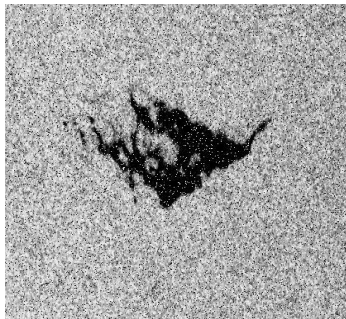
Vertical-like Direction



Horizontal-like Direction

IMAGE AUGMENTATION STRATEGY

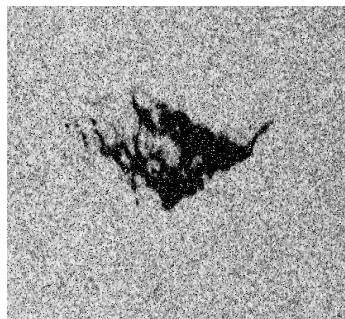
Add 300 pics.



01

NON-LINE SHAPE
AUGMENTATION
FOR 300 PICs.

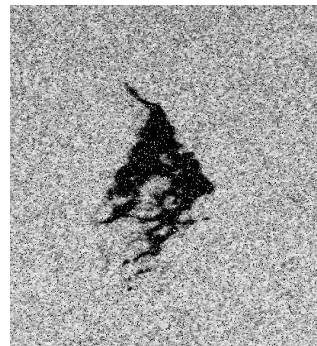
Add 900 pics.



02

NON-LINE SHAPE
AUGMENTATION
FOR 900 PICs.

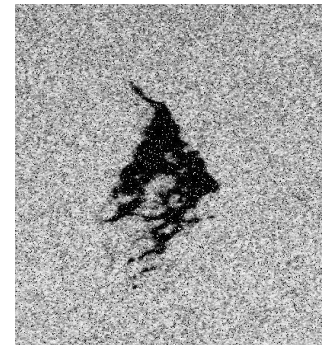
Add 300 pics.



03

Version 01 with
90 deg Rotation
AUGMENTATION.

Add New augmentation



04

Version 03 with
New Augmentation
Properties.

MODEL COMPARISON

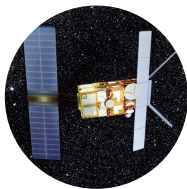
Metrics (Test Set)	300 Non-line augmentation	900 Non-line augmentation	300 Non-line augmentation with 90-rotation.	300 Non-line with additional augmentation.
Precision	0.675	0.659	0.599	0.599
Recall	0.352	0.429	0.457	0.449
Max F1 Score	0.51 @ 0.413 conf	0.52 @ 0.29 conf	0.52 @ 0.29 conf	0.51 @ conf
mAP @ IoU > 0.5	0.367	0.429	0.453	0.439

ERROR ANALYSIS ON REAL CASE



SENTINEL-1/2

EUROPEAN SPACE AGENCY



ERS-1/2

EUROPEAN SPACE AGENCY



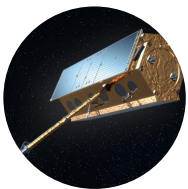
ENVISAT

EUROPEAN SPACE AGENCY



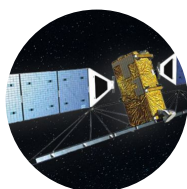
ALOS-2

JAXA



Terra SAR-X

German Aerospace Center



RADARSAT-1/2

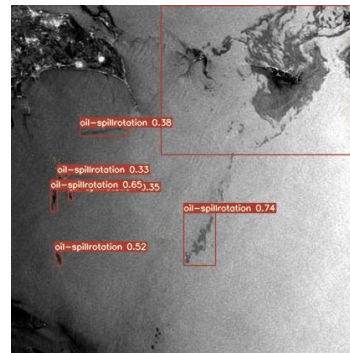
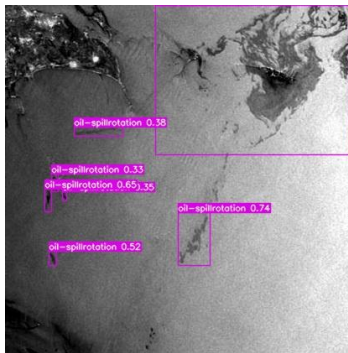
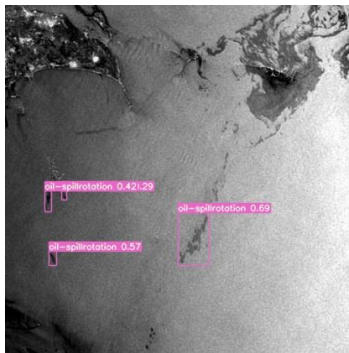
Canadian Space Agency

15 Major Oil Spill Events around the world

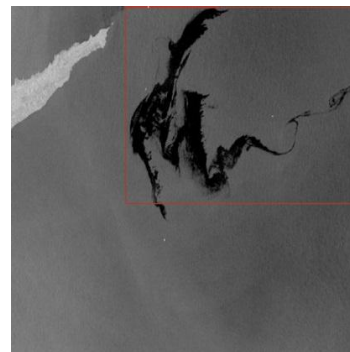
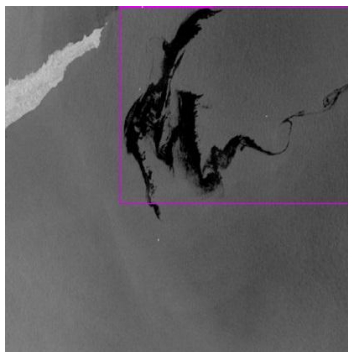
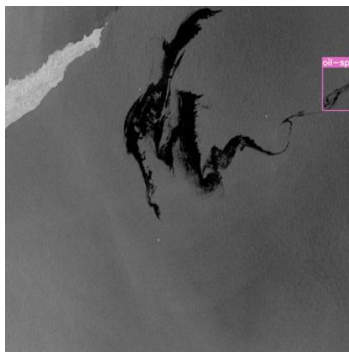


POST-AUGMENTATION RESULTS

Oil Spill Case from Caspian Sea, Azerbaijan



Oil Spill Case from Latakia, Syria to Cyprus



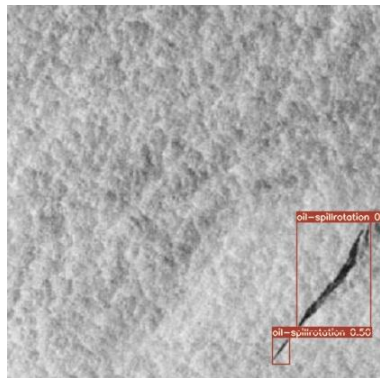
Before non-line shape
augmentation

After non-line shape
augmentation ver. 1

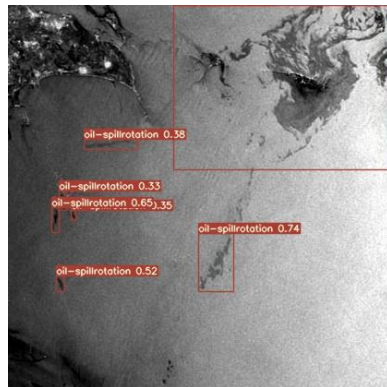
After non-line shape
augmentation ver. 3

BEST MODEL PREDICTION

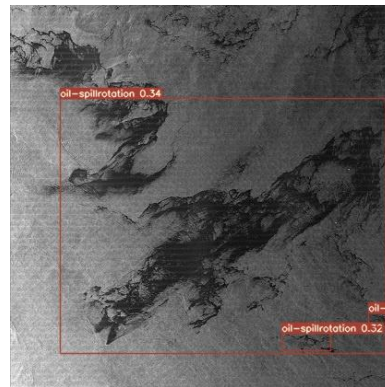
East China Sea, China/Japan



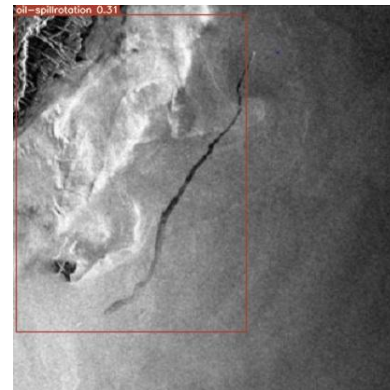
Caspian Sea, Azerbaijan



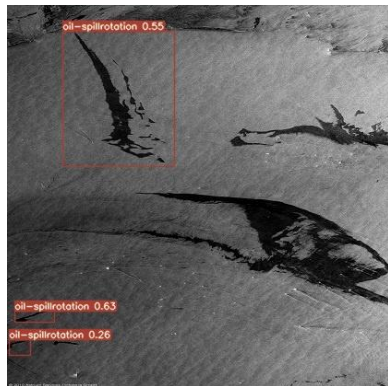
Montara, Australia



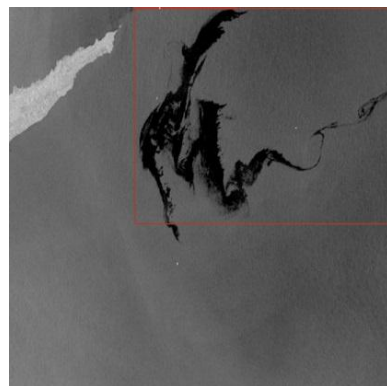
Kuroshio, Taiwan



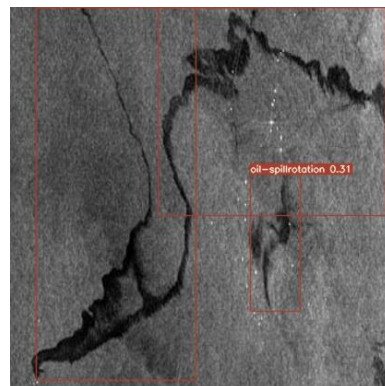
Gulf of Mexico, USA



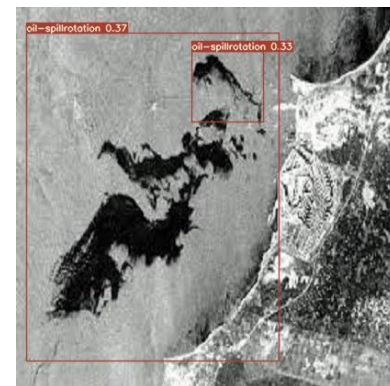
Latakia, Syria to Cyprus



Coast of Galicia, Spain

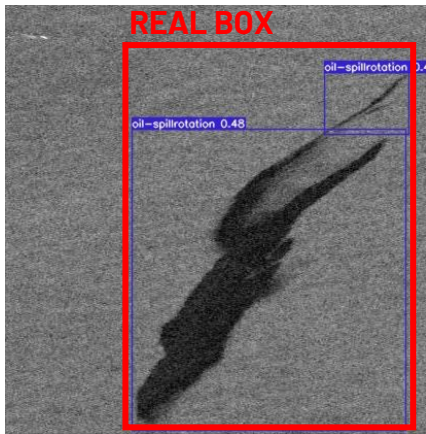
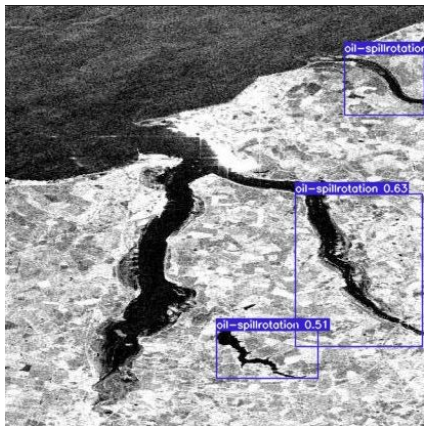


Al Khafji, Kuwait



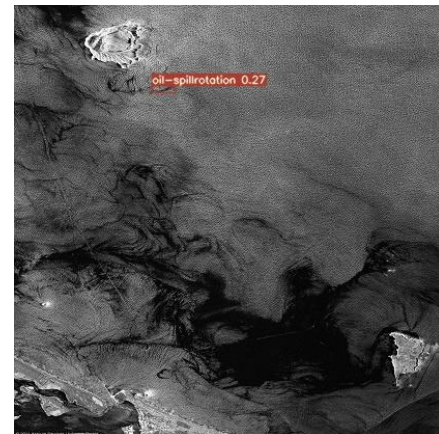
MODEL CONSTRAINTS

FALSE POSITIVE (FP)



- Model misclassified the label of real objects such the **river or water** resource near the ocean.
- Model framed the over/undersize of the real label which caused the low score in **mAP@IoU > 0.5**.

FALSE NEGATIVE (FN)



- Model could not detect high complexity of oil spill shape.

CONCLUSION

01

Final Model : YOLOV7

02

Augmented Method

- 90 Rotate : Clockwise, Counter-Clockwise and Upside Down.
- Crop : 0 - 37 % of Zoom
- Rotation : -5 to 5
- Brightness : -30 - 0 %
- Blur : Up to 2 Px
- Noise : Up to 10% of pixel

03

Hyperparameter

- **Defaults Hyperparameter** as per developer suggestions such as Lr0 = 0.01, Lrf = 0.1 etc.

04

Model Performance

Metrics (Test Set)	Baseline	Best Model
Precision	0.598	0.599
Recall	0.438	0.457
Max F1 Score	0.51 @ 0.29 conf	0.52 @ 0.29 conf
mAP @ IoU > 0.5	0.432	0.453

05

Model Constraints

- Real water resource misclassification
- Wrong size of box prediction
- Complexity of Shape misclassification

WAY FORWARD



Consult Domain Experts

To explore more features of **oil spill and ocean characteristics** with domain experts to boost spill detection.



Satellite Data

To gather a higher size of train data to increase the **model performance** and **reliability**.



Explore more data in Thailand

To seek out Thailand-based satellite data to **domestically utilize** oil spill detection in risk location i.e. Gulf of Thailand.



Alternative Options

To explore more advanced options such as **image segmentation** in order to reduce the limitations of object detection.



THANK YOU
for listening!