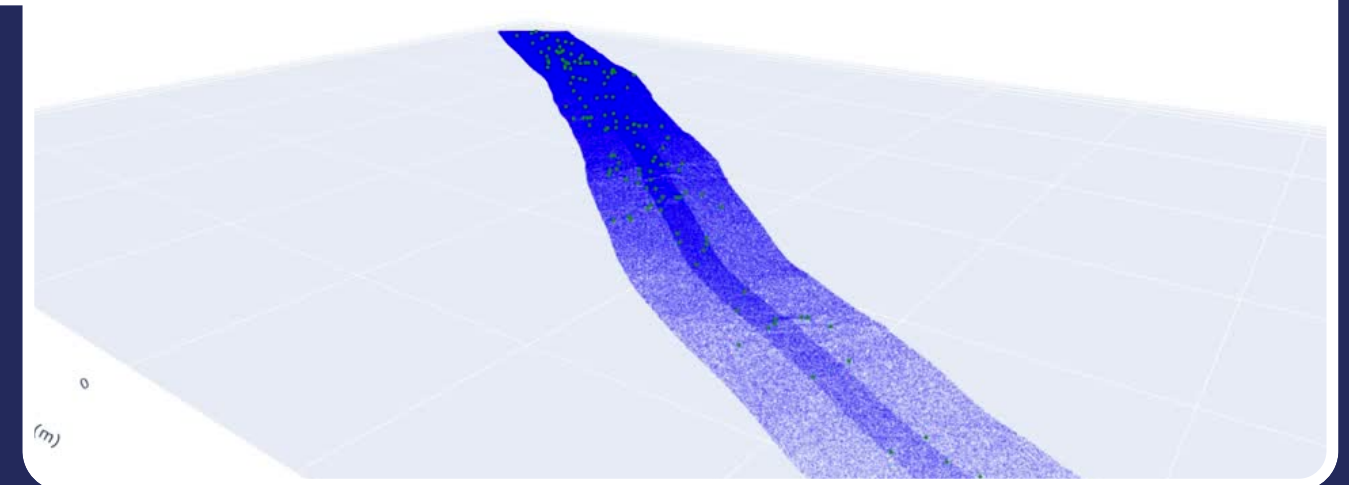


Everything within Reach

Future Subsea Robotics

Sustainable access to ocean space.
Haugesund, August 2026

George Jones



The Vision

2030 targets

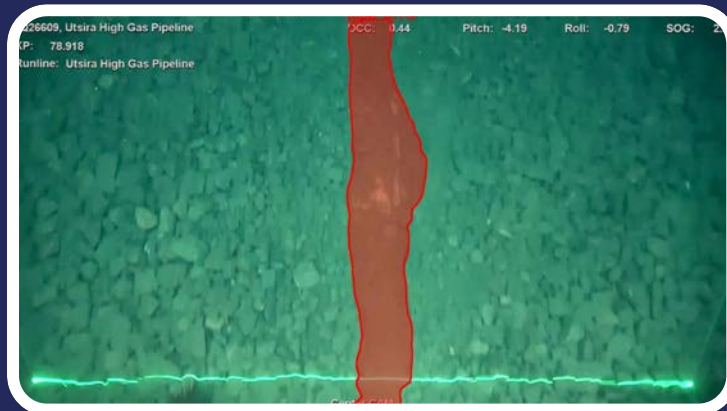
Advancing towards highly autonomous operations by 2030.

Enabled by Reach Remote & intelligent ROV.

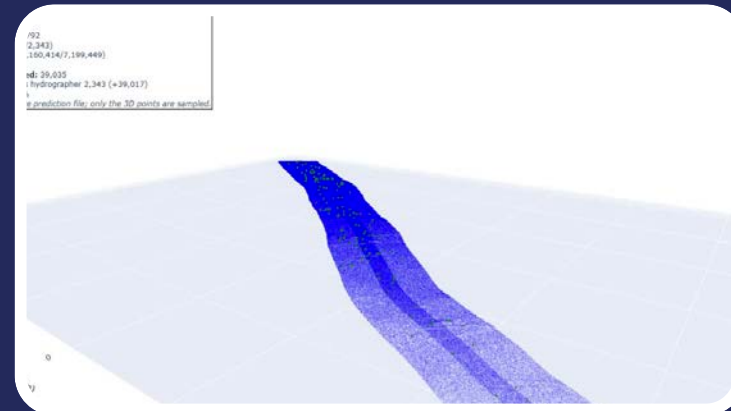
Contents

Four proof of concept projects

Asset Tracking



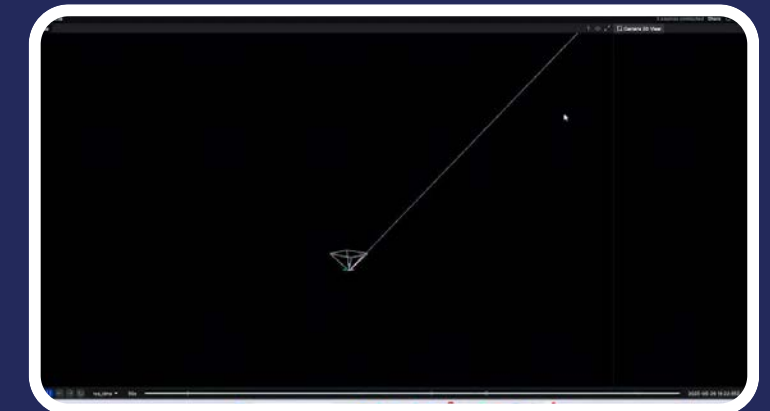
Automated Cleaning



AI Supervision



Intelligent ROV



Pipeline Tracking



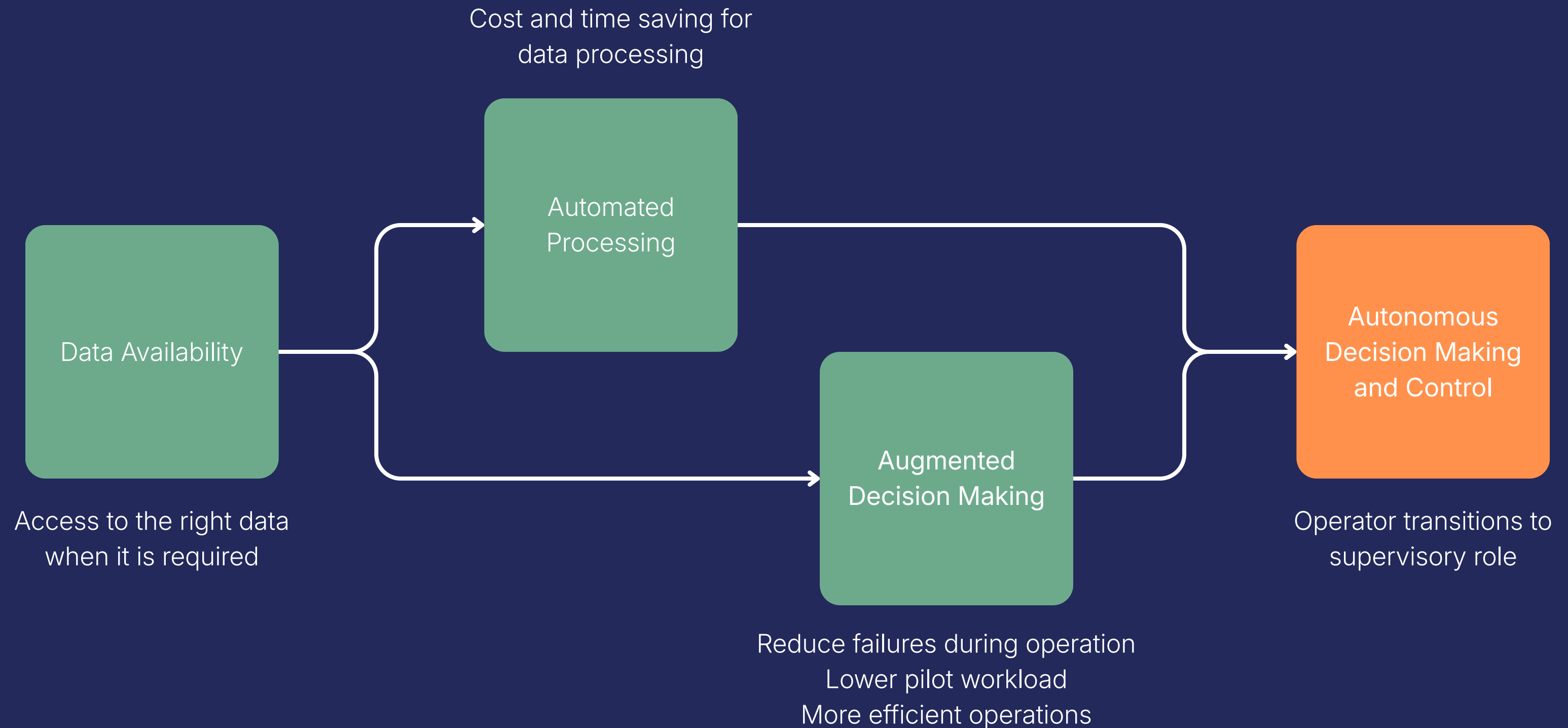
Aims & Objectives

- Track pipelines during surveys to:
 - Provide a safety failover for current autonomous operation.
 - A multi-modal approach to **increasing probability of maintaining track of the pipeline.**
- Augmenting the pilot view to make pipeline tracking easier, and more robust.



Roadmap to 2030 Autonomous Operations

From data to actions



Pipe Video

Center Camera



The Challenge

REACH
SUBSEA

PROJECT EVENTING

LAST EVENT ON PROJECT 600087 @ 04:22:50 12-01-2026 | TOTAL: 4

MAN-MADE OBJECT

DAMAGE

CORROSION

MARINE GROWTH

PIPELINE FEATURE

SEABED FEATURE

ANOMALY

REPLAY CONTROL | LAST 24H

VIDEO CHANNELS: ROV_CENTER_DVR

TIME PERIOD: 01/02/26 19:31:00

TIME BEFORE: 1 MIN

TIME AFTER: 1 MIN

MINUTES BACK: 1440 1320 1200 1080 960 840 720

ROV_CENTER_DVR

2026-02-01 19:33:01 RR01/VIDEO/0-CC-HC-RR01-VM05-Kytdesign/ROV_CENTER_DVR
01/02/26
19:33:00
KP: 4.398
DCC: -0.49
SN2016_04_C1
E 302214.60
N 6624730.30

SN2025_14_Bamlafjorden, SN2016_04_Cable_01

Gyro 182.21
Pitch 1.37
Roll -0.12
Depth 23.24
Altitude 2.54
Total Depth 25.78
Laser separation 0.1m

Cable was not where it was supposed to be!

19:31:20 19:31:30 19:31:40 19:31:50 19:32:00 19:32:10 19:32:20 19:32:30 19:32:40

EVENT LOG

FIELD LOG | REACH REMOTE 01 | PROJECT 600087

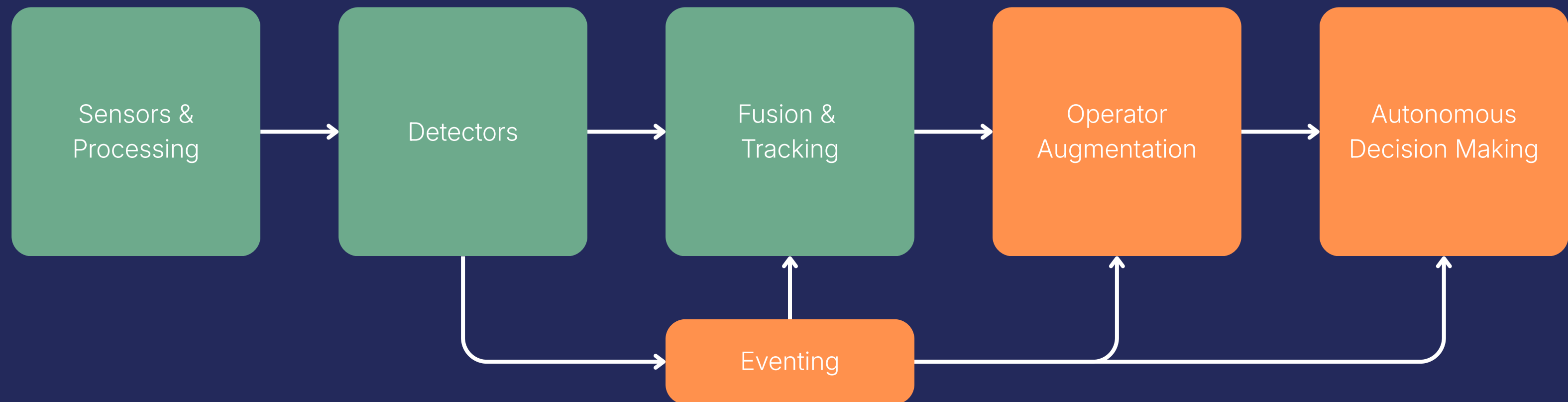
SELECT PERIOD: 27/01/26

Filter table...

INDEX	UTC DATE TIME	LOCAL DATE TIME	TASKPLAN	RUNLINE	KP	KP REF	EVENT
-------	---------------	-----------------	----------	---------	----	--------	-------

Architecture

■ Implemented @ TRL3-4
■ In progress / TBD



Laser Profiler



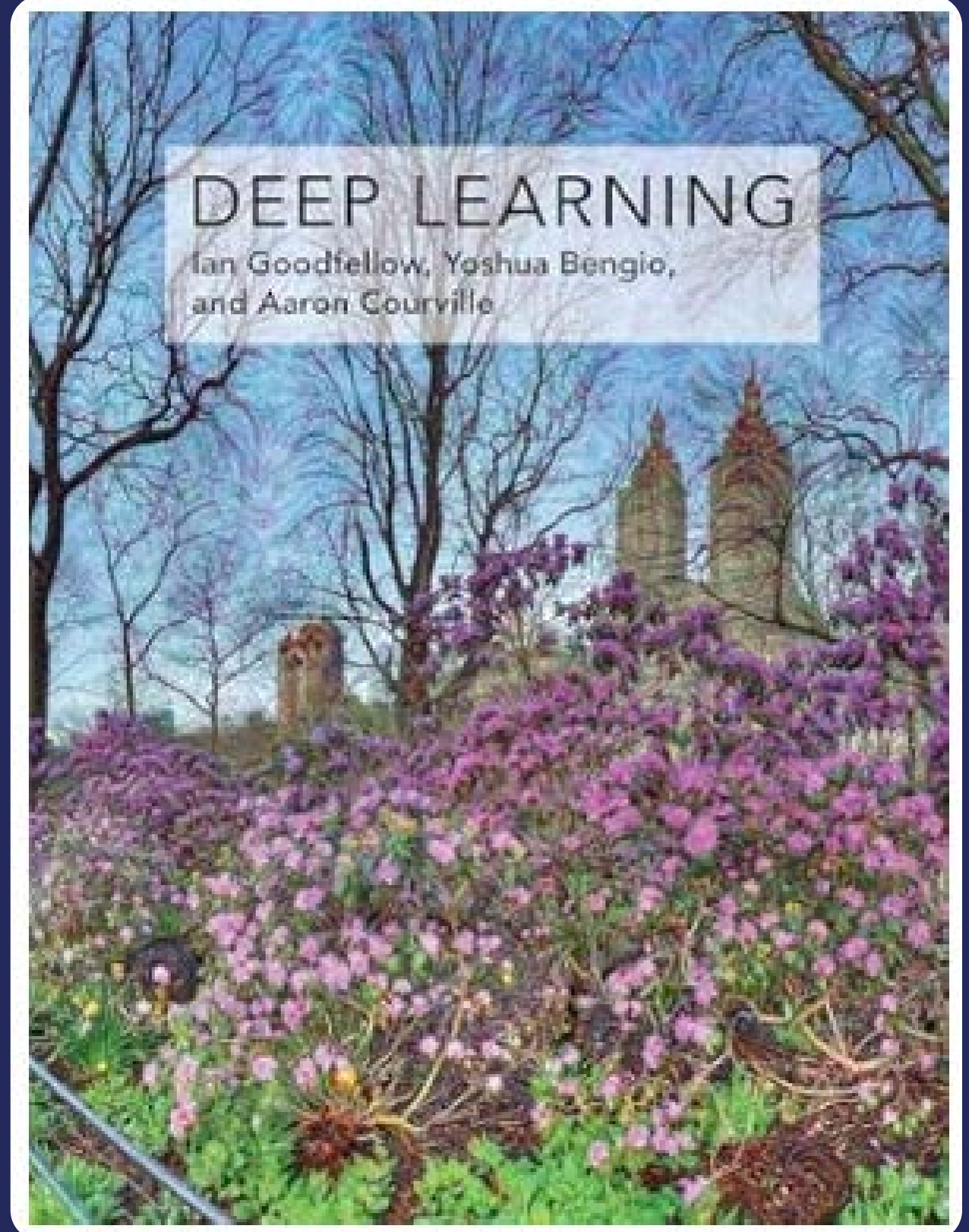
Laser Profiler

But there are two main challenges:

1. Requires a reference laser.
2. Rocks (or other non-planar objects) can cause detections.



But what about...



But what about Deep Learning

- First we utilised open source datasets

- MarinaPipe

- SubPipeMini

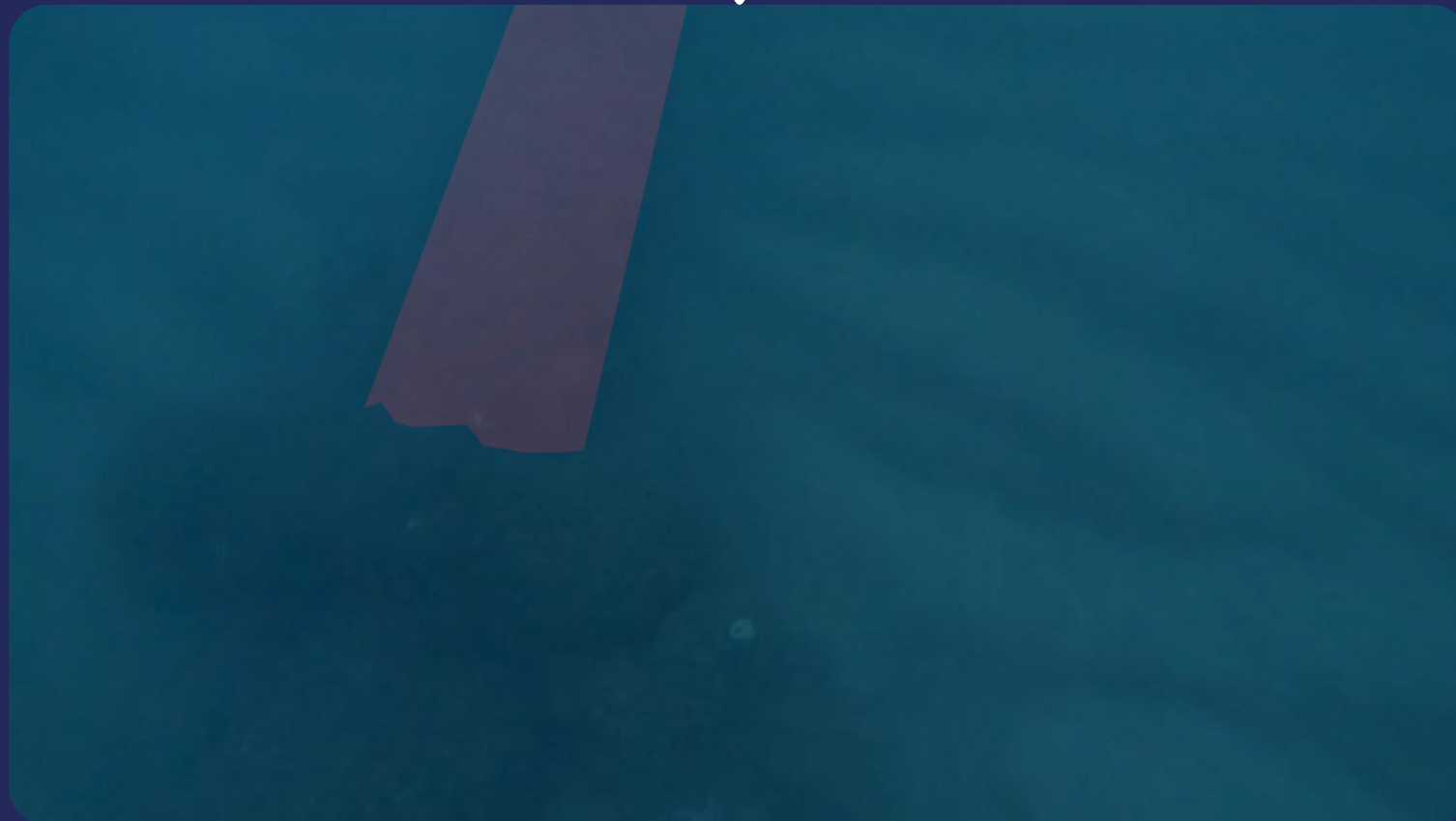


But what about Deep Learning

- First we utilised open source datasets

- MarinaPipe

- SubPipeMini



But what about Deep Learning

- First we utilised open source datasets:
 - MarinaPipe
 - SubPipeMini
- But it had issues:
 - A lot of the images had no pipe in them.
 - Licensing considerations.
 - Camera quality.
 - Not really representative of what we wanted to use it on.

So we labelled some of our own...



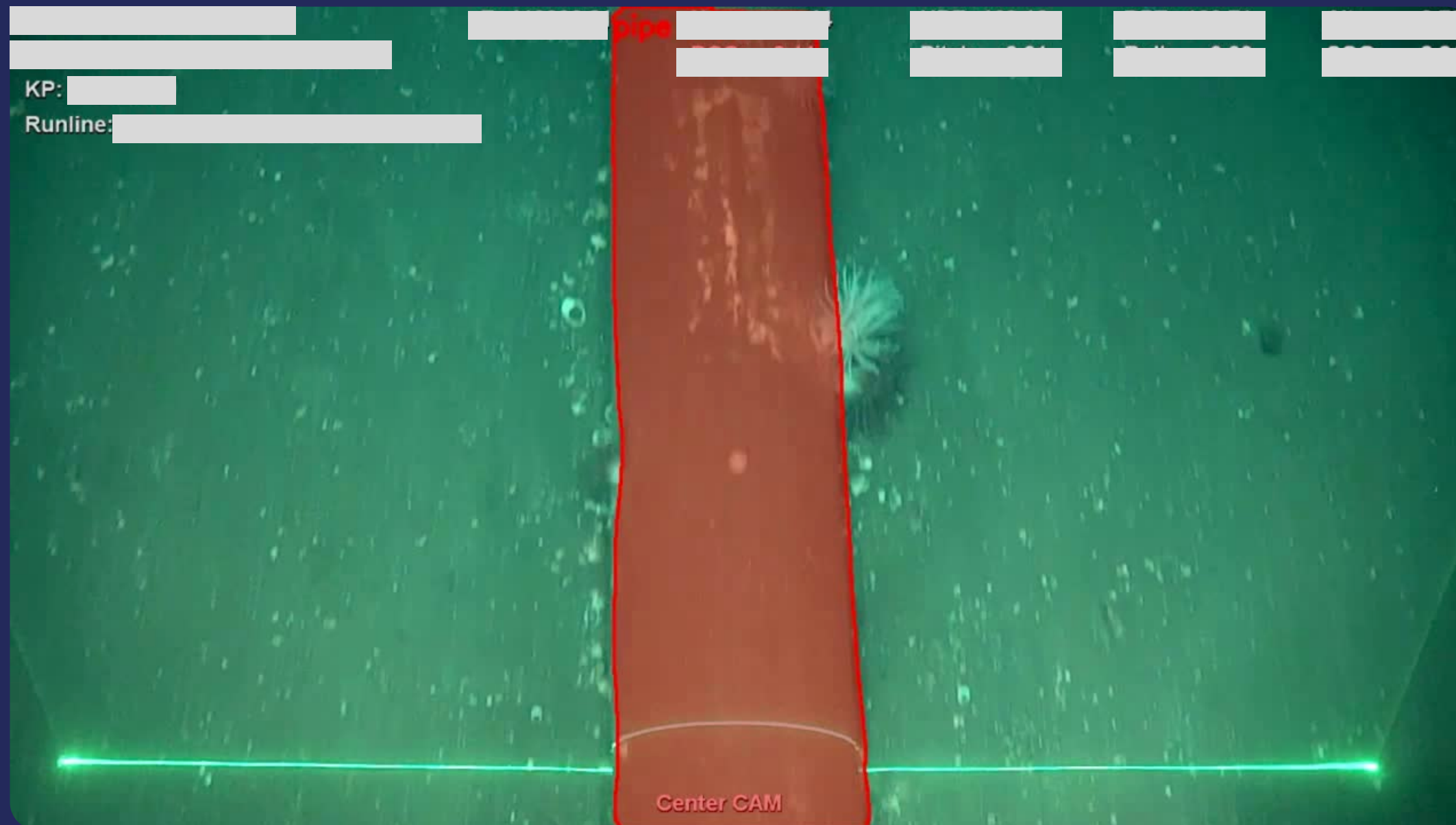


REACH SUBSEA

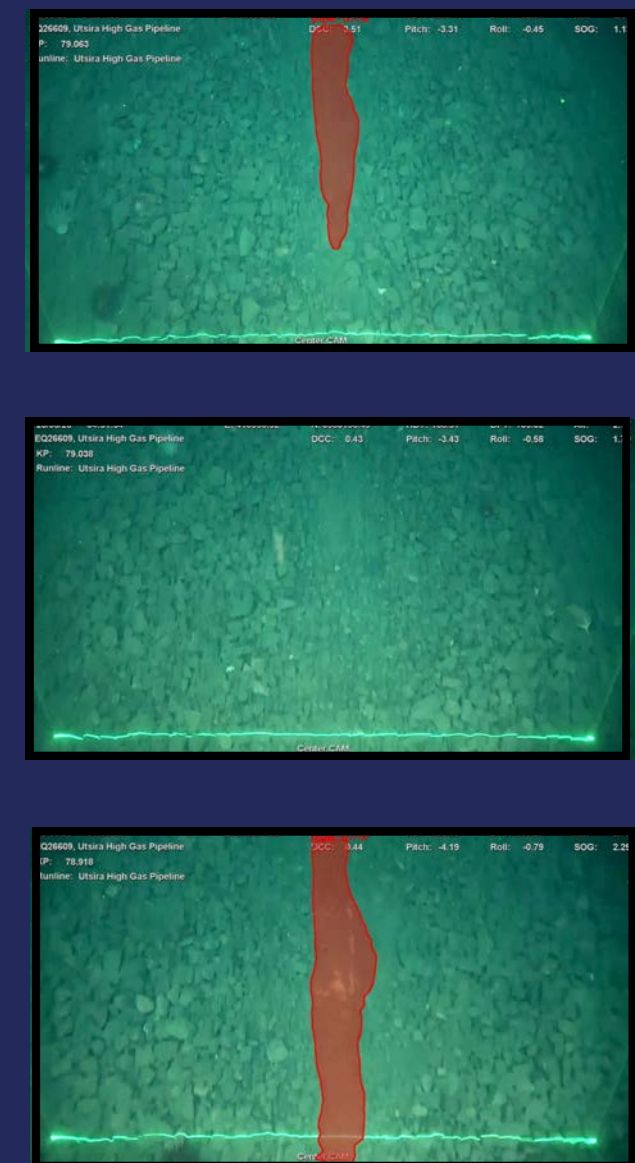


Computer Vision - Pipe Segmentation

Pipe in full view

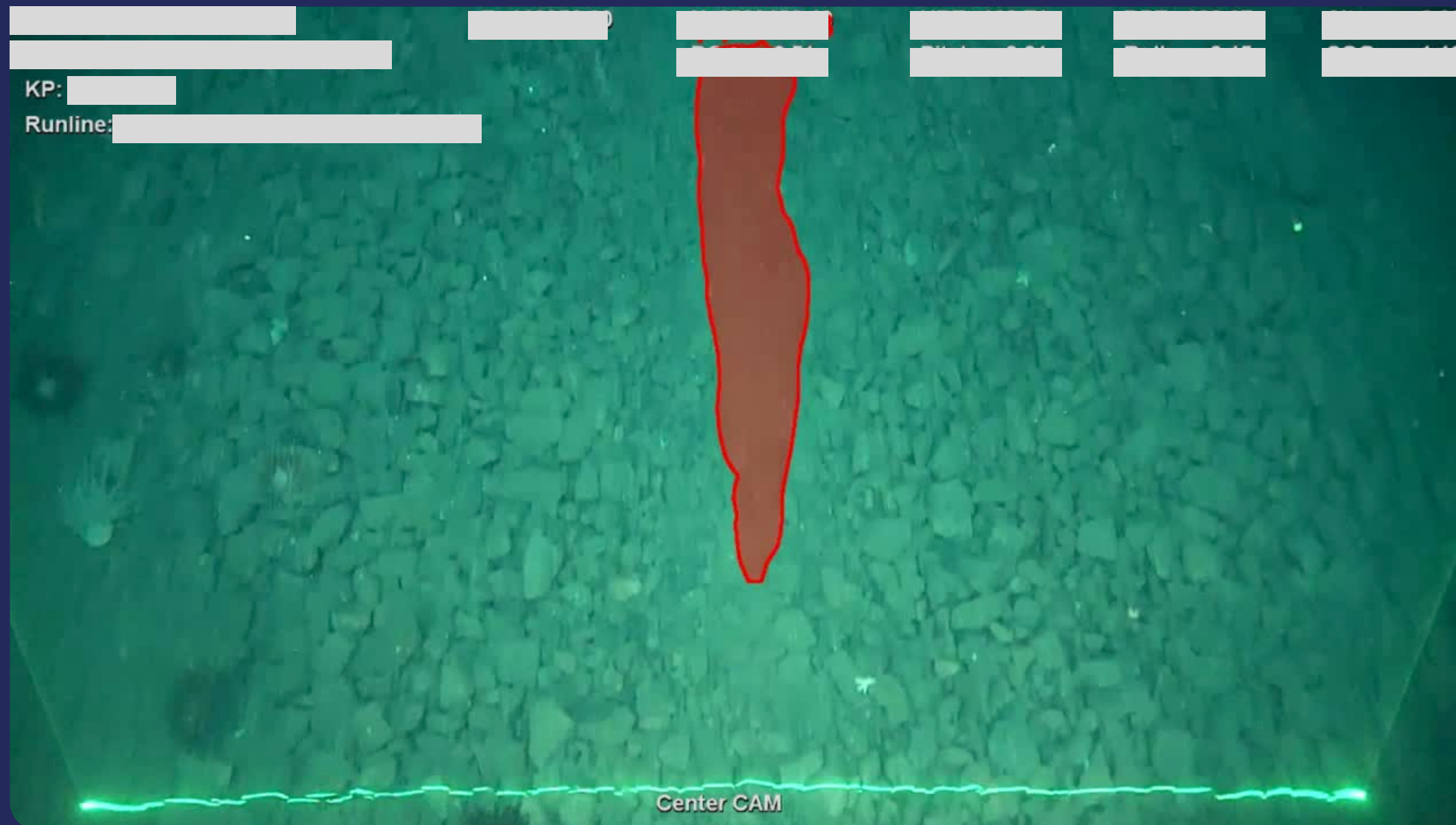


Click to enlarge

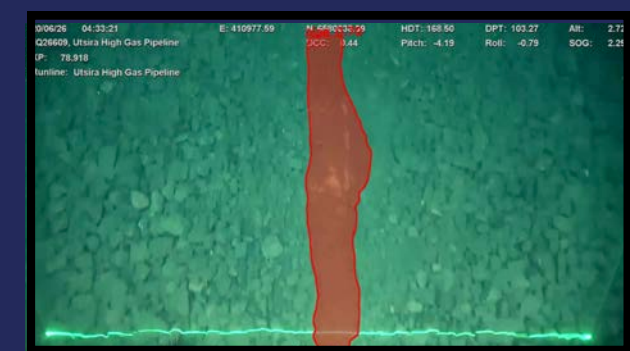
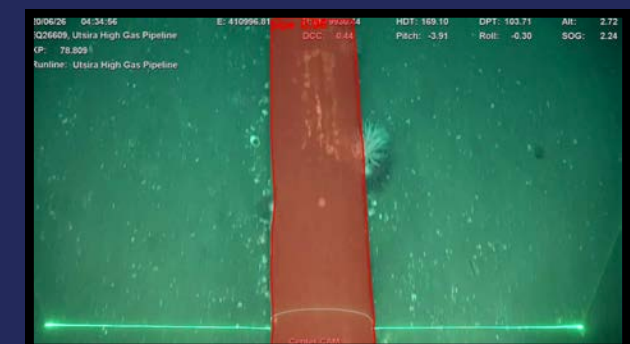


Computer Vision - Pipe Segmentation

Pipe entering gravel dump



Click to enlarge

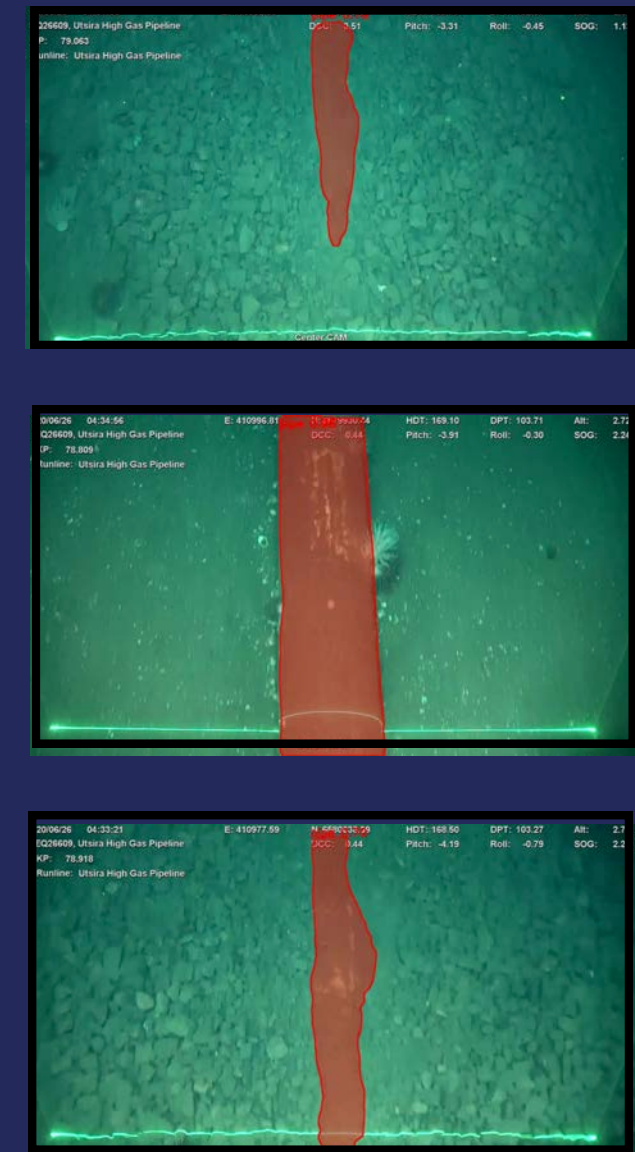


Computer Vision - Pipe Segmentation

Small pipe segments showing through gravel dump



Click to enlarge



Computer Vision - Re-ran on cable example

CV applied blindly to video of losing track of cable



Target Hardware Deployment with Tracker in Realtime

Jetson Thor



Summary

How it performed

- Tracks pipe well.
- Identifies and maintains track of pipe in difficult situations.
- Can run in realtime on edge hardware.

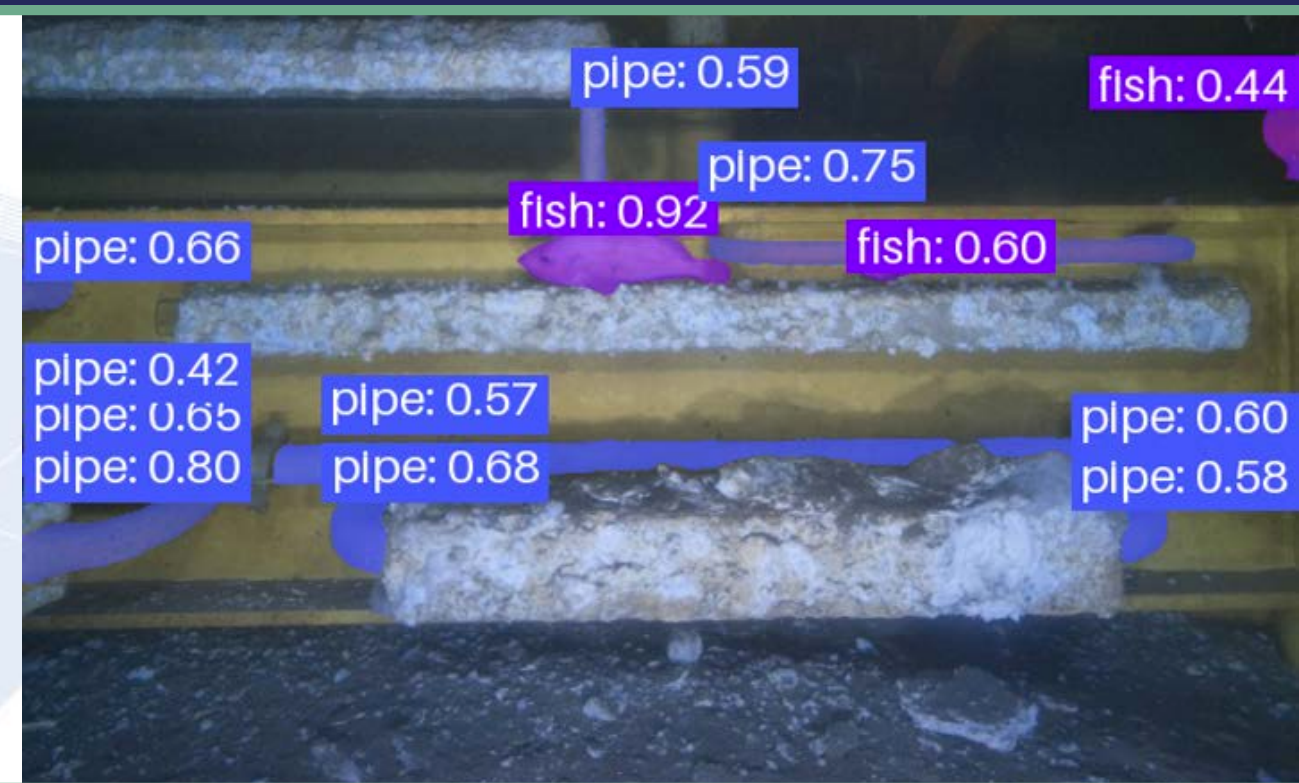
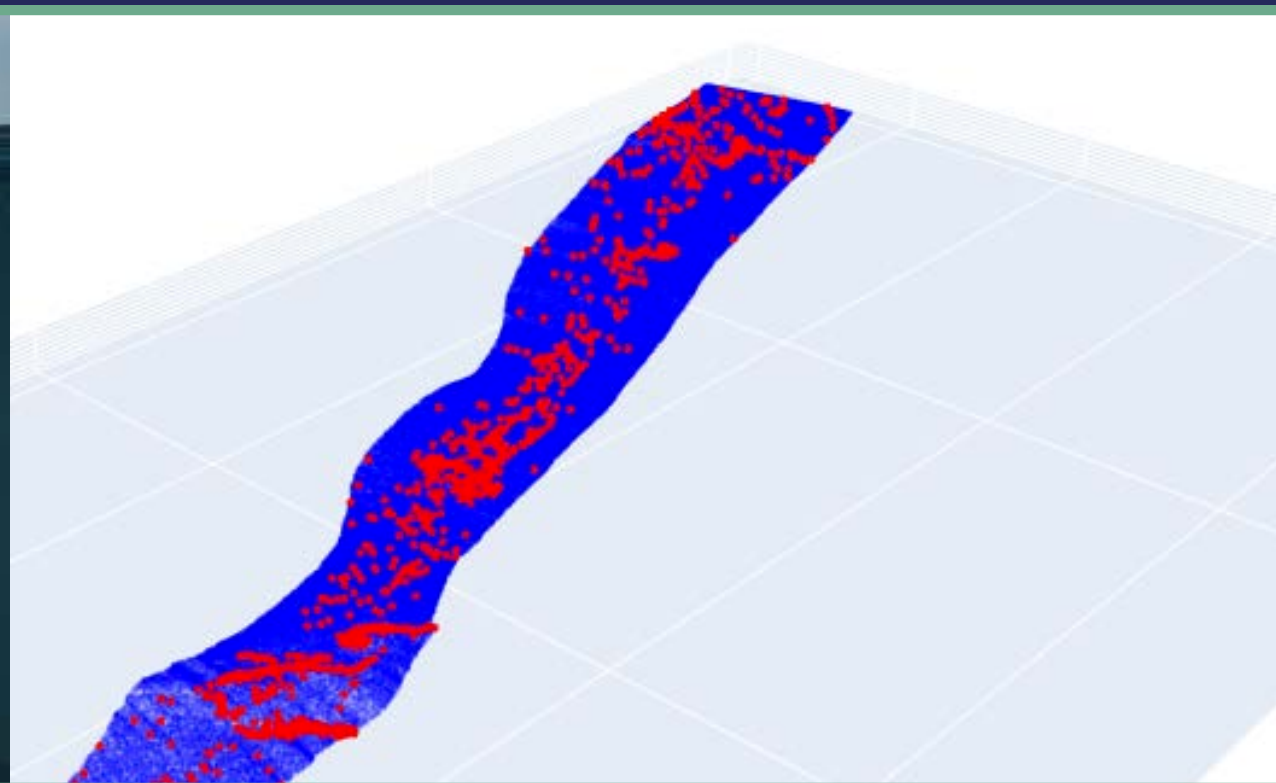
What next?

- Work with ROV pilots to understand how to best present this information to them.
- Sonar detector.
- Sensor fusion (ready for multi-modal)
- **Automatic eventing...**

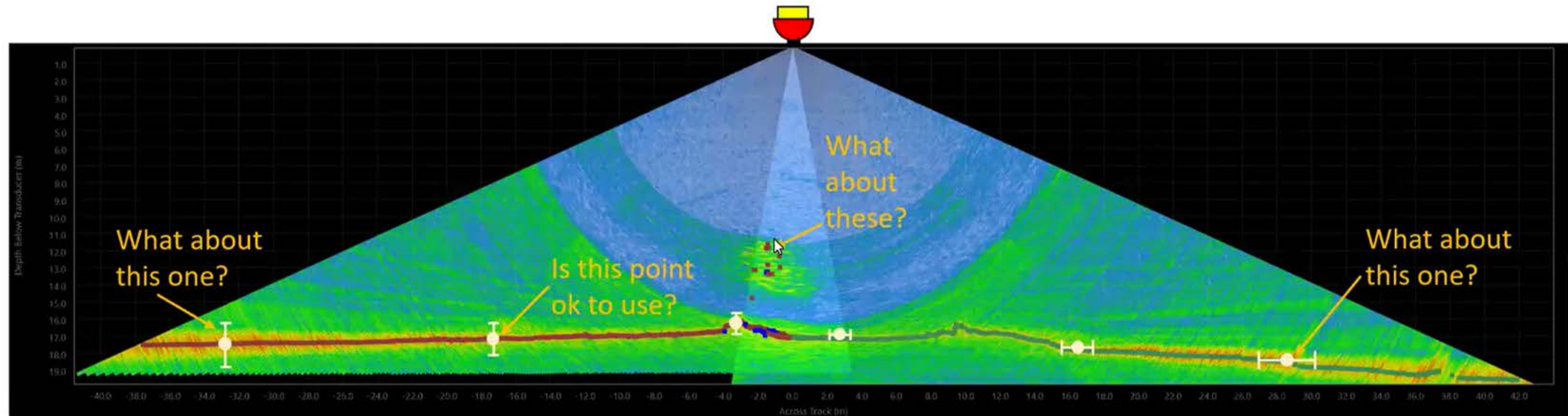
Automatic Eventing / Feature Extraction



Machine learning for MBES cleaning

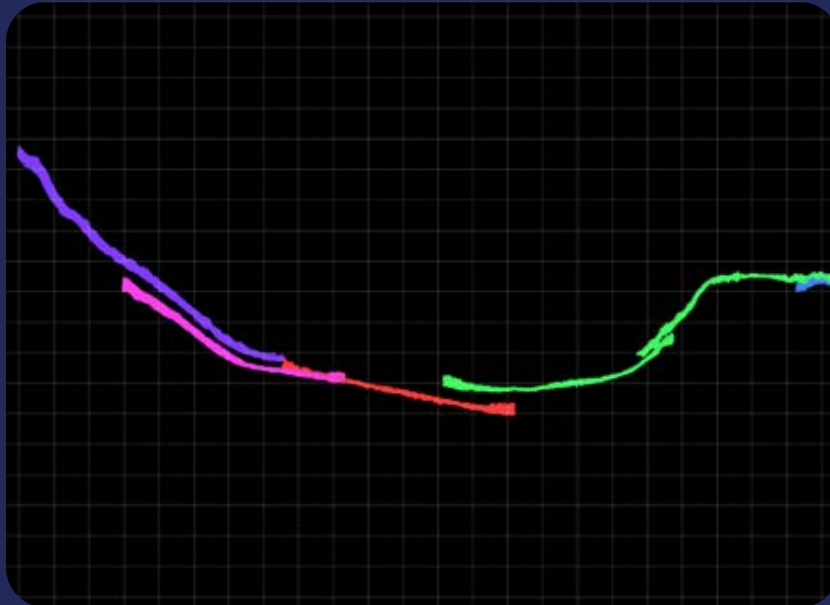


What is going on?

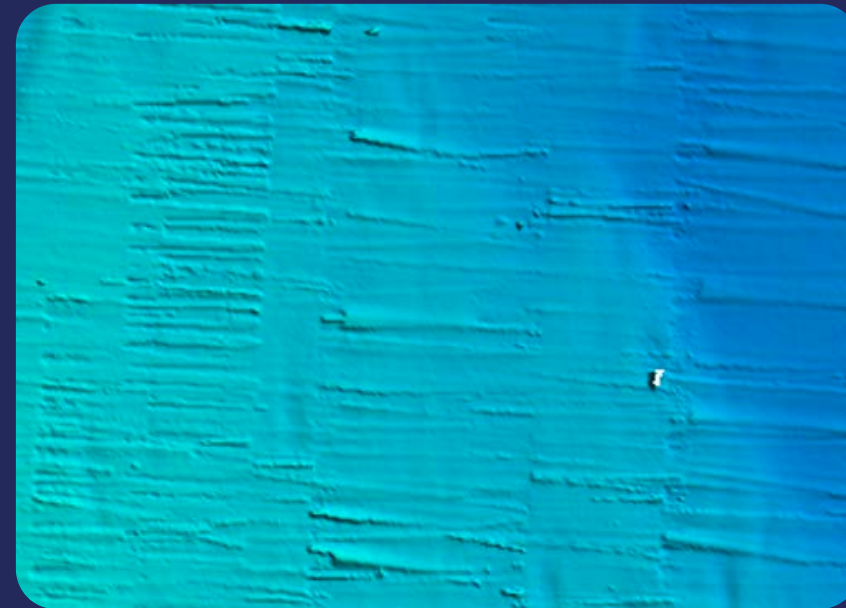


Sources of uncertainty

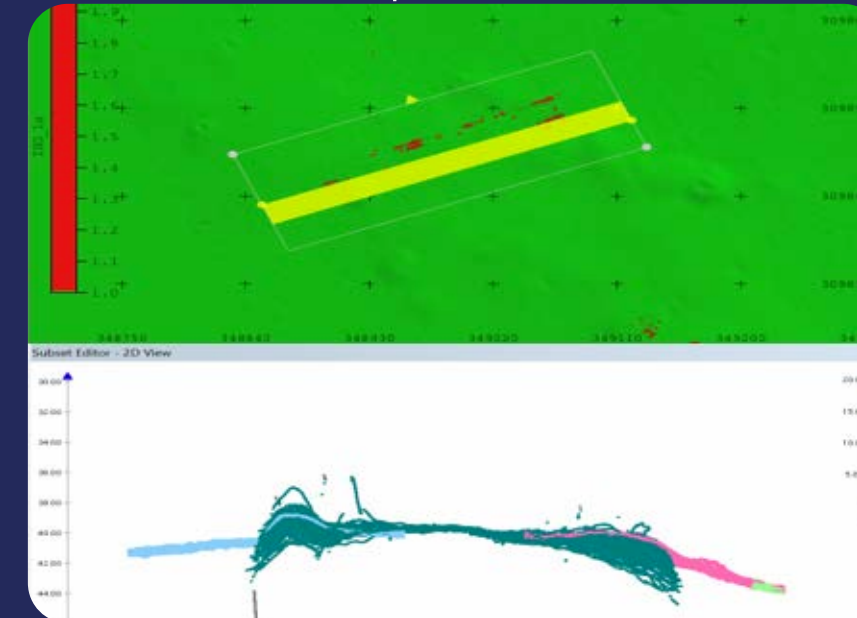
Tide misalignment



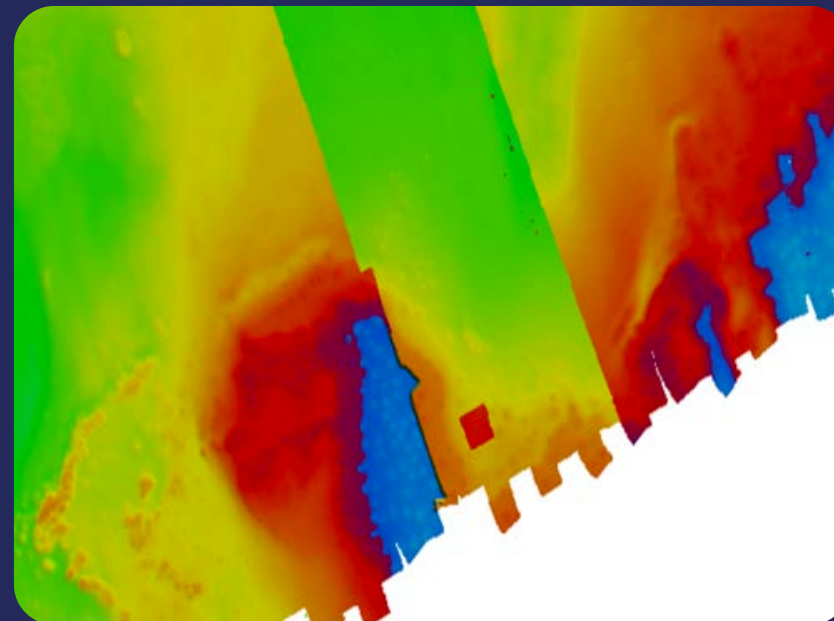
Bad weather



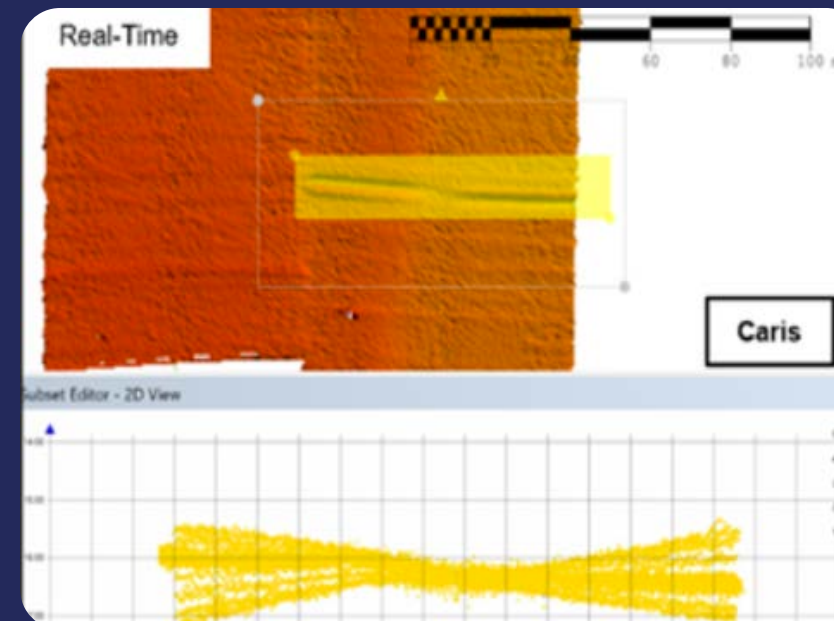
Swath open too much



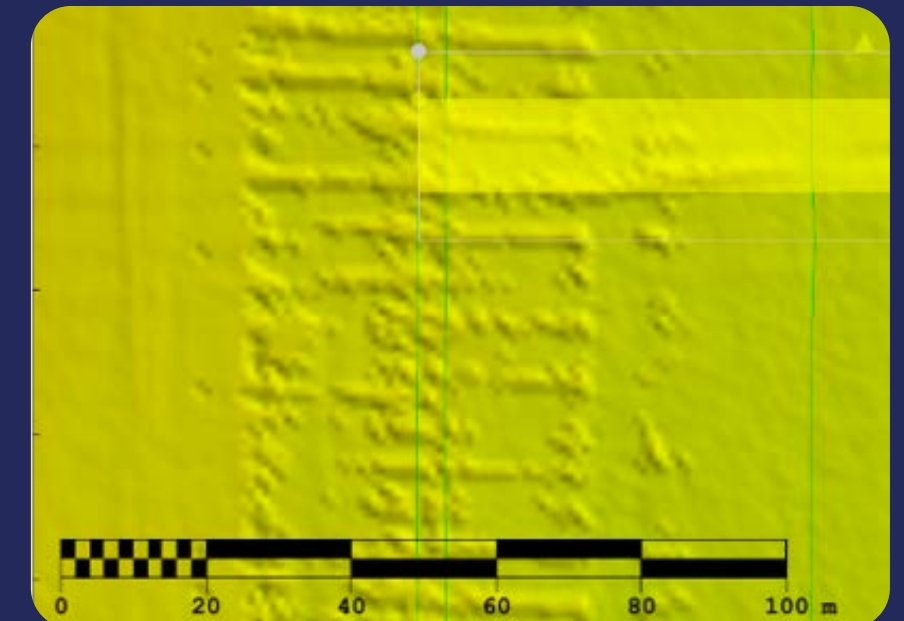
GPS Loss



Motion Artifact

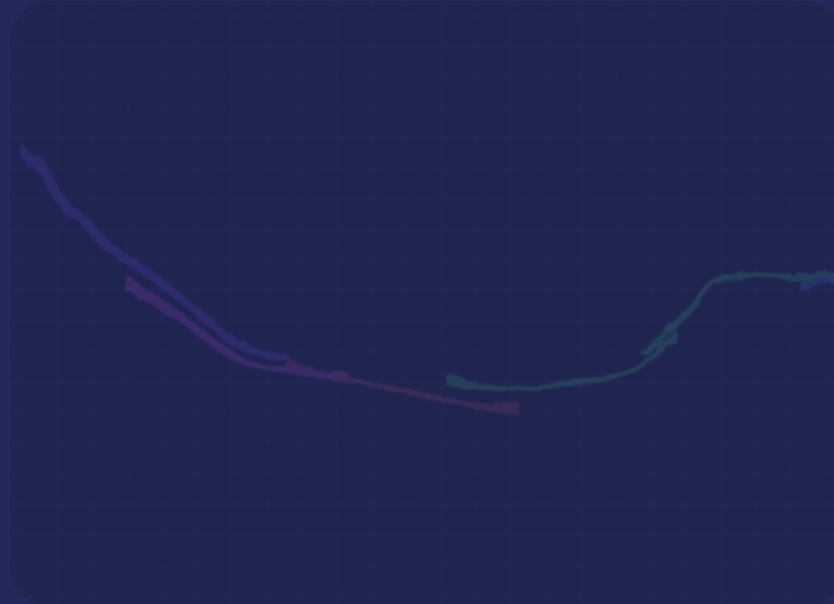


Poor GPS

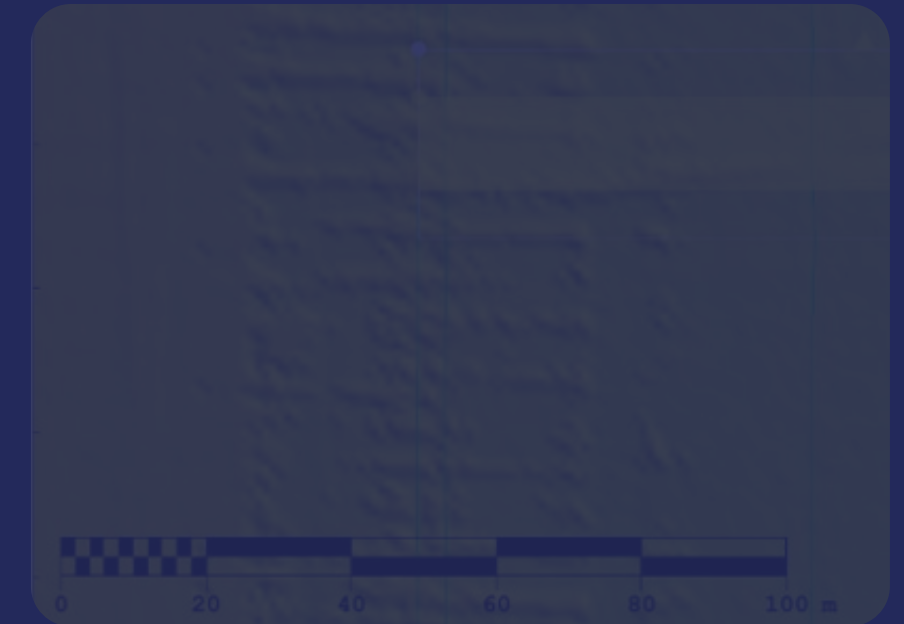


Sources of uncertainty

Tide misalignment



Poor GPS



The Challenge

1:4

Collection : Processing

For every 15 minutes spent collecting data, 1 hour must be expended processing.

~€15k

per day currently spent

24 + 8 data processors across reach operating at ~€60p/h.

$32 * 8 \text{ hrs/day} * €60 = \$15,360 / \text{day}.$

1200hrs

backlog of data

Significant backlog of data waiting to be processed.

1200 hrs processing time.

CLEANBYMACHINE

The (Proposed) Solution

What?

- A software tool to clean multibeam bathymetry using machine learning techniques, specifically gradient boosted trees (XGBoost).
- Uses Reach data assets (archived data) for training
- Training data is all hand-cleaned, formally approved multibeam recordings.
- One model per sonar type.
- During model training, feature combinations which indicate 'noise' are learned.

Why?

- Reduce manual data processing step by **X%**
- Increased consistency, repeatability and predictability compared to human operators.
- Time (and therefore) capital savings
 - at 60% reduction (assuming numbers from previous slide) - \$15,000 per day saving.
 - at 80% reduction - \$20,800 per day saving.

Training, testing and unseen survey data

Training



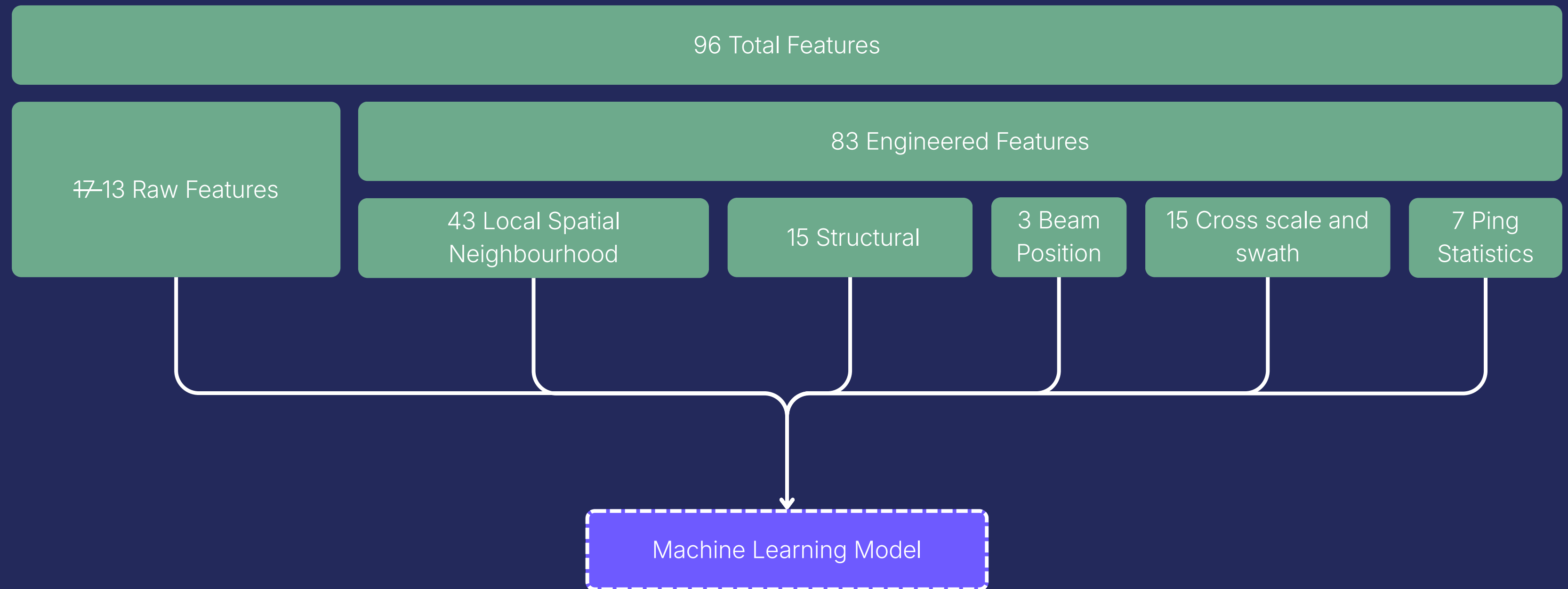
Across several surveys &
several vessels.

Unseen



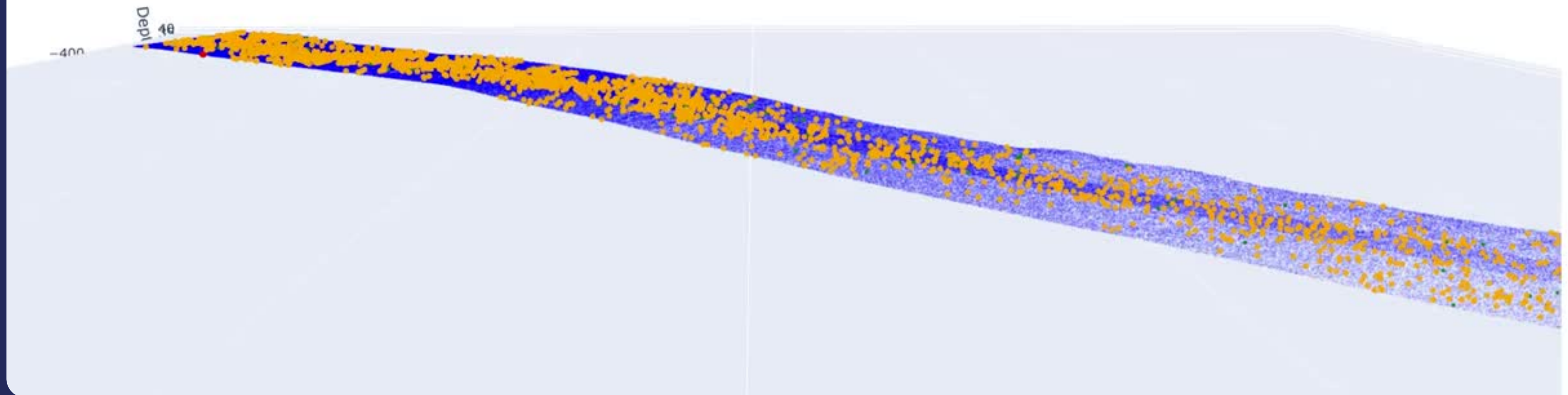
Ringfenced from training to
assess blind transferrability.

Features Summary - 96 total



Results (run 26)

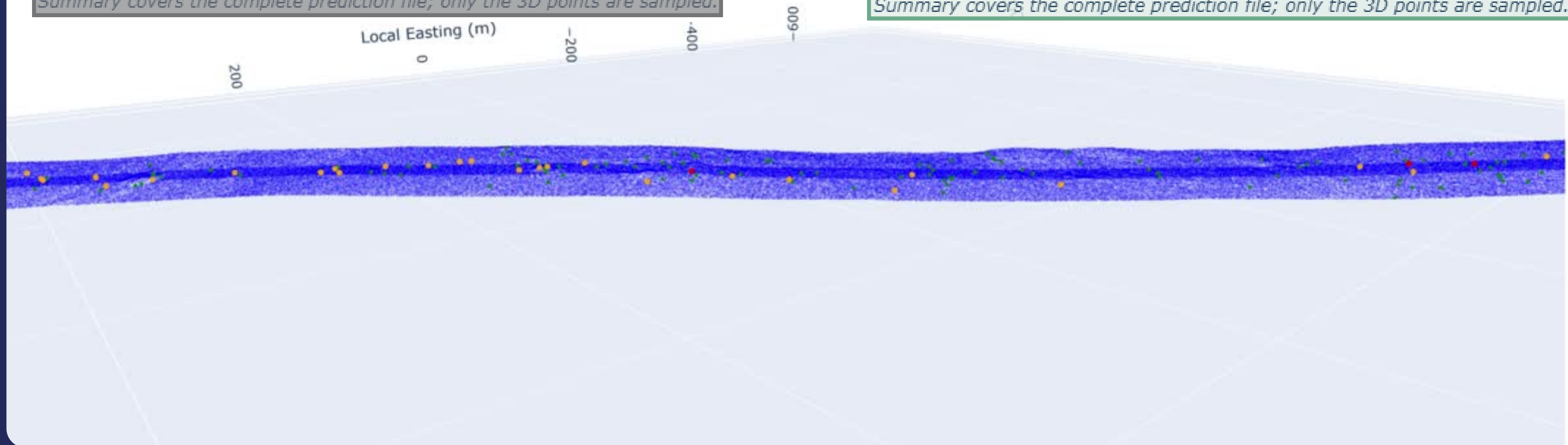
Threshold: 0.50
Full-file rows: 7,201,792
Valid labelled rows: 7,201,792
Noise found: 99.1% (2,322/2,343)
Good data kept: 99.41% (7,156,715/7,199,449)
Noise missed: 21
Good data wrongly removed: 42,734
Cleaner rejected: 45,056 vs hydrographer 2,343 (+42,713)
Overall agreement: 99.41%
Summary covers the complete prediction file; only the 3D points are sampled.



Results (run 26)

Threshold: 0.50
Full-file rows: 7,201,792
Valid labelled rows: 7,201,792
Noise found: 99.1% (2,322/2,343)
Good data kept: 99.41% (7,156,715/7,199,449)
Noise missed: 21
Good data wrongly removed: 42,734
Cleaner rejected: 45,056 vs hydrographer 2,343 (+42,713)
Overall agreement: 99.41%
Summary covers the complete prediction file; only the 3D points are sampled.

Threshold: 0.01
Full-file rows: 7,201,792
Valid labelled rows: 7,201,792
Noise found: 94.8% (2,221/2,343)
Good data kept: 99.99% (7,198,796/7,199,449)
Noise missed: 122
Good data wrongly removed: 653
Cleaner rejected: 2,874 vs hydrographer 2,343 (+531)
Overall agreement: 99.99%
Summary covers the complete prediction file; only the 3D points are sampled.



Results - mapped back to the challenge

1 : 1.2

Collection : Processing

With a 70% reduction in workload, the ratio of collection to processing is much closer to 1:1.

~€11k

per day saved

70% reduction in time spent processing results in ~€11k savings per day.

360hrs

backlog of data

Should take 360 hours to process, as opposed to 1200 hrs.

Machine learning for PPE Detection

- Trained on open source dataset
- Capable of detecting
 - Person
 - Ear
 - Ear-muffs
 - Face
 - Face-guard
 - Face-mask
 - Foot
 - Tool
 - Glasses
 - Gloves
 - Helmet
 - Hands
 - Head
 - Medical-suit
 - Shoes
 - Safety-suit
 - Safety-vest

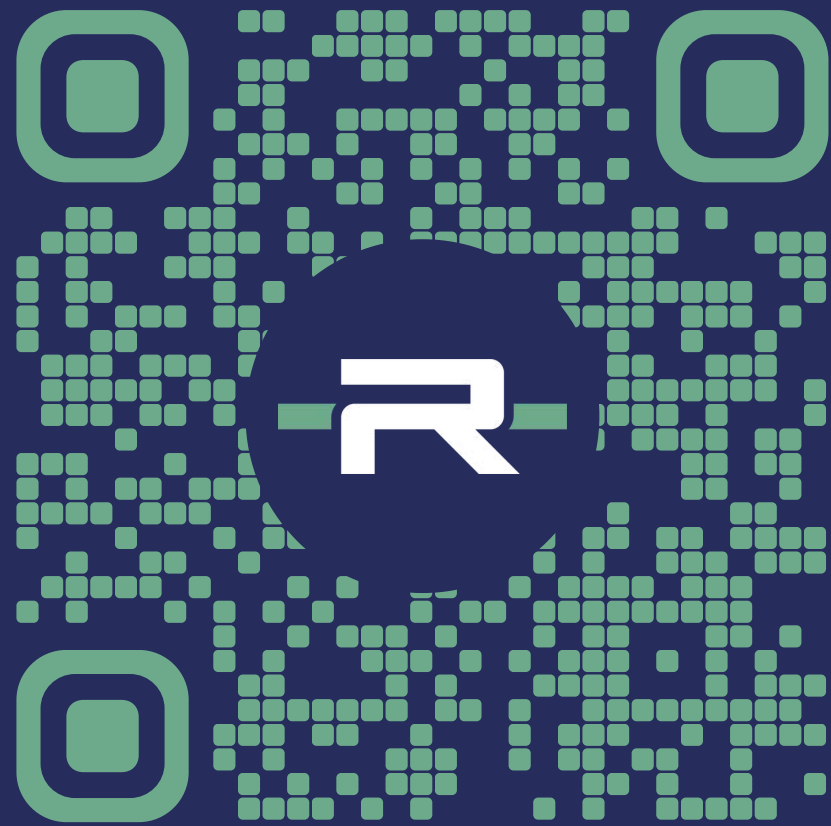




Everything within Reach

Future Subsea Robotics

Link to Slides



Reach - LinkedIn



George - LinkedIn

