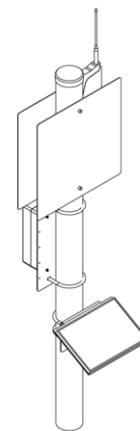




Smart Sign Technology

for Continuous Easement Interference Monitoring



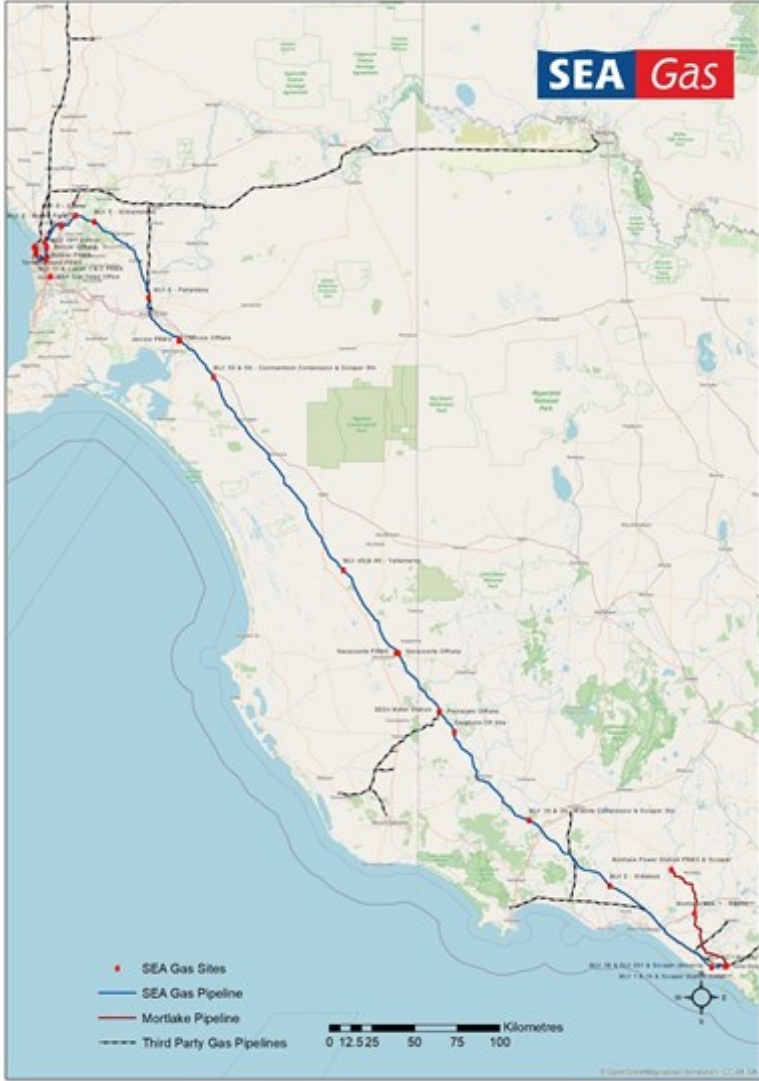
Demi Vlass – SEA Gas

Matthew Lama – Fleet Space Technologies

Dr Johan Barthelemy – University of Wollongong



- Own and operate 800km of high-pressure natural gas transmission pipeline system
- Provide 40-60% of South Australia's natural gas energy needs
- 1184 landholders on easement across SA & Vic
- Seeking to leverage operational technology to *Work Smarter*



Purpose

To enhance pipeline safety management and public safety

- Australian pipeline operating standards are world leading
- Need to maintain high safety standards to avoid potentially catastrophic consequences



[1] Ghislenghien, Belgium: July 2004

Key Public Safety Controls

External interference pose a significant risk to pipelines

- Gas transmission pipelines are buried to reduce possible interference
- Removal of this layer of protection increases risk
- Typical assurance activity is through patrolling the Right of Way, both air and ground, to monitor for threats
- Signage and pipeline awareness education as additional controls



Industry Need

Current patrol methods have limitations

- Aerial and road patrols only detect threats for the duration they are in the vicinity of any section of pipeline
- Patrols are resource intensive
- As easement activity increases so do the number of possible threats



Alternate Technology Explored

- Satellite Photogrammetry
 - Issues with weather
 - High resolution is expensive
 - Economy of scale
- Drones
 - Need to be able to traverse 800kms in any weather
 - Not yet competitive at the scale required

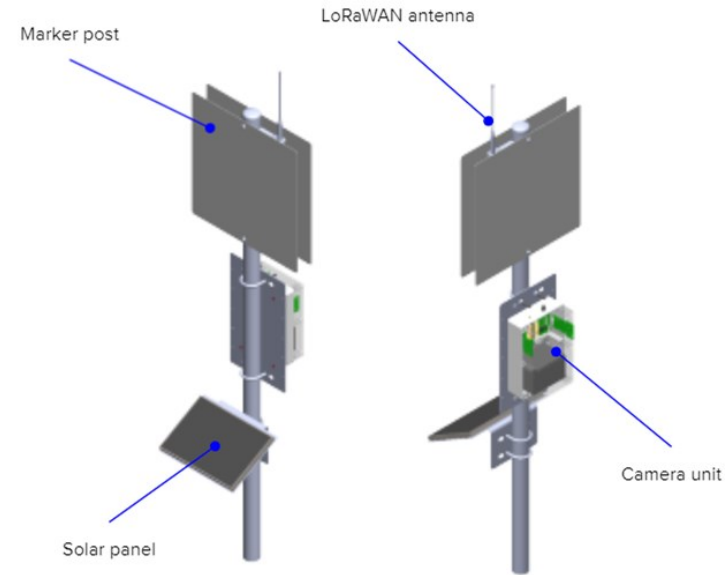


Image from [2]

Proposed Solution

A continuous and intelligent monitoring device – making the sign "smart"

- Sensors on existing pipeline easement marker signs
- Detect, analyse and identify threats using AI
- Identified threats are communicated to pipeline operator

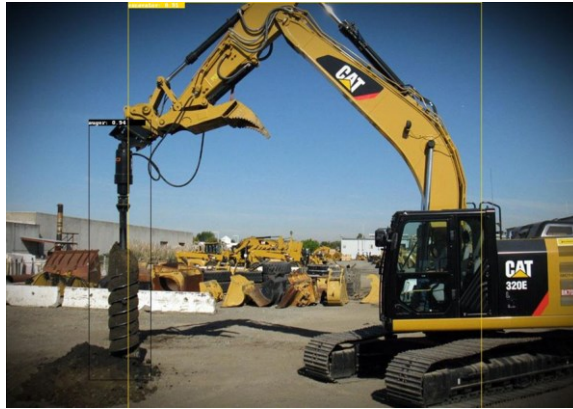


The Artificial Intelligence

Threat identification

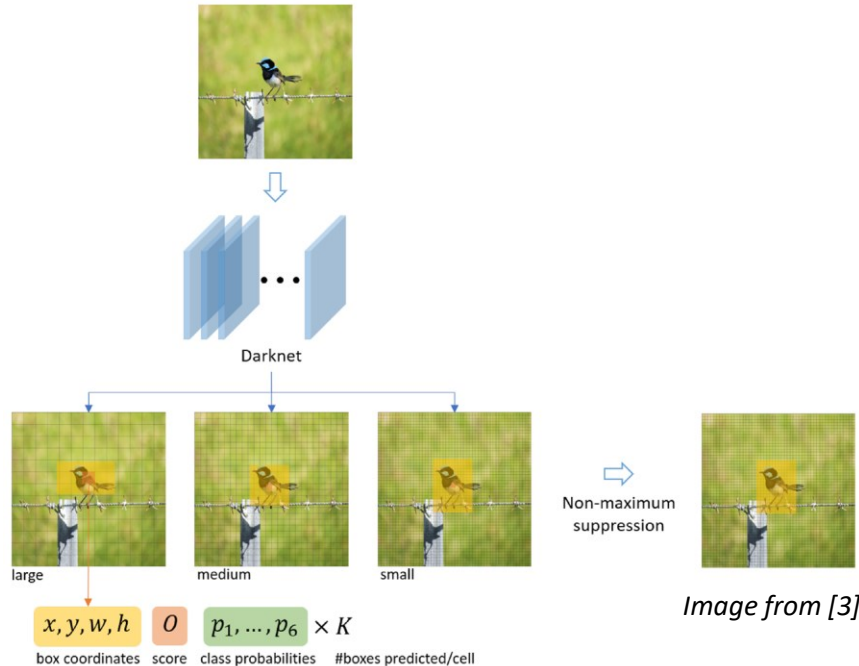
Detecting Threats in Images

An object detection approach



You Only Look Once v4

An object detection approach



- Based on Darknet Deep Neural Network [3]
- 49 millions parameters to optimise
- Pre-trained on ImageNet (14 million images)
- Able to detect objects on 3 different scales

- [3] Barthélemy, J., Verstaavel, N., Forehead, H., & Perez, P. (2019). Edge-computing video analytics for real-time traffic monitoring in a smart city. *Sensors*, 19(9), 2048.
- [4] Bochkovskiy, A., Wang, C. Y., & Liao, H. Y. M. (2020). Yolov4: Optimal speed and accuracy of object detection. *arXiv preprint arXiv:2004.10934*.

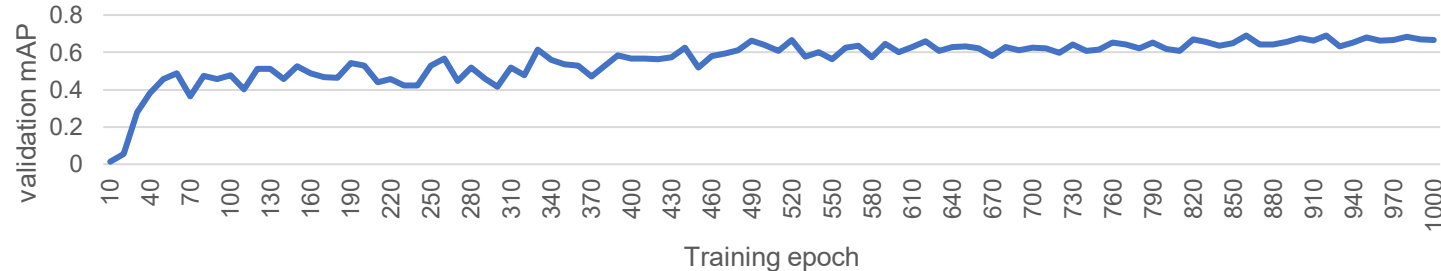
Training the AI

Edge computing architecture

- Database: 10,000+ images of different threats, including

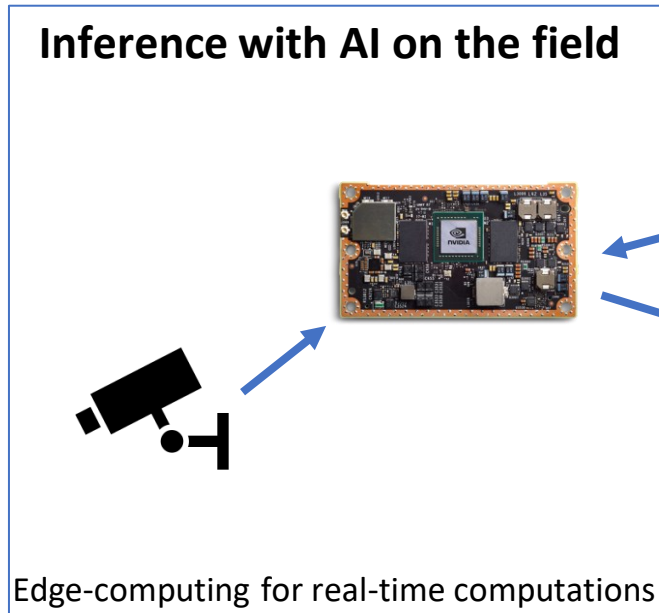
Person	Cable plough	Tractor
Car	Truck	Clay delver
Excavator	Bobcat	Ditch witch
Boring rig	Auger	Post driver

- AI training and validation: Accuracy on par with SOTA – 0.70 mAP



AI Deployment

Edge computing architecture



Trainin



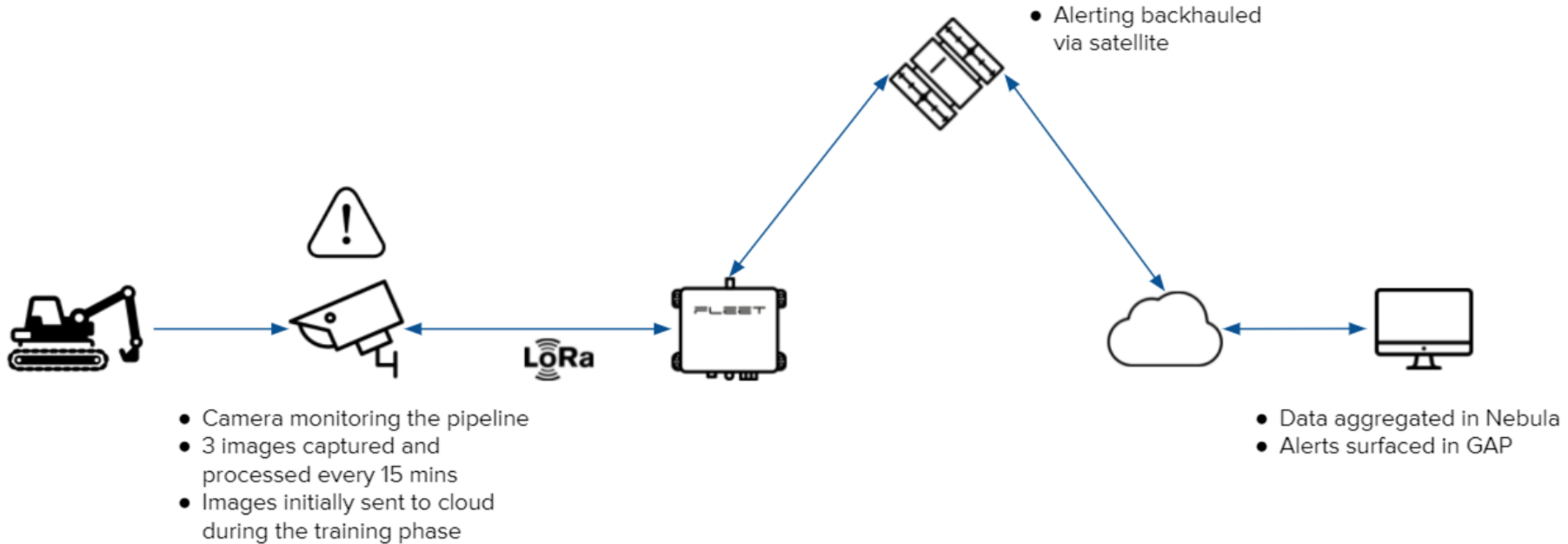
Application



The Sensor

An intelligent edge-computing device

Threat Identification Process



Threat Identified by the AI

- Detection
- Confidence level calculation



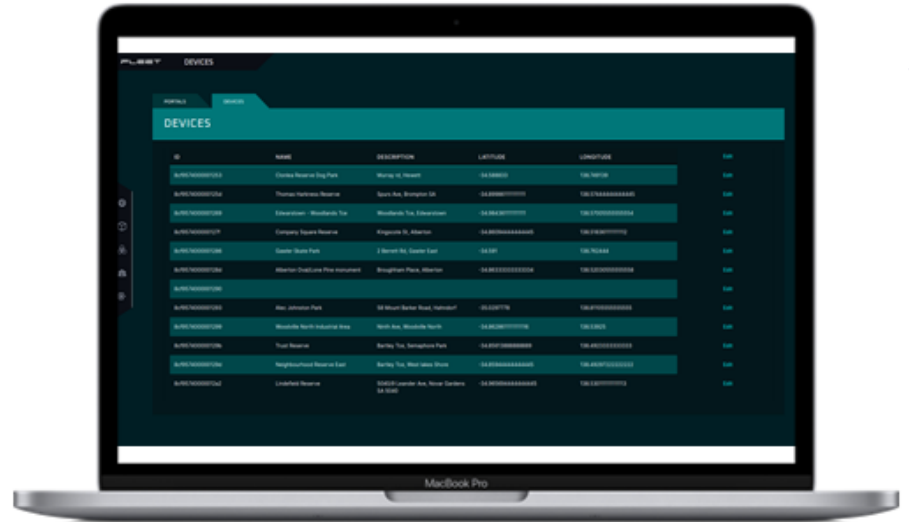
Alert Data Received in the Portal

- Message aggregation
- Satellite backhaul



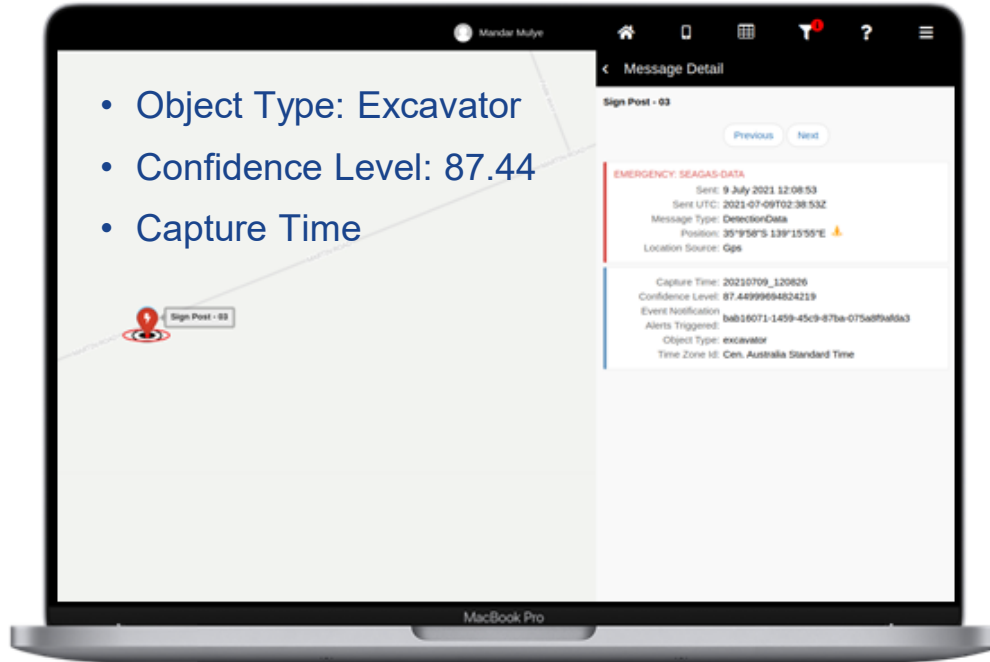
Alert Received in Management Platform

- Network management
- Downstream connectivity



*Alert delivered to operational platform/s, action to be taken as required

Alert Data Visualised



- Object Type: Excavator
- Confidence Level: 87.44
- Capture Time

- Surfacing alerts for users
- Driving notifications

Staff Receive Alert

- Call to action



*Alert delivered to operational platform/s, action to be taken as required

Challenges

X Pink tint and unstable brightness



✓ Pink tint removal and brightness control



X Before calibration



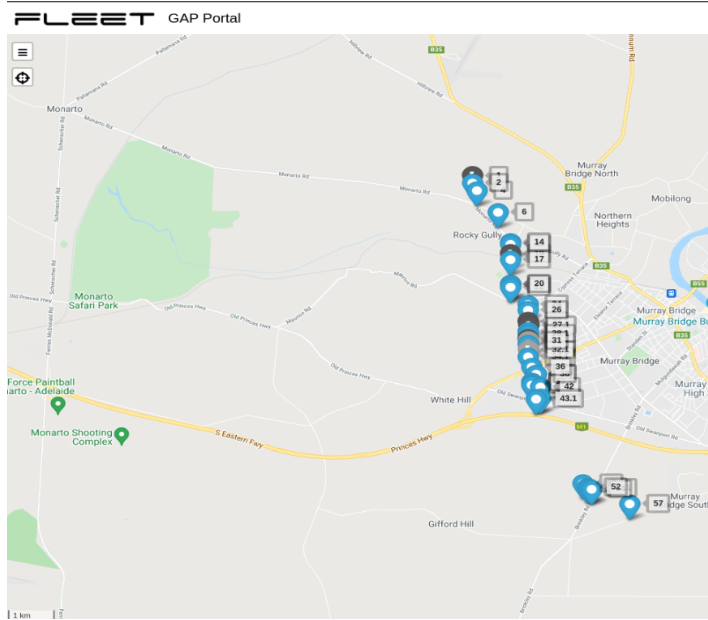
✓ After calibration



Field Test and Demo

Current Deployment

3 Portal and 48 sensors



Demo



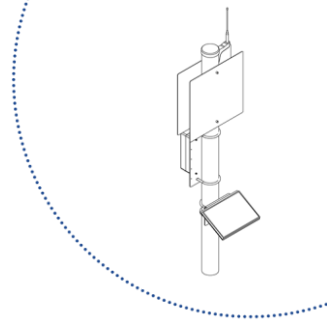
Conclusions and Future Work

Conclusions

- **AI successfully trained**
 - 10k images
 - mAP of 70%
 - 12 types of threats
- **End-to-end solution currently being tested in the field**
 - 48 devices and 3 portals deployed
 - Long-range data transmission



Next steps



Further training the AI

- Collecting more data
- Installing temporary devices

Sensors

- Review the unit design with a focus on:
 - Deployability
 - Power consumption
- Device evaluation
 - Testing in urbanised environment
 - Integration of improvements and learnings into the next generation of sensor



Enabling the decarbonisation of Australia's energy networks

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