



## Industrial Use Case and Lessons Learned

**UDRC – Deep Learning and Defence – 14<sup>th</sup> Nov 2019**

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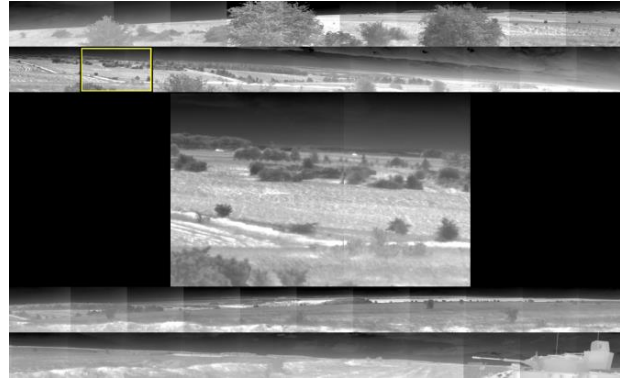
# Introduction

# Introduction – What is the customer problem ?



Mission & comms

+



Watch this ...

+



Remember this ...

## Cognitive overload

- Multiple sensors with no spare capacity to observe and exploit

## Desire to

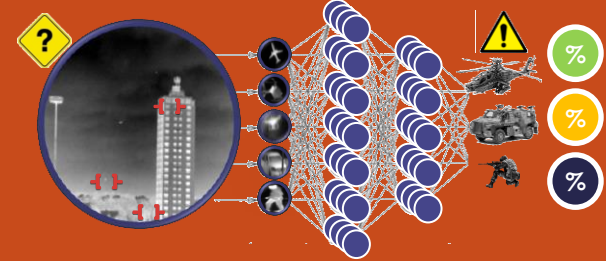
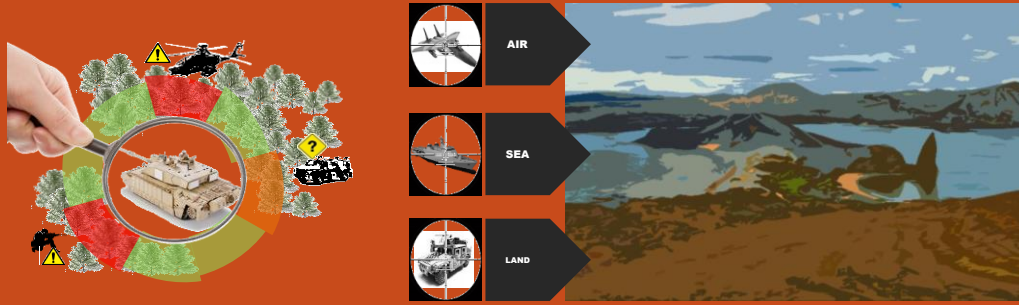
- Automate the observation of sensors
- Look deeper into images
- Receive warnings and alerts of threats

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# Introduction – PhD/EngD

## Context and scene understanding



Patrick Harding, Task relevant image content segmentation for compression (2005-2009)

1<sup>st</sup> EngD

Calum Blair, Real-time video scene analysis with heterogeneous processors (2008-2012)

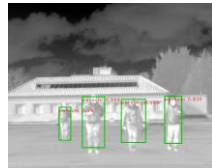
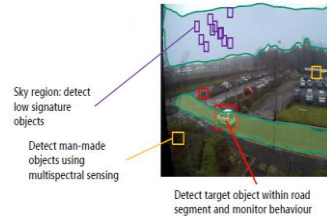
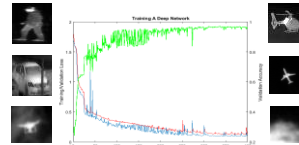
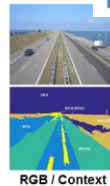
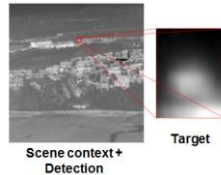
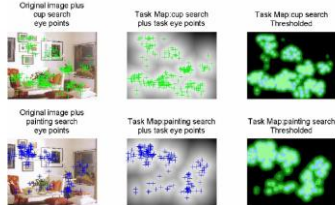
Jonathan Letham, Context based image segmentation (2008-2011)

Chris Dickson, Analysis of IR polarisation signatures for vehicle detection (20011-2015)

Iain Rodger, Target classification in multimodal video using CNN and context (2012-2016)

Impact on Thales AI strategy

Rachael Abbott, Deep and intelligent scene understanding (Queens University Belfast) (2017- )



Classifier Output										
Class	C1	C2	C3	C4	C5	C6	C7	C8	C9	AM
C1	0	0	0	0	0	0	0	0	0	0
C2	0	0	59	0	233	163	4703	5146	0	0
C3	0	14	748	0	5	5	0	0	0	0
C4	0	0	0	0	0	0	0	0	0	0
C5	0	0	0	0	0	0	0	0	0	0
C6	0	3	0	0	82	2096	50	2833	0	0
C7	0	0	0	0	0	0	0	0	0	0
AM	0	17	608	0	110	2064	4703	5146	0	0

Classifier Output										
Class	C1	C2	C3	C4	C5	C6	C7	C8	C9	AM
C1	0	0	0	0	0	0	0	0	0	0
C2	2	4623	124	0	232	163	0	5146	0	0
C3	0	4	748	0	5	5	0	0	0	0
C4	0	0	0	0	0	0	0	0	0	0
C5	0	0	0	0	0	0	0	0	0	0
C6	0	15	0	0	82	2096	0	2833	0	0
C7	0	0	0	0	0	0	0	0	0	0
AM	7	4602	886	0	316	2064	0	4703	0	0

Natalie Flaherty, AI for land platforms (2018- )

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# Introduction – Thales TrUE AI approach

## Transparent AI

- where users can see the data used to arrive at a conclusion.

## Understandable AI

- that can explain and justify the results.

## Ethical AI

- that follows objective standards protocols, laws and human rights.

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# AI Across Thales UK





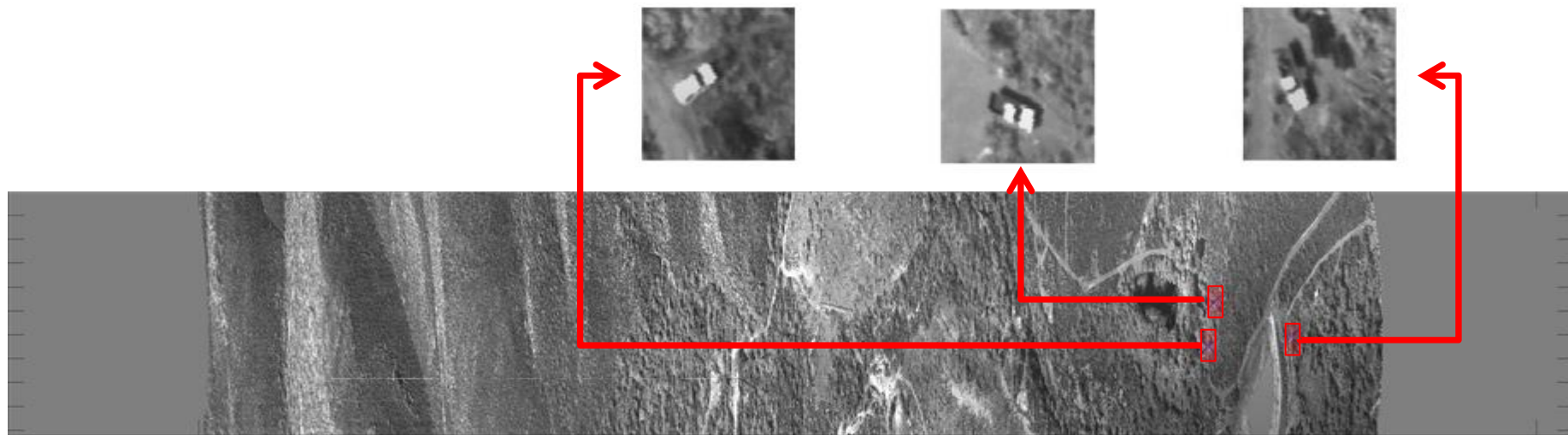
# AI Across Thales UK – Impact on air to ground image analysis

## Two highly trained human analysts

- 30 minutes
- 31 identified objects

## CNN algorithm

- 0.5 seconds
- 36 identified objects
  - Including all 31 identified by analysts
  - Other 5 confirmed as misses by analysts



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# AI Across Thales UK – Environments



## Tailored Machine Learning across many domains

- Airborne reconnaissance
- Maritime
- Thermal & Visible data
- Small targets in high resolution imagery



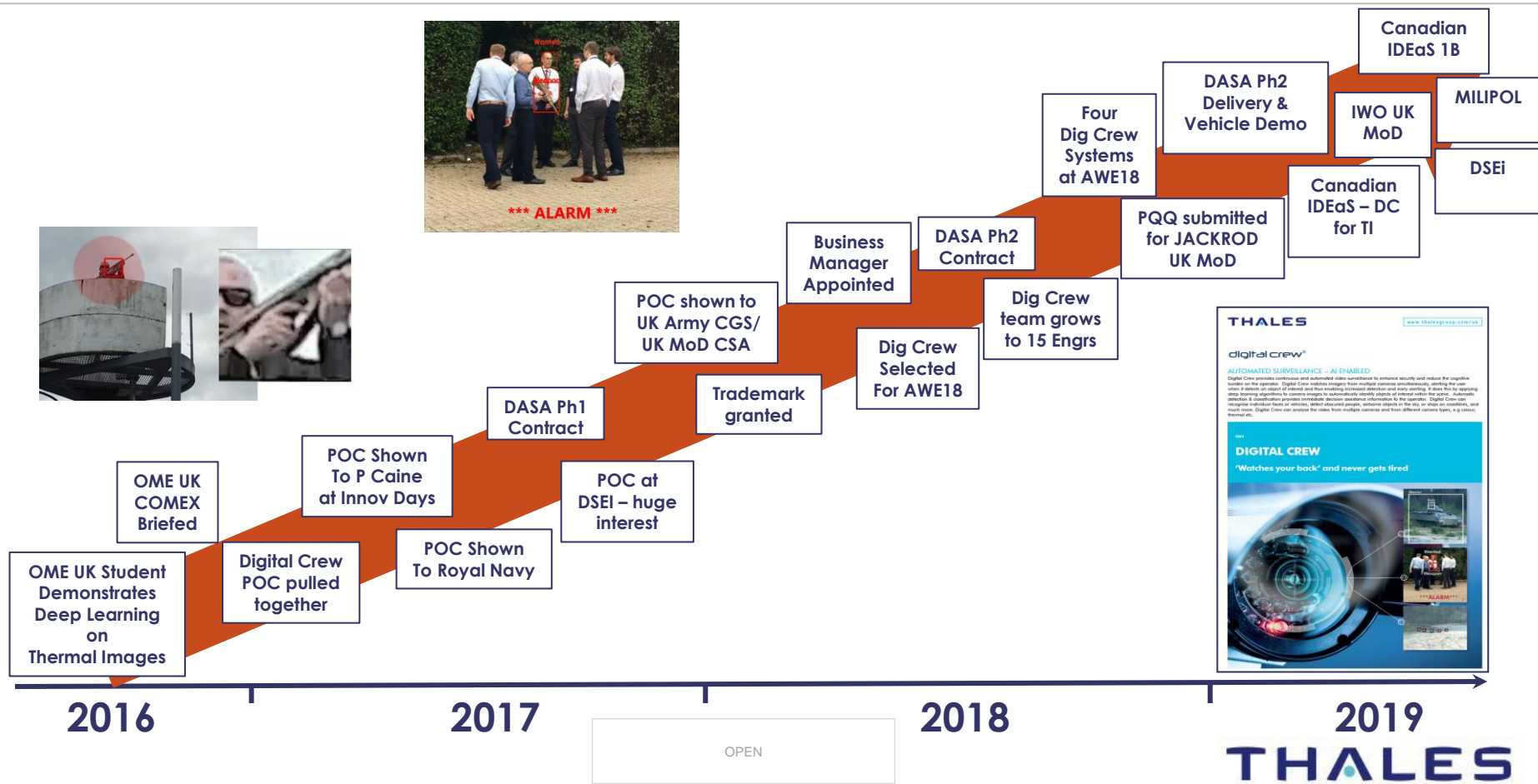
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## Academia & Research



## Development & Users



Defence and Security  
Accelerator



Australian Government



Government  
of Canada

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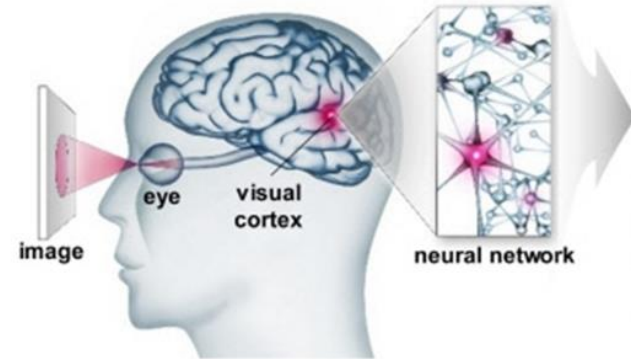
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# AI Across Thales UK – Digital Crew<sup>®</sup>

## COTS & MOTS hardware & software, open standards



Linux



## Free, permissive licences, readily available and huge community support

## CNN - YoloV3, YoloV3-tiny, ResNETxx, ...



Adapted for the military domain, stand alone & no external links

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## ■ Potential Use Case

➤ <Video>

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## Challenges



# Challenges – Performance Metrics

## Need confidence in performance

- Models need to generalise to the real world
- Implies larger test sets compared to training sets → Opposite of literature?

## Metrics, metrics, metrics

- Customer
- Business
- Technical

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# Challenges – Following Published Literature

## ■ All that glistens is not gold

- Some published methods are very hard to re-create
- Datasets may have unidentified bias
- Papers do not always provide enough explanation to reproduce results

## ■ Lessons Learned

- Deep understanding of datasets, and pro-active identification of bias
- Reproducible development environment
- Configuration control of experiments

# Challenges – Adaptation Required

## Open source doesn't mean the work is done!

- Domain specific use cases
- Constrained environments

## Re-implementation for deployment

- Minimise Size, Weight, Power and Cost – SWaP C
- Reduce memory footprint
- Enable use of higher resolution imagery
- Increase frame-rate & reduce latency

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# Thank you



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