



DAEDALEAN

The Flightpath To Autonomy

Goes Through OPV

IFATCA Geneva 2024-03-22

Dr. Luuk van Dijk
CEO & Founder • Daedalean

We are Daedalean

150+
staff

18
PhDs

17
pilots



In 2016, we set out to build the world's first fully autonomous flight control system...

...making flying safer, scalable and smarter by replacing and outperforming the human pilot on every measurable dimension



Source: [FAA-S-8081-16B](#)

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1

Aviate

Perception; control; aircraft state monitoring. Always know where to land and avoid obstacles

2

Navigate

Find optimal flight path, use existing air spaces, traffic patterns

3

Communicate

Talk to air traffic controllers; understand other pilots

4

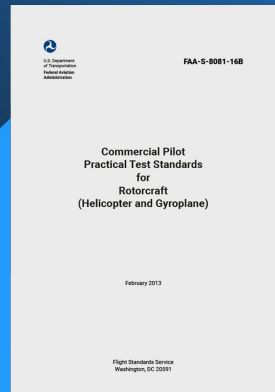
Decide

Real-time decisions on board in normal and contingency situations

5

Plan

Plan the mission, modify the mission — pre-flight and en-route

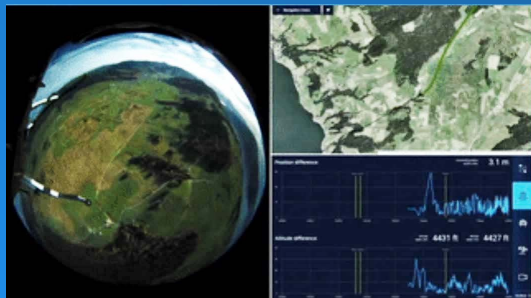




VXS – Daedalean's Computer Vision suite to assist and to pilot in VMC

Making decisions and judgement calls that are currently the exclusive domain of the human pilot

Where am I?



Visual Navigation

High-precision positioning in GPS-denied situations

Where can I fly?



Visual Traffic Detection

Detection of cooperative (e.g. planes) and non-cooperative (e.g. drones, bird flocks) airborne hazards

Where can I land?

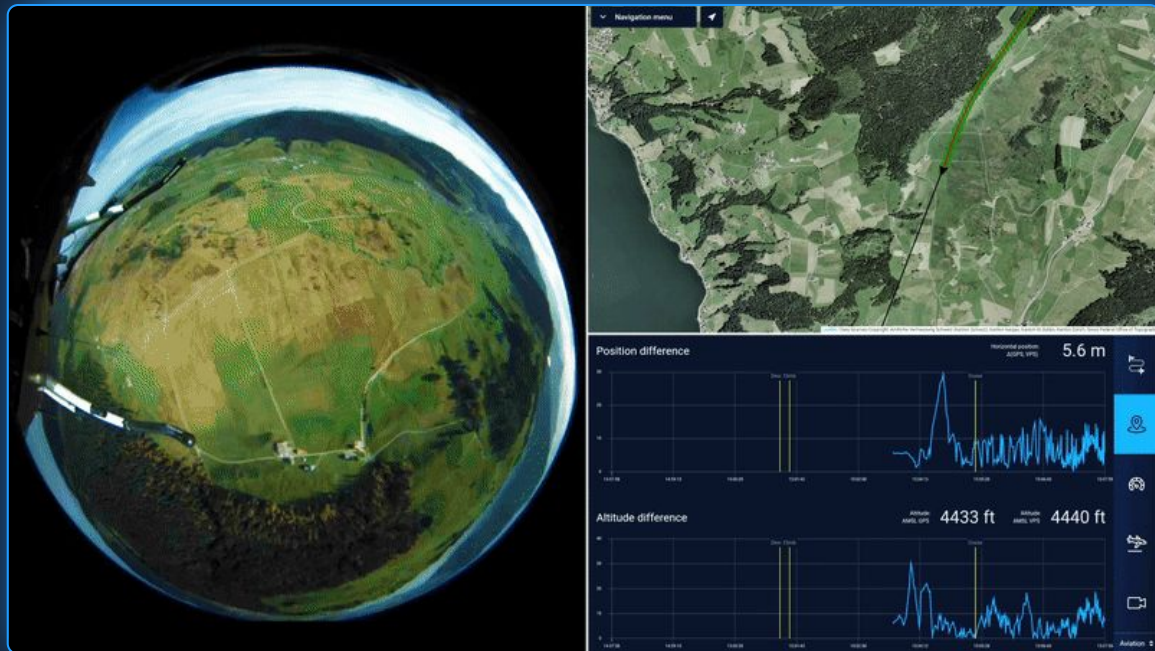


Visual Landing Guidance

Runway and helipad landing guidance

Visual Positioning

Camera-based navigation and landing guidance



Functions

Continuously outputs position, heading, velocities, height above ground, landing guidance and corresponding uncertainties

Properties

Navigation data on par with high-end INS/GNSS and ILS is designed to integrate with existing INS-/GNSS-based navigation systems while being dissimilar

Safety certification level

DAL-C+



Visual Traffic Detection

Noticing all fixed wing, rotorcraft and drones in time



Functions

- Detects and tracks traffic
- Classifies traffic
- Identifies size, position, direction of movement, time to collision

Properties

- Better accuracy than human pilots
- Detection range up to 3.5 km
- Up to 10 Hz
- Integration with existing flight deck instruments as a pilot-aid

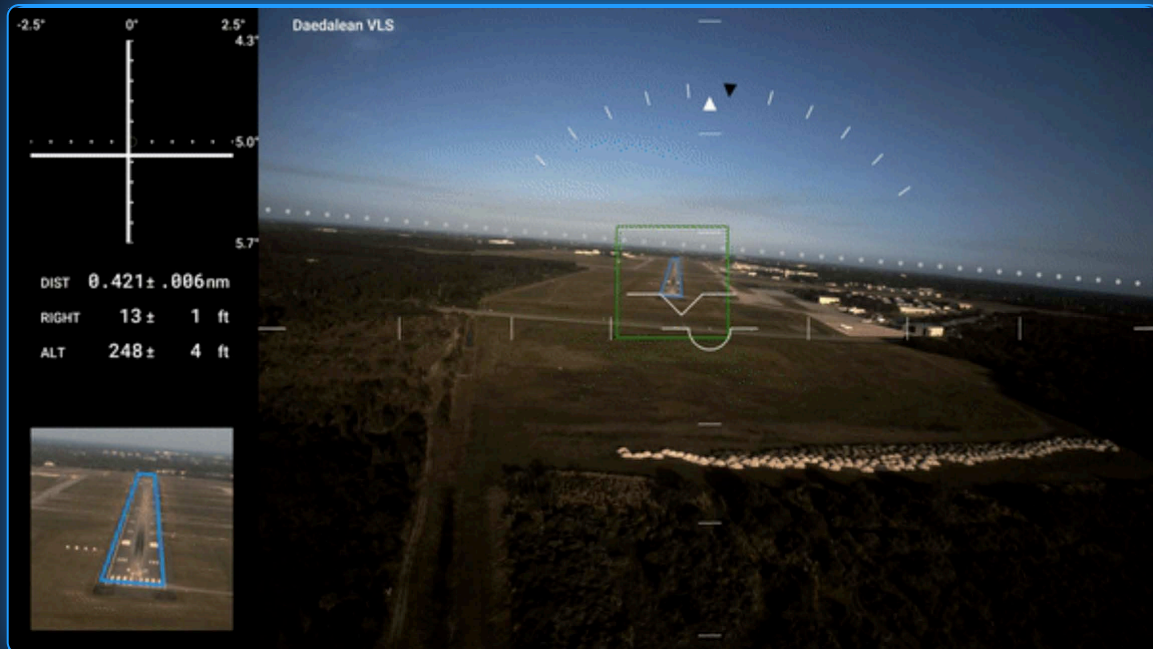
Safety certification level

DAL-C+



Visual Landing Guidance

Onboard GNSS-independent landing guidance for VTOL and fixed-wing aircraft



Functions

- Lists runways, helipads or landing markers in view
- Provides guidance on final approach to touchdown to the selected runway, helipad or landing marker

Properties

- Continuous real-time tracking of ownship position relative to the runway, helipad or landing marker
- No equivalent instrument exists

Safety certification level

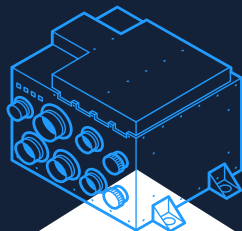
DAL-C+ target

Underlying ~~magic~~ technology

Certifiable in-house machine-learned algorithms (neural networks)

System level

Cameras



Aerospace-grade
computing
hardware

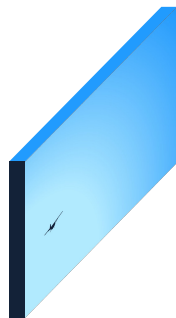
Updated path plan
for the pilot or for the
FCC

Control

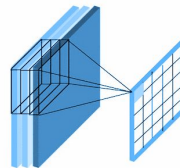
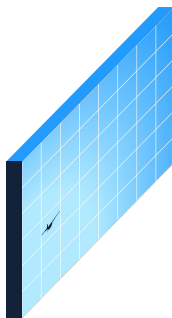
Software level



Camera



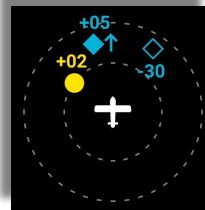
Preprocessing



Certifiable
Convolutional
Neural Network

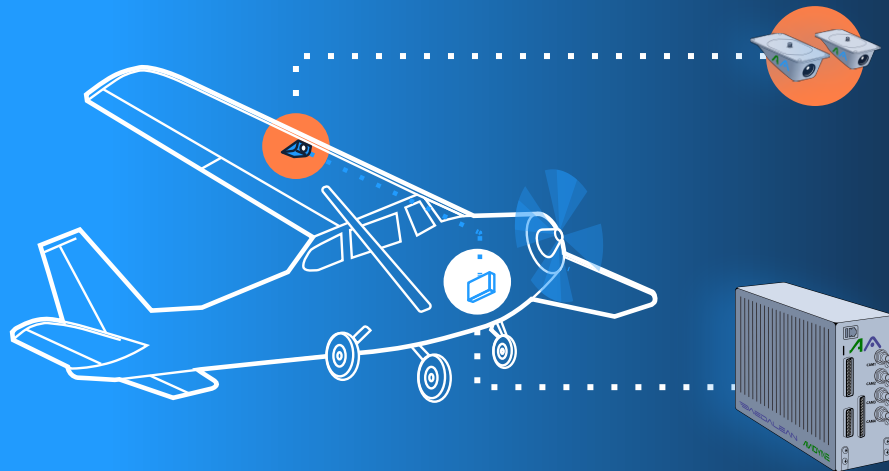


Post-processing



Certification

- DAL-C certification of Traffic Detection the first vision-based ML application
- Required creating new methodology for showing fitness for purpose and absence of unintended function



We are here
2024-03



1
A concept for Learning Assurance proposed, used by EASA in their "First usable guidance for Level 1 AI/ML application"

2
STC application for Traffic Detection

3
FAA issues Issue Paper, EASA issues a Certification Review Item

4
A specific application gets certified, referencing ML policy as means of compliance



Paving the way to certify the use of AI in safety-critical avionics



Federal Aviation
Administration

Executive summary

This report is one of the main outcomes of a joint research project between Daedalean AG and the Federal Aviation Administration, carried out between September 2020 and September 2021. The subject of the project was the study of a Hybrid Neural Network (HNN) for the certification of an Artificial Intelligence (AI) Machine Learning (ML) Neural Network.



Figure 3: An aircraft cockpit display showing a flight path and various instruments.

Following up on the two joint reports [CodaNN 2020, CodaNN 2021] by EASA, the goal was to provide a comprehensive overview of the state of the art in the field of AI in safety-critical applications, as well as to inform future work on the use of machine learning and AI in safety-critical applications. The report is structured as follows: Chapter 1: Introduction. This chapter provides an overview of the project and its objectives. Chapter 2: Introduction. This chapter provides an overview of the project and its objectives. Chapter 3: Introduction. This chapter provides an overview of the project and its objectives.

Figure 3: An aircraft cockpit display showing a flight path and various instruments.

1 Introduction

1.1 Goals and scope of the project

The subject of this report is a joint research project between Daedalean AG and the Federal Aviation Administration, carried out between September 2020 and September 2021. The subject of the project was the study of a Hybrid Neural Network (HNN) for the certification of an Artificial Intelligence (AI) Machine Learning (ML) Neural Network.

Chapter 2 Introduction

2.1 Background

In recent years, the scientific discipline known as machine learning (ML) has become increasingly important in the field of safety-critical applications. This is due to the fact that ML algorithms are able to learn from data and make predictions based on that data. This makes them very useful for tasks such as fault detection, anomaly detection, and predictive maintenance. However, the use of ML in safety-critical applications is still a challenge, as there are many factors that can affect the reliability of these algorithms. This report aims to provide an overview of the state of the art in the field of ML in safety-critical applications, as well as to inform future work on the use of ML in safety-critical applications.

- "Traditional Development Assurance Frameworks are not adapted to the challenges of ML."
- "Difficulties in keeping a comprehensive description of the overall behavior of the system under development."
- "Lack of predictability and explainability of the ML application behavior."
- "Lack of guarantee of robustness and of no 'surprise' behavior."
- "Lack of standardized methods for evaluating the operational performance of ML applications."
- "Issue of bias and variance in ML applications."
- "Complexity of architectures and algorithms."
- "Adaptive learning processes."

This report investigates these challenges in more detail. The current state of the art in the field of ML in safety-critical applications is reviewed. The challenges are identified and discussed. The report also provides an overview of the state of the art in the field of ML in safety-critical applications, as well as to inform future work on the use of ML in safety-critical applications.

EASA Daedalean - EASA CodaNN IPC EXERCISE - Chapter 2

2.2 Learning Assurance process elements

The report has identified several elements for Learning Assurance in a V-shaped development cycle. It is a process to ensure that the quality and completeness of the development cycle is maintained throughout the project. The process is based on the V-model, which is a common model for software development. The V-model is used to ensure that the development cycle is maintained throughout the project. The process is based on the V-model, which is a common model for software development.



Figure 2.2: V-shaped development cycle for Learning Assurance. The diagram shows a V-shape with various stages of development and testing. The left side of the V represents the development phases, and the right side represents the testing phases. The bottom of the V represents the final product. The stages are labeled with various terms related to development and testing.

2.3 Other key takeaways from the IPC

To meet with the challenges in keeping a comprehensive description of the intended function, the report identifies several key takeaways from the IPC. These takeaways are based on the V-model, which is a common model for software development. The process is based on the V-model, which is a common model for software development.

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EASA and Daedalean' joint reports: [CodaNN 2020](#), [CodaNN 2021](#)

introduce the W-shaped Learning Assurance process for validation of deep neural networks in safety-critical applications.

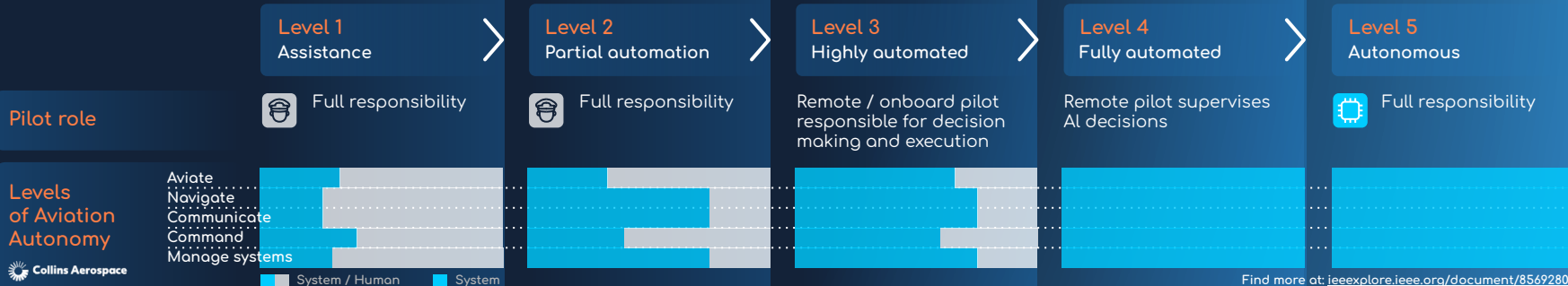
[The FAA and Daedalean joint report](#): evaluates how the W-process fits the FAA's future certification policy.

“ System engineering has its V — now machine learning has its W ”

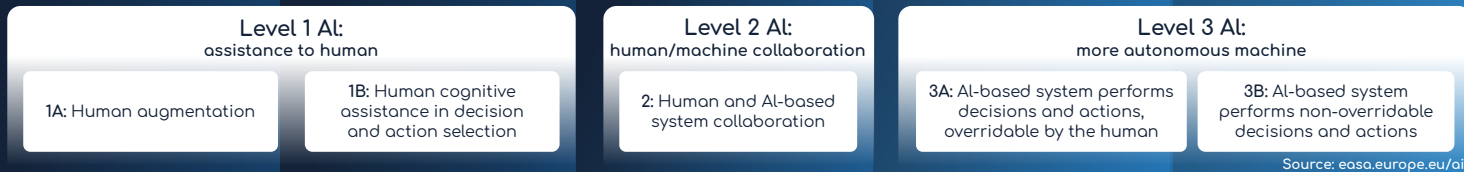
Graham Warwick
Managing Editor,
[Technology with Aviation Week](#),
aka [The Woracle](#)



Stages Toward Autonomy



Classification of AI applications



Technical roadmap



Aviation certification level

Interface

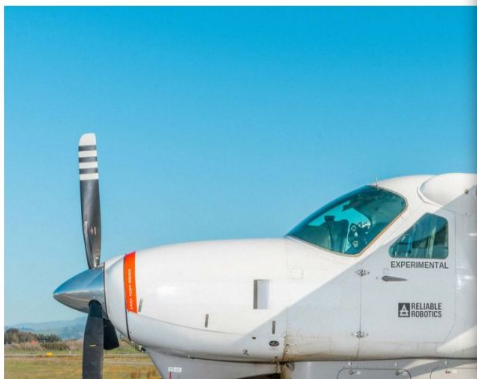


Automate or autonom-ate?

This Startup Wants to Fill the Skies With Cargo-Filled Robot Planes

● Reliable Robotics has been flying Cessnas through a mixture of autonomy and remote control. FedEx, for one, is a believer.

By Ashlee Vance +Get Alerts



Autonomous aviation startup Xwing hits \$400M valuation after latest funding round

Aria Alamalhoda | @breadfrom / 6:00 PM GMT+2 • April 15, 2021

Comment



The safety pilot has his hands off the controls during an Xwing demonstration flight. **Image Credits:** Xwing

Xwing has scored another win two months after it completed its first gate-to-gate autonomous demonstration flight of a commercial cargo aircraft. The company said Thursday it has raised \$40 million at a post-money valuation of \$400 million.

The company is setting its sights on expansion — not only tripling its engineering team, eventually running regular fully unmanned commercial cargo flights.

THE VERGE

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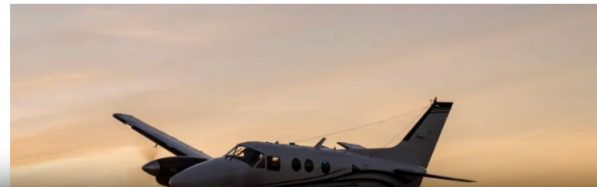
TECH / TRANSPORTATION / AVIATION

Merlin Labs wants to fill the sky with pilotless planes carrying cargo and passengers

Autonomous flight startup announces \$25 million funding round

By Andrew J. Hawkins | @andyjayhawk | May 26, 2021, 12:00pm EDT

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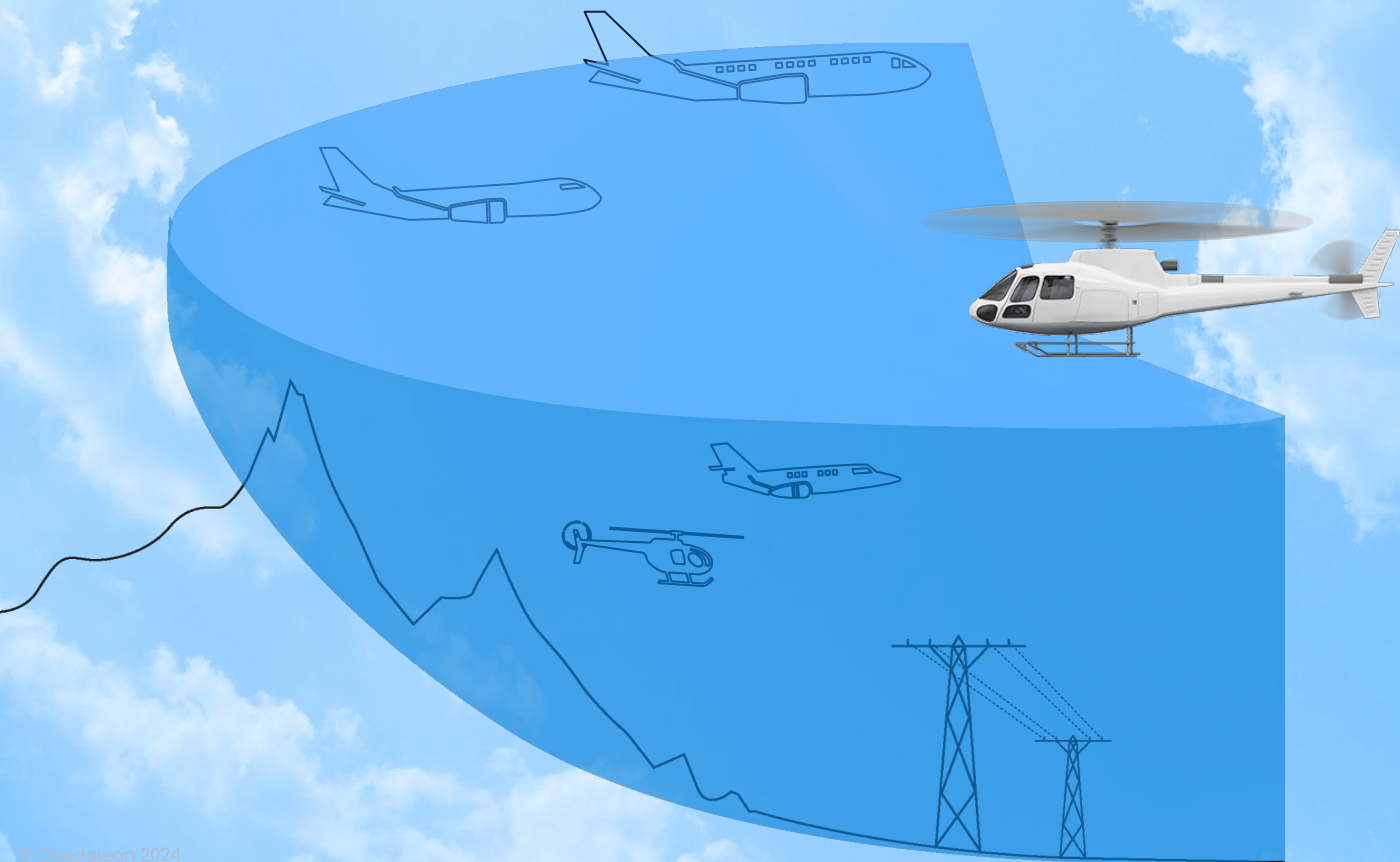
Email (required)

Autonomous flight
technology for unmanned
aircraft.

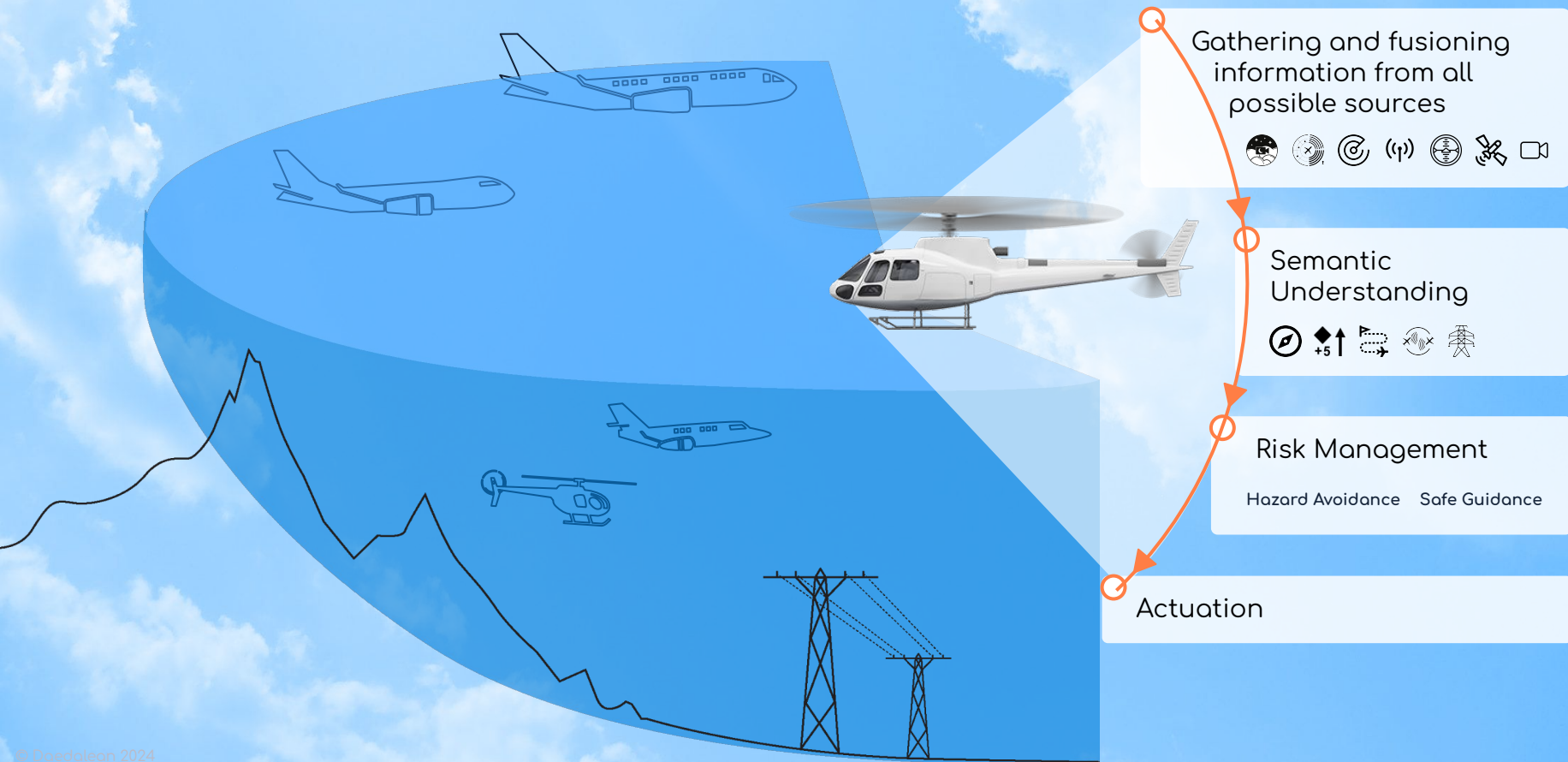
Our software integrates sensors and maps with onboard flight control systems to allow regional aircraft to navigate, take off and land safer than ever before.



What it is that a human pilot actually does?



What should a computer do?





Key capabilities at key phases of flight



- 1 Detect and avoid hazards (on ground and airborne)

Take-off and initial climb



- 2 Emergency landing guidance to a safe site (known and unknown)

Cruise



- 3 Detect and avoid airborne hazards (also non-cooperative)



- 4 Weather awareness



- 5 Position and navigate (also in GNSS-denied environment)

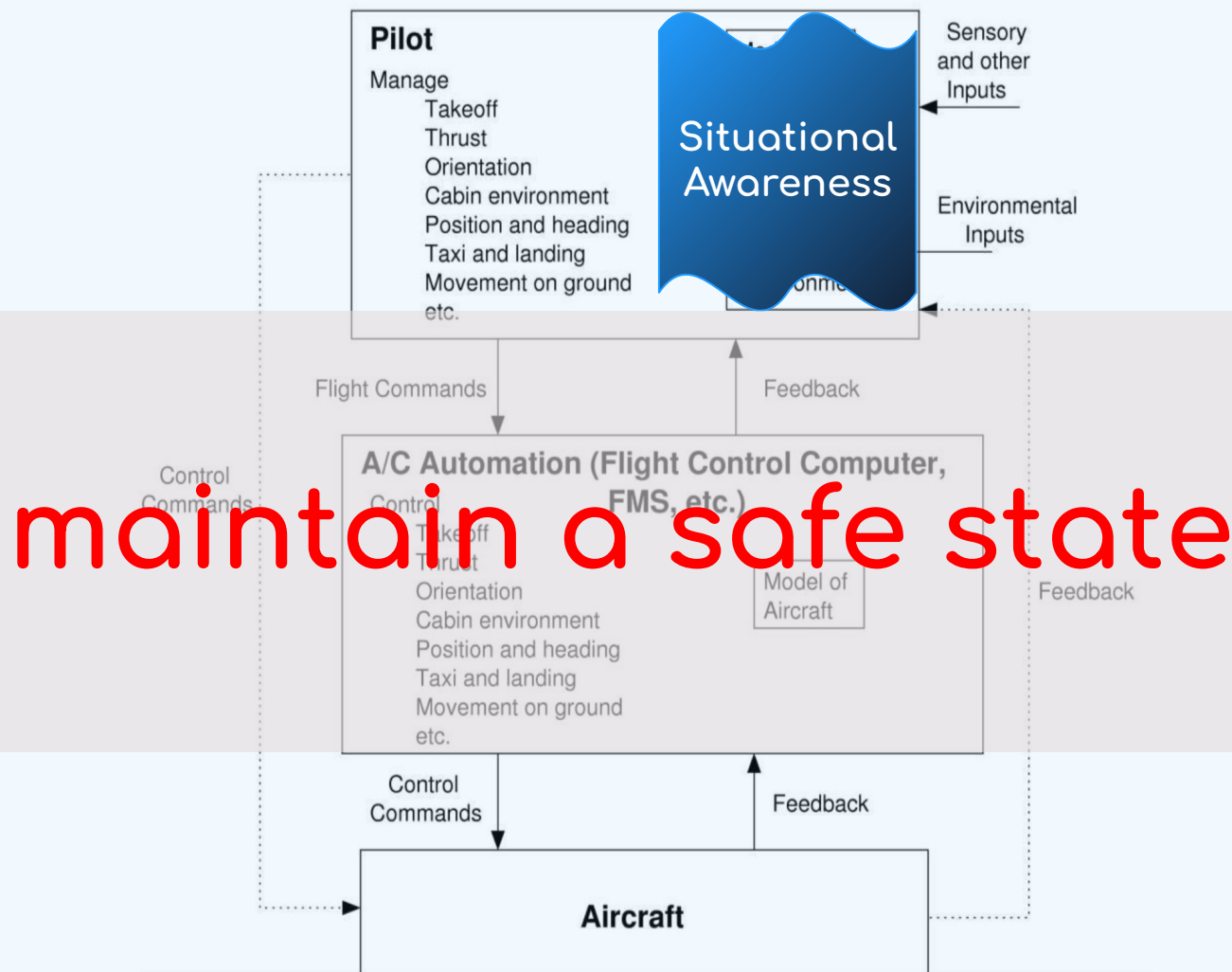


- 6 Precision landing on pre-marked site

Final approach and landing



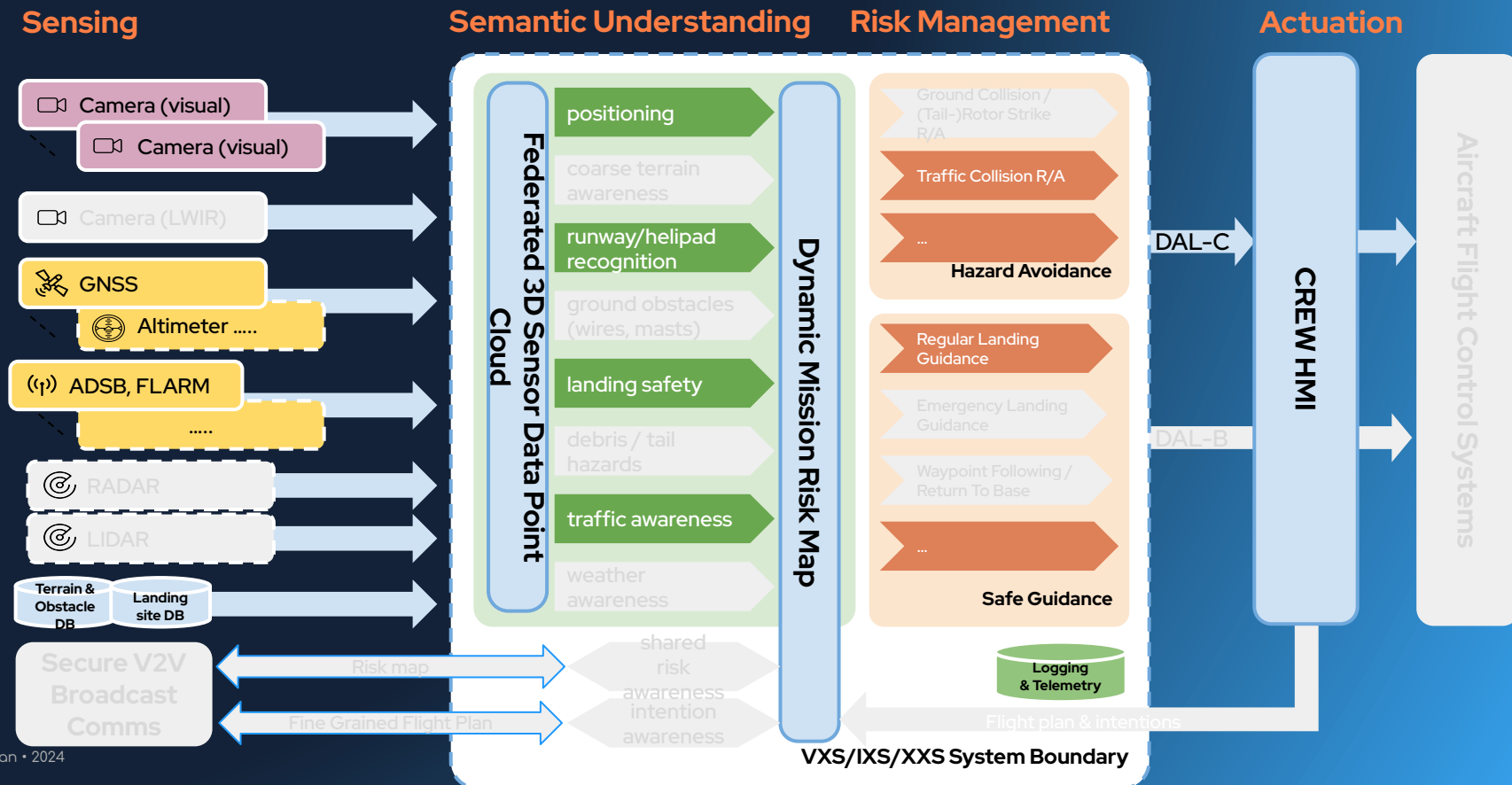
- 7 Safe landing on unmarked site



A Systems-Theoretic
Approach to Safety in
Software-Intensive
Systems.

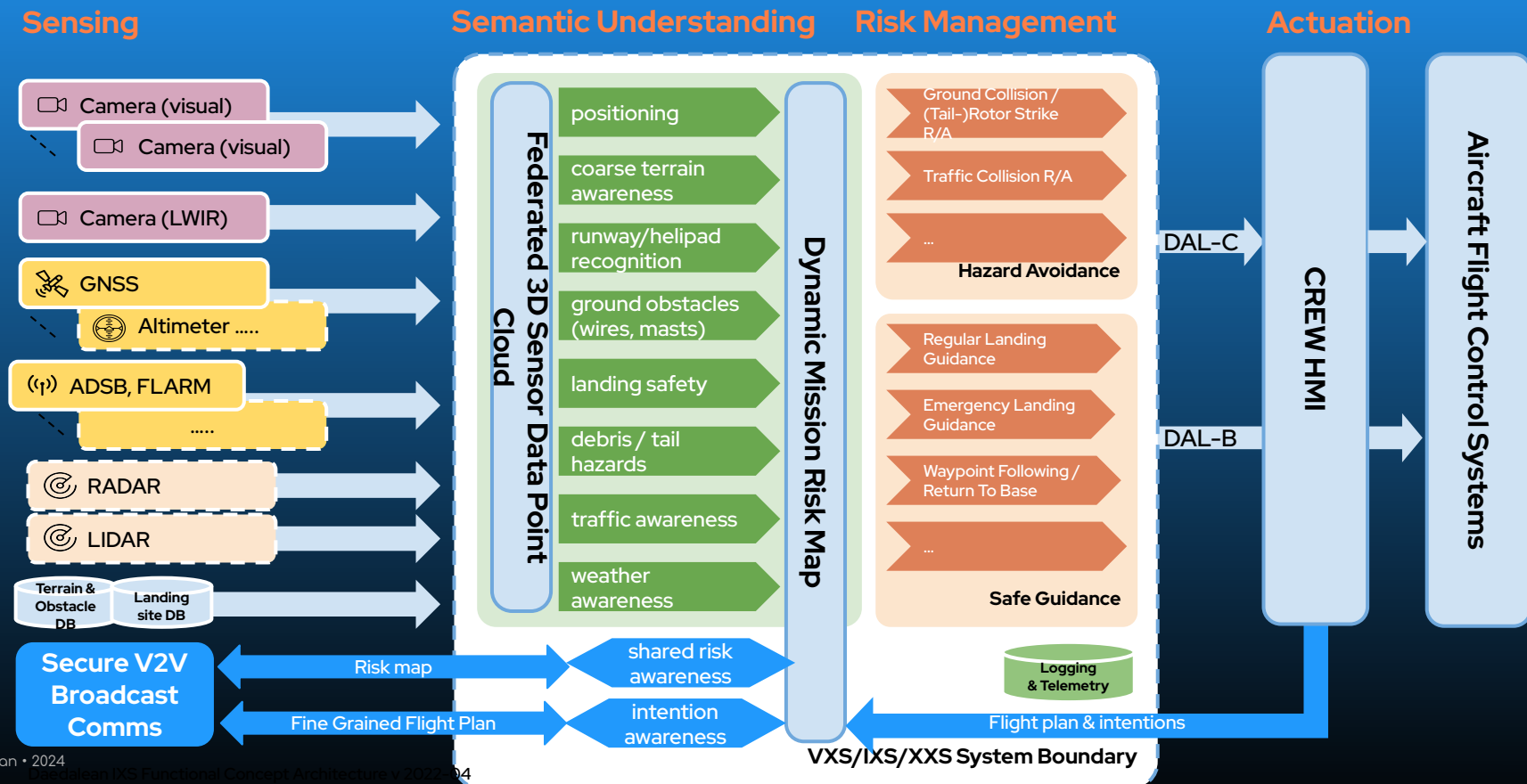
Nancy G. Leveson

Daedalean Situational Awareness Today





Daedalean Situational Awareness Tomorrow





Every single pilot gets a copilot

- Create the aircraft that doesn't kill anyone if you take your hands off the controls
- Make pilot interface with the flight plan, not with the aircraft
- Guiding HMI principle: reduce information overload, don't add to it











Questions?

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