

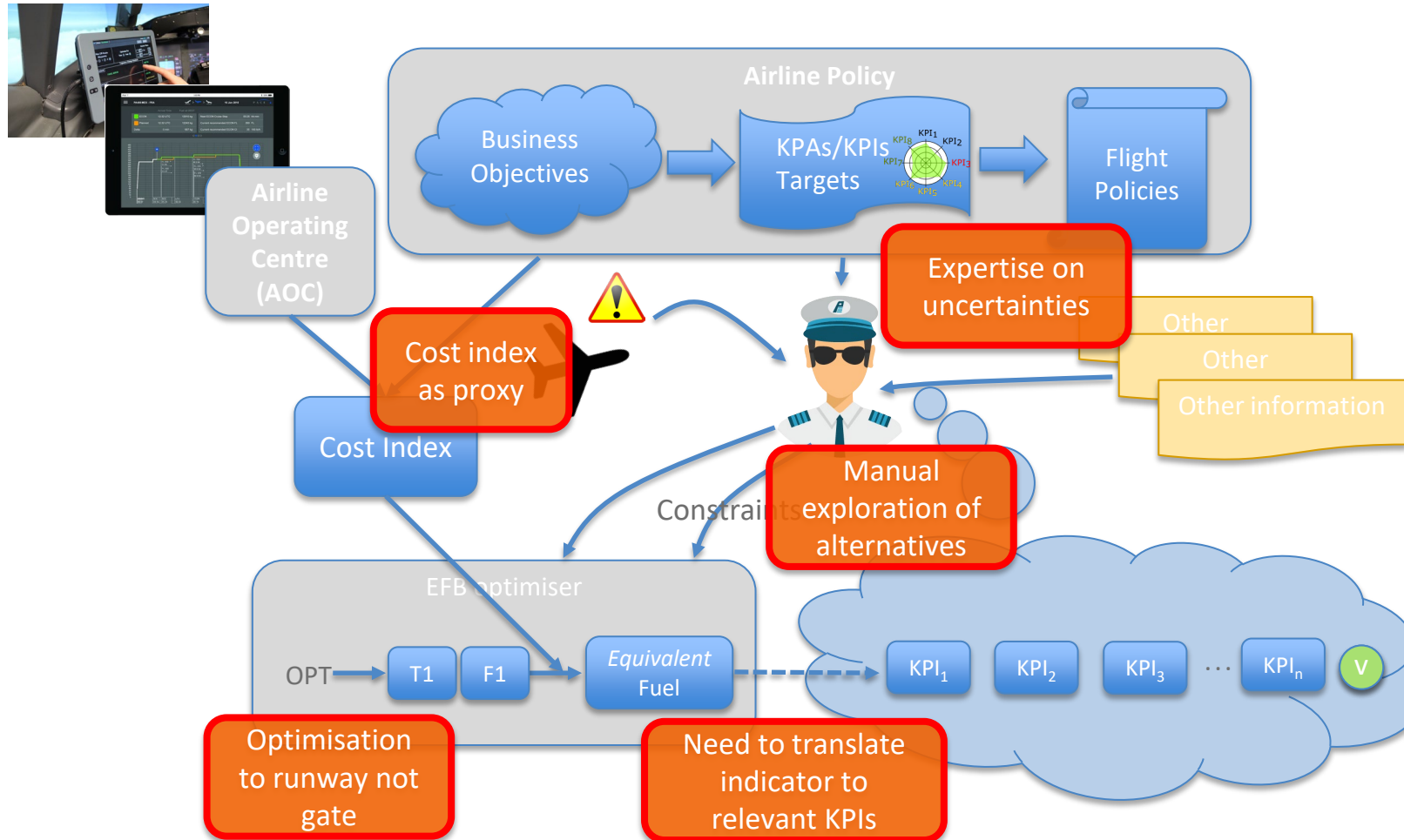
**Autum  
School 2025**

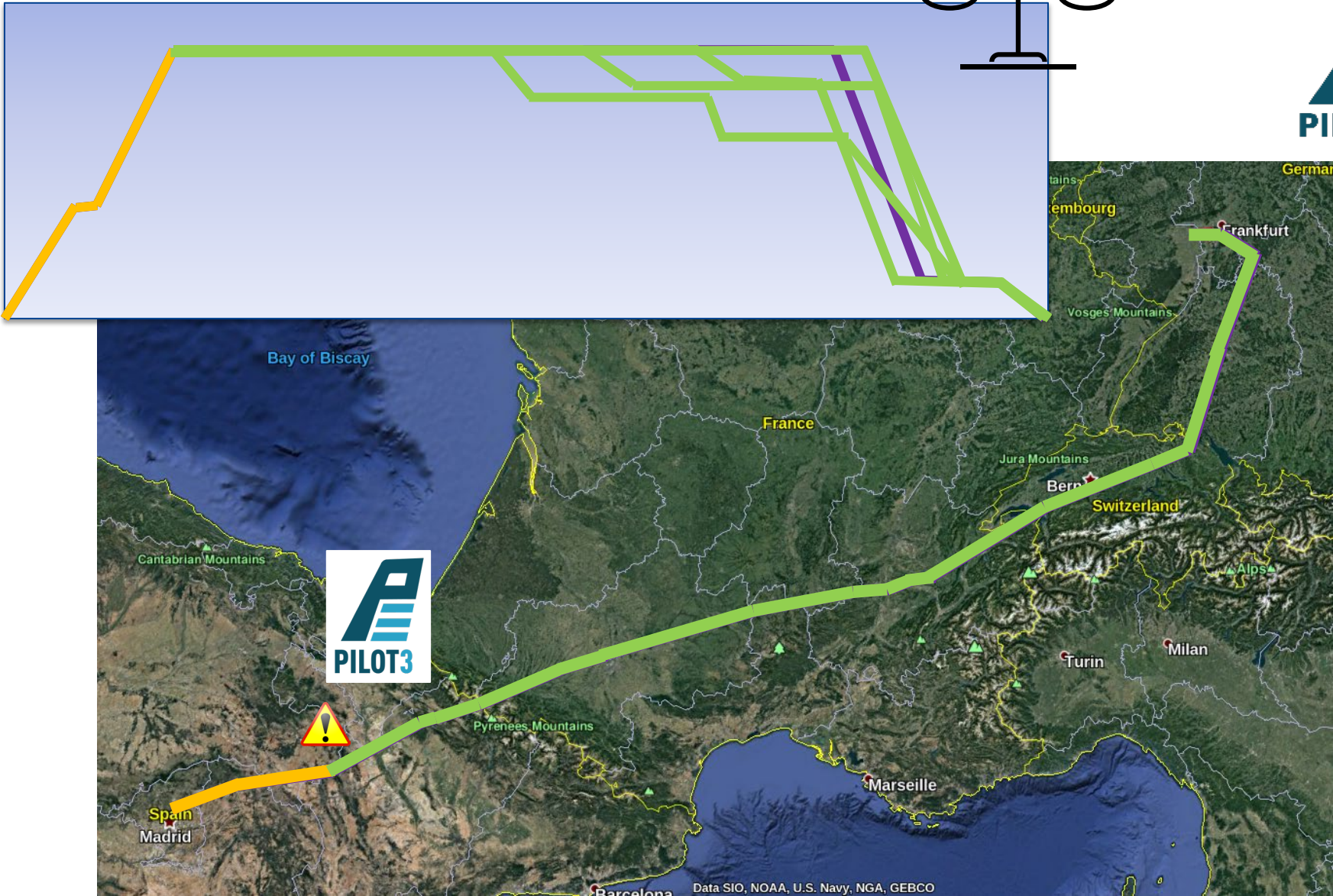
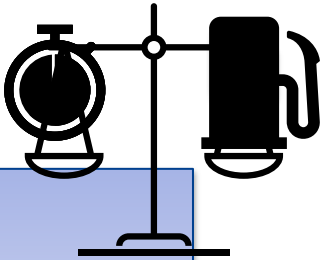
# **Models for flight operations**

## **Examples of ML models and challenges**

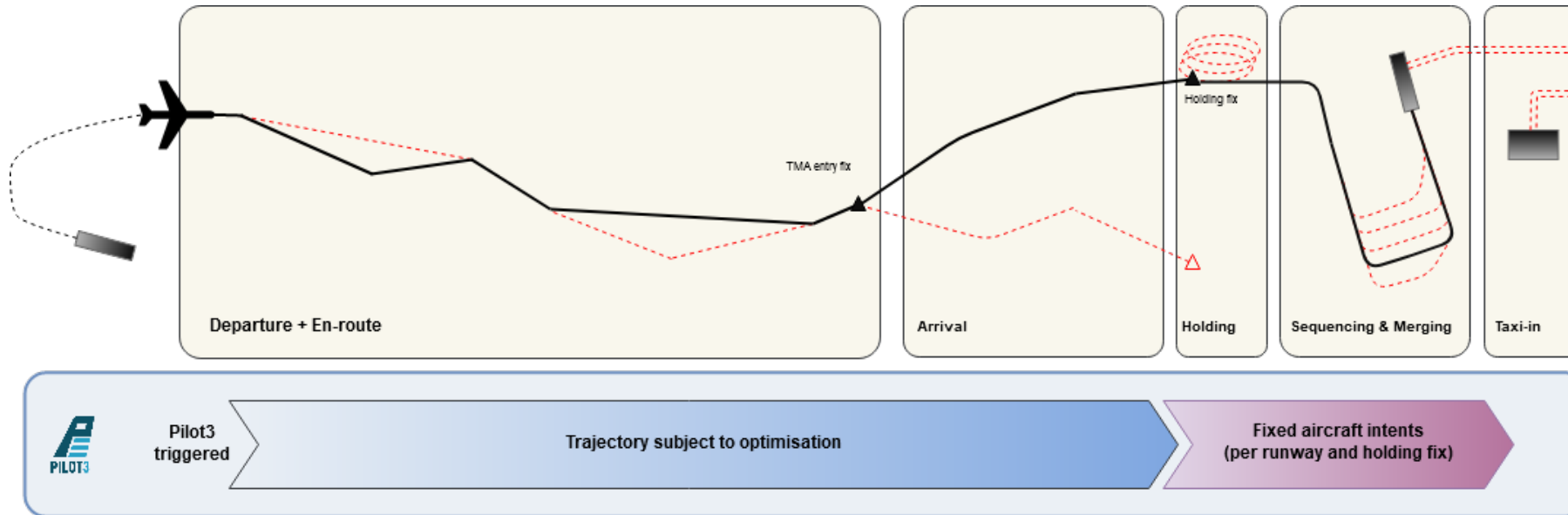
### **Pilot3 and Dispatcher3 projects**

**Dr Luis Delgado**  
**Centre for ATM Research**  
**University of Westminster**  
**27 Oct 2025**

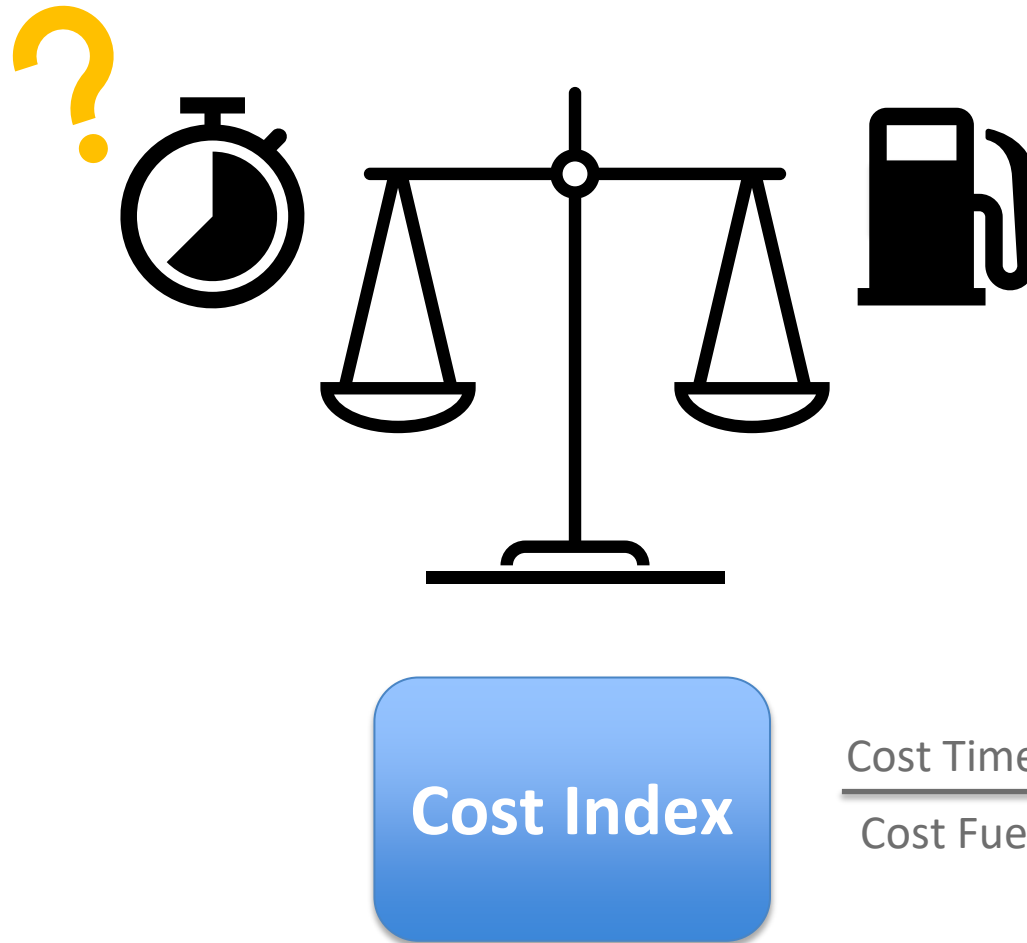




# Pilot3

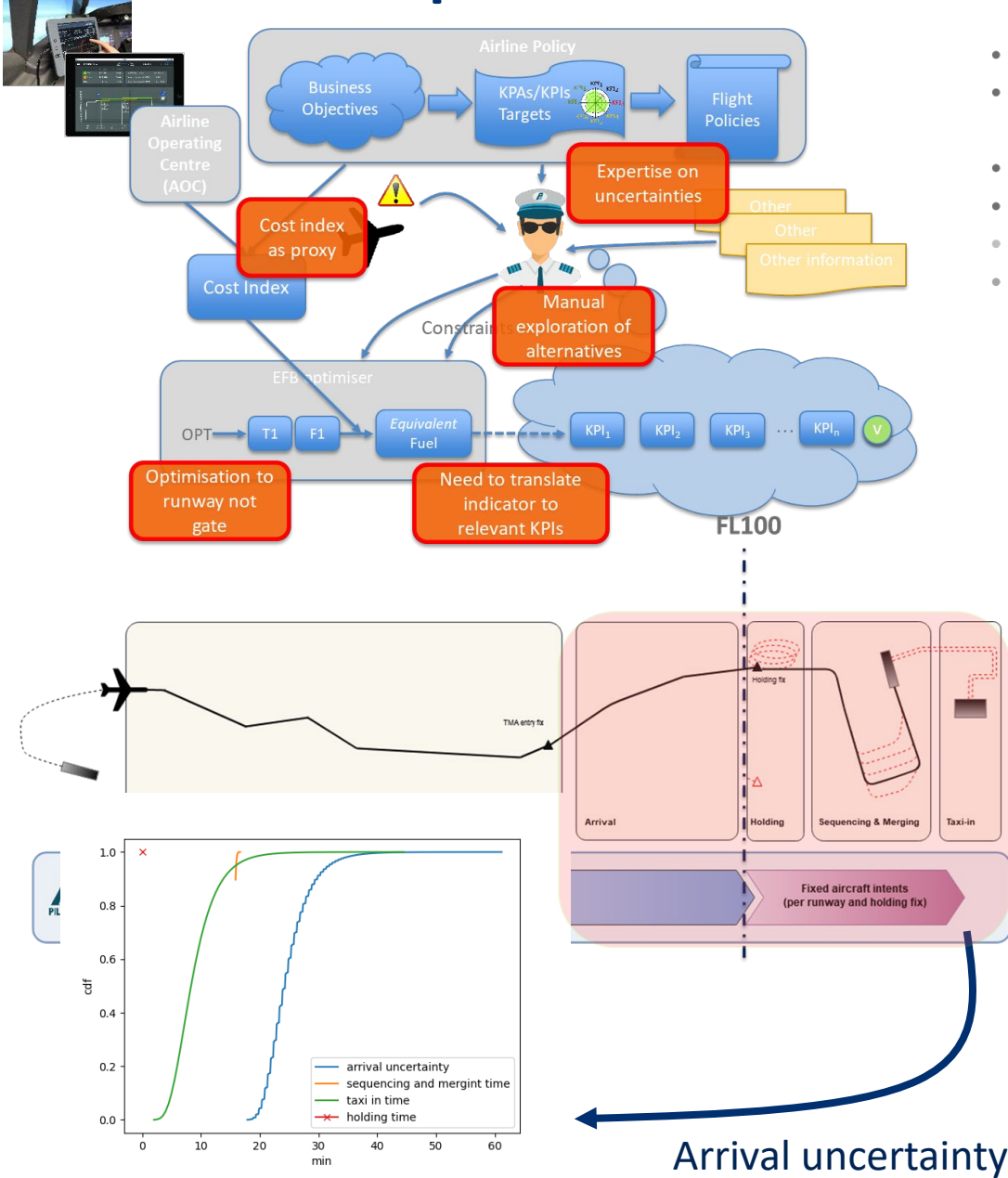


- Optimisation of vertical profile (from triggering point to FL100)

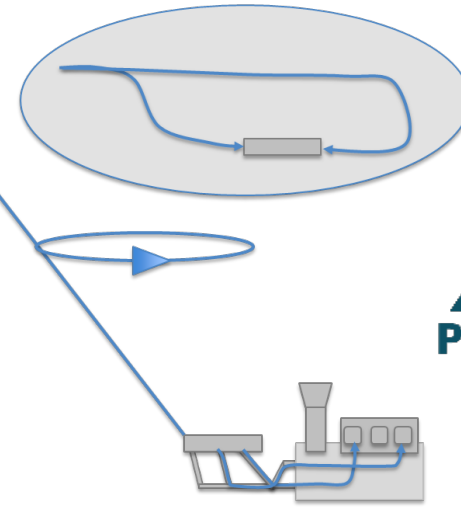




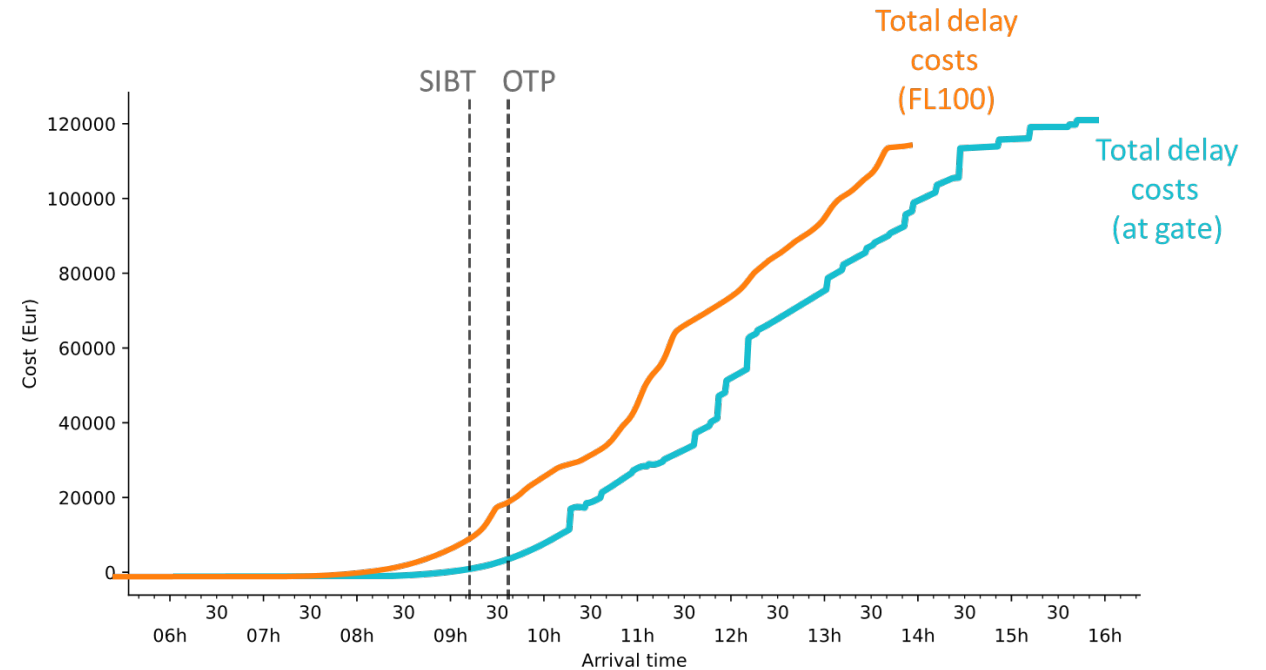
# Pilot3 and Dispatcher3



- Procedure at TMA
- Sequencing and merging distance
- Holding
- Taxi-in time
- Cruise variation
- Wind/Weather uncertainty



Transportation  
Research  
Laboratory

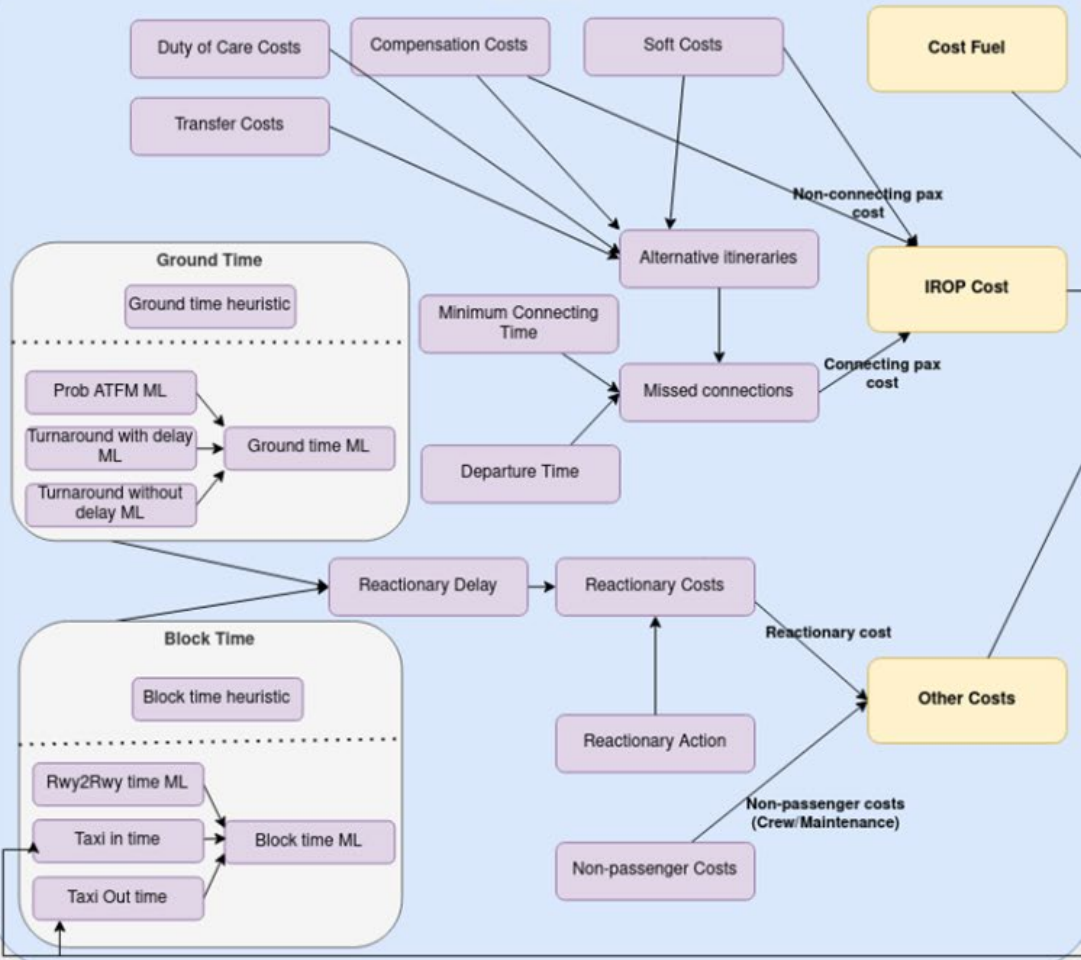




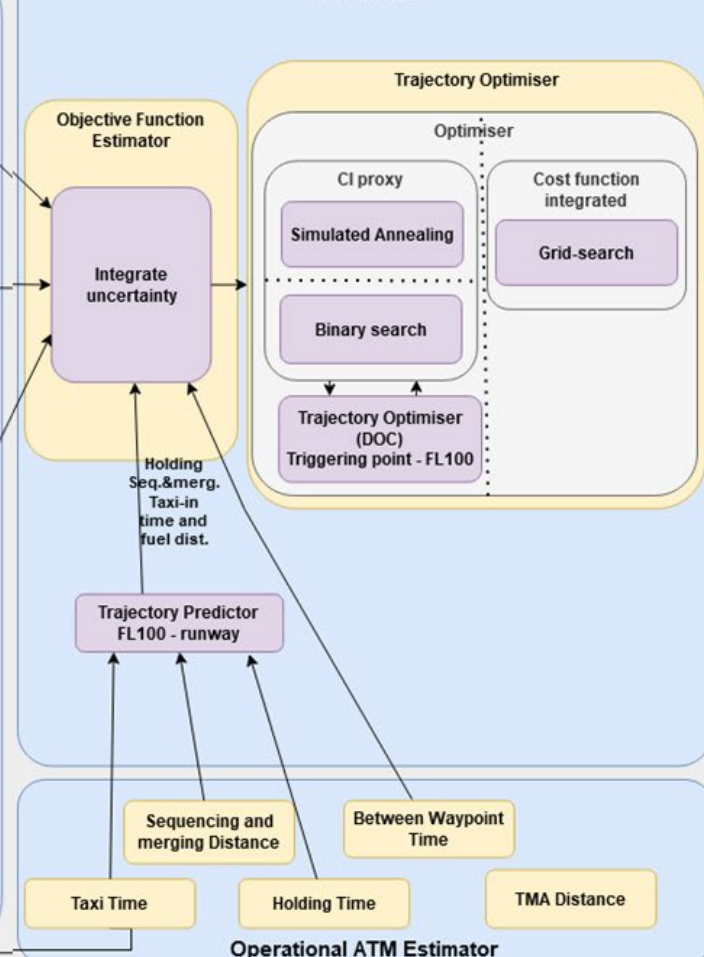
## Multi-objective decision support system

HMI

### Performance Indicators Estimator



### Alternatives Generator



### Performance Assessment Module



Data Manager

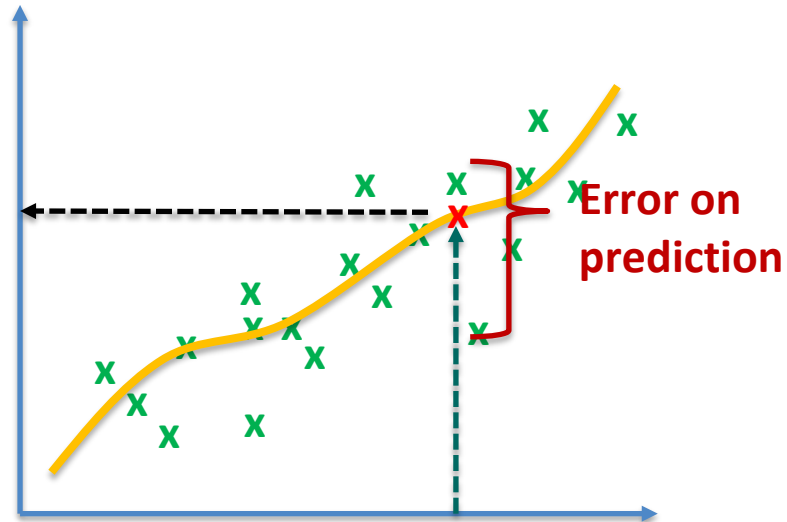
Total delay costs  
(at gate)

Arrival uncertainty

# Pilot3 – Machine learning models challenges

## Challenges

1. Need of distribution not only average expected value

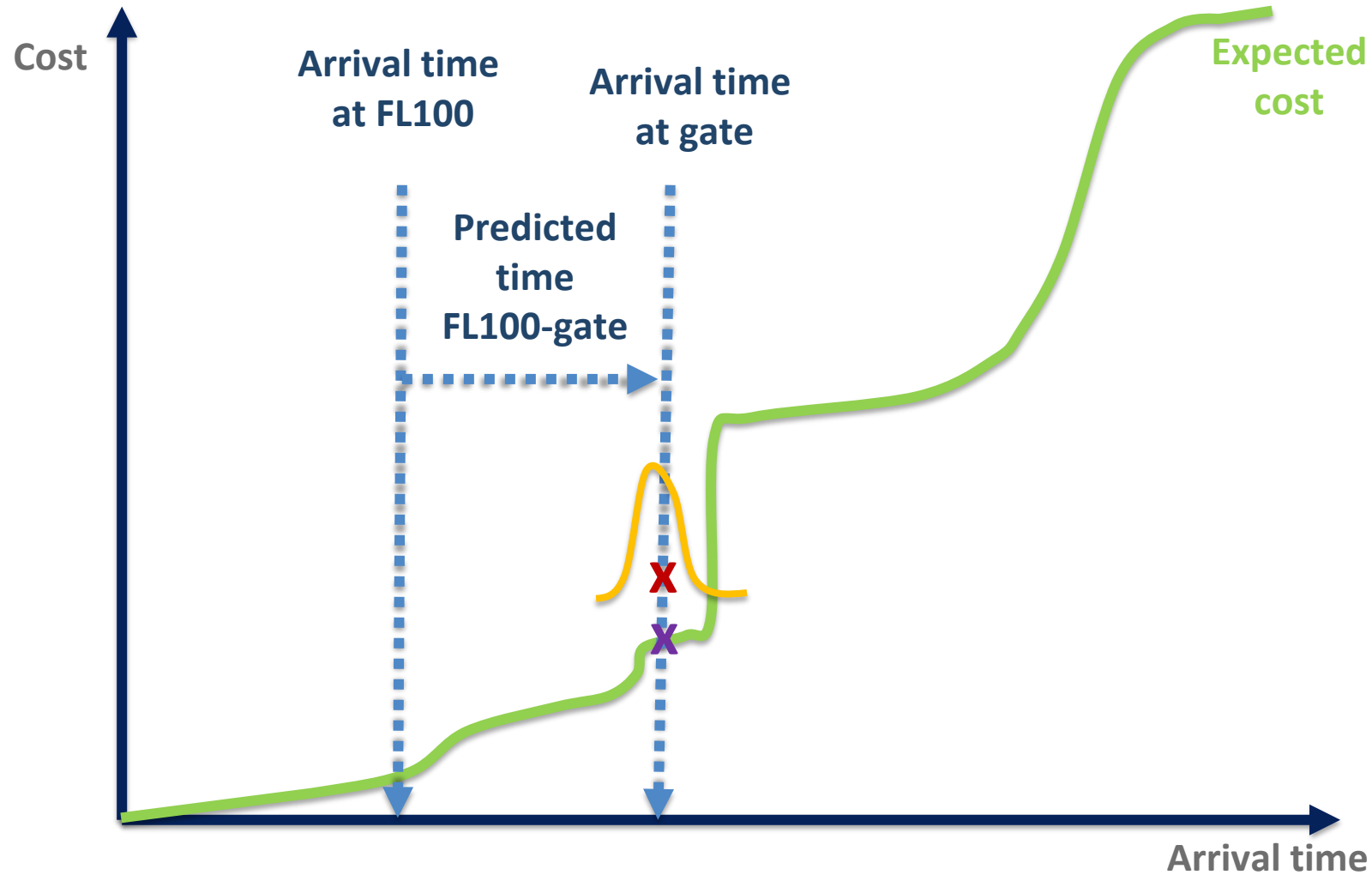




# Pilot3 – Machine learning models challenges

## Challenges

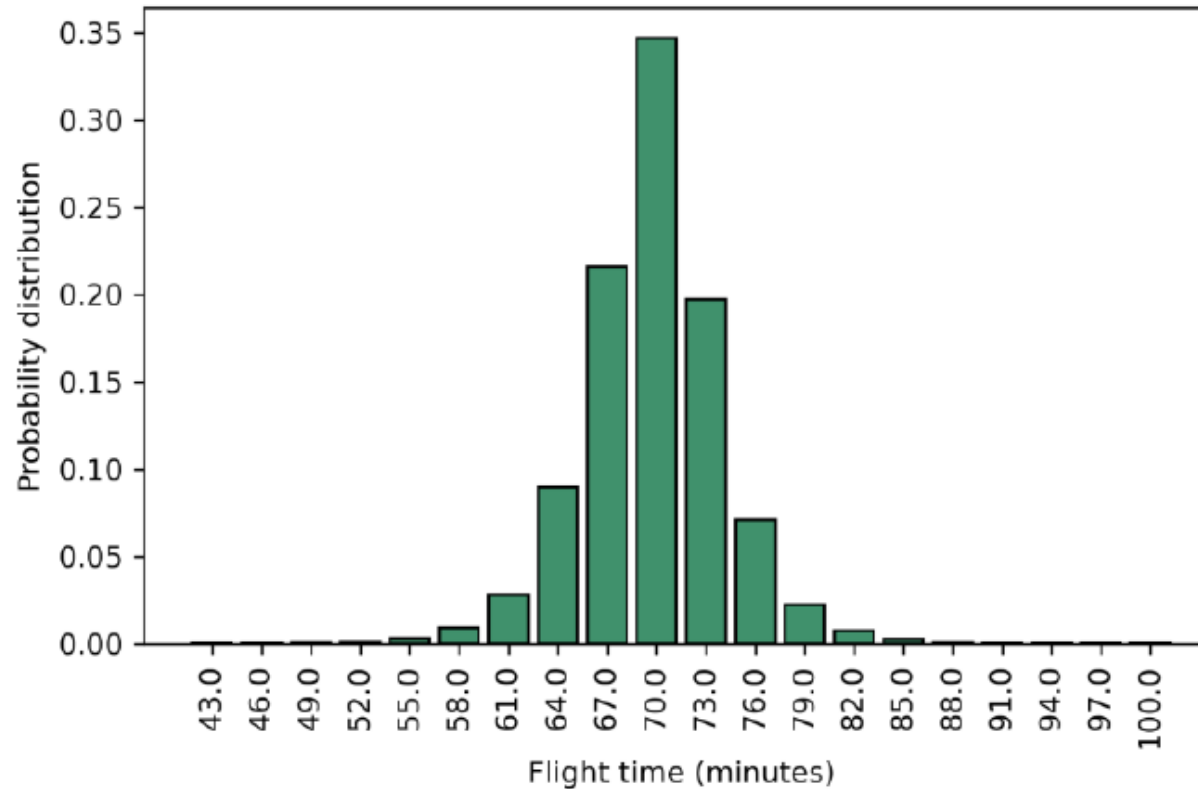
1. Need of distribution not only average expected value



# Pilot3 – Machine learning models challenges

## Challenges

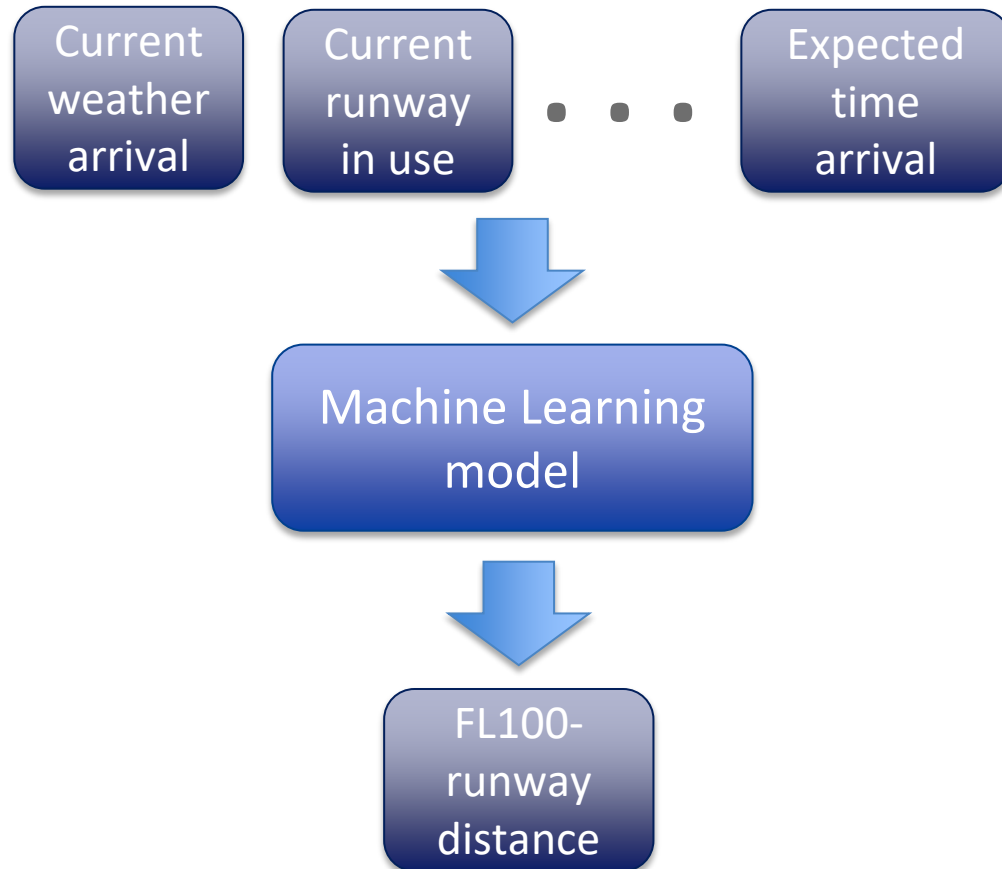
1. Need of distribution not only average expected value



# Pilot3 – Machine learning models challenges

## Challenges

### 2. Prediction horizon



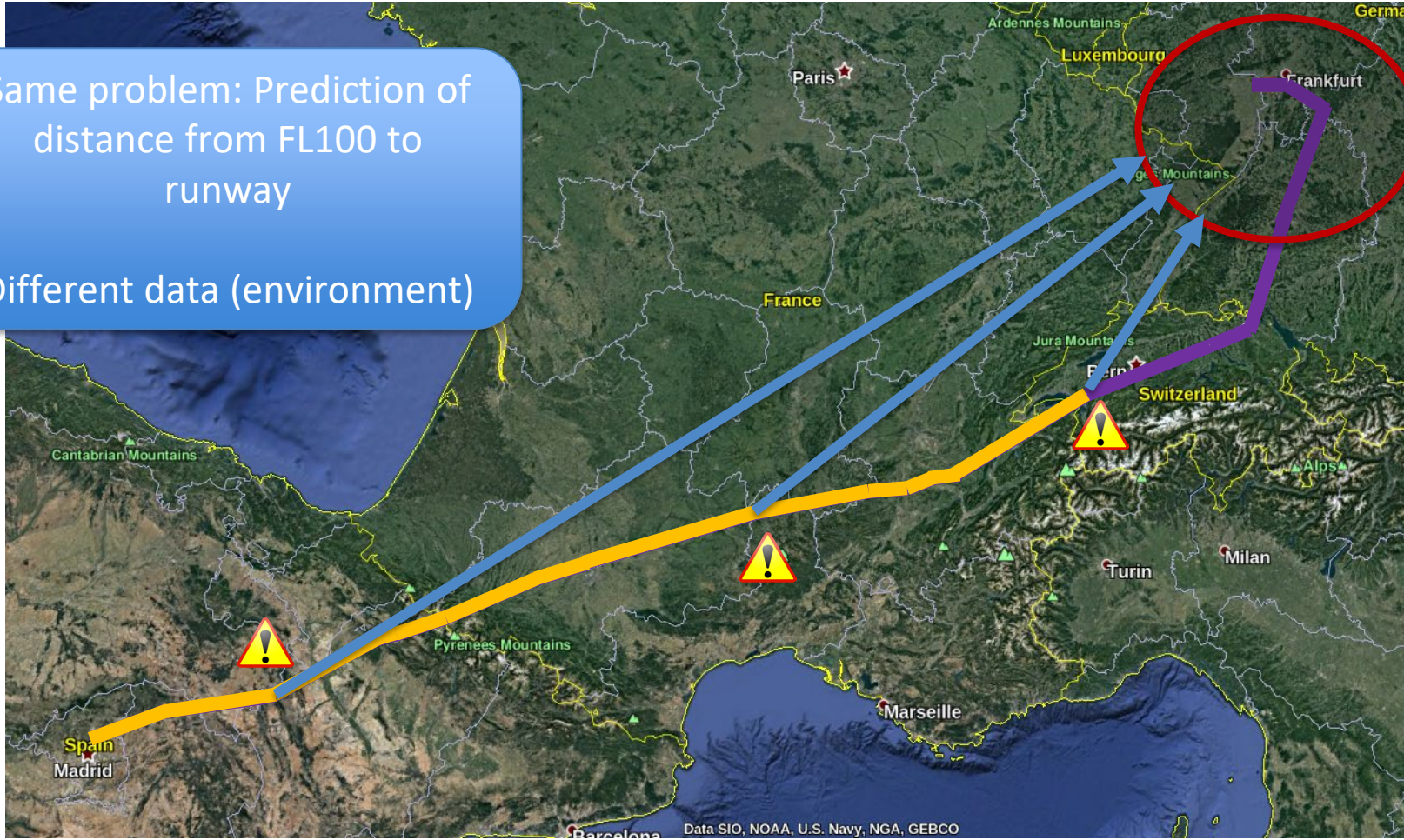
# Pilot3 – Machine learning models challenges

## Challenges

### 2. Prediction horizon

Same problem: Prediction of distance from FL100 to runway

Different data (environment)



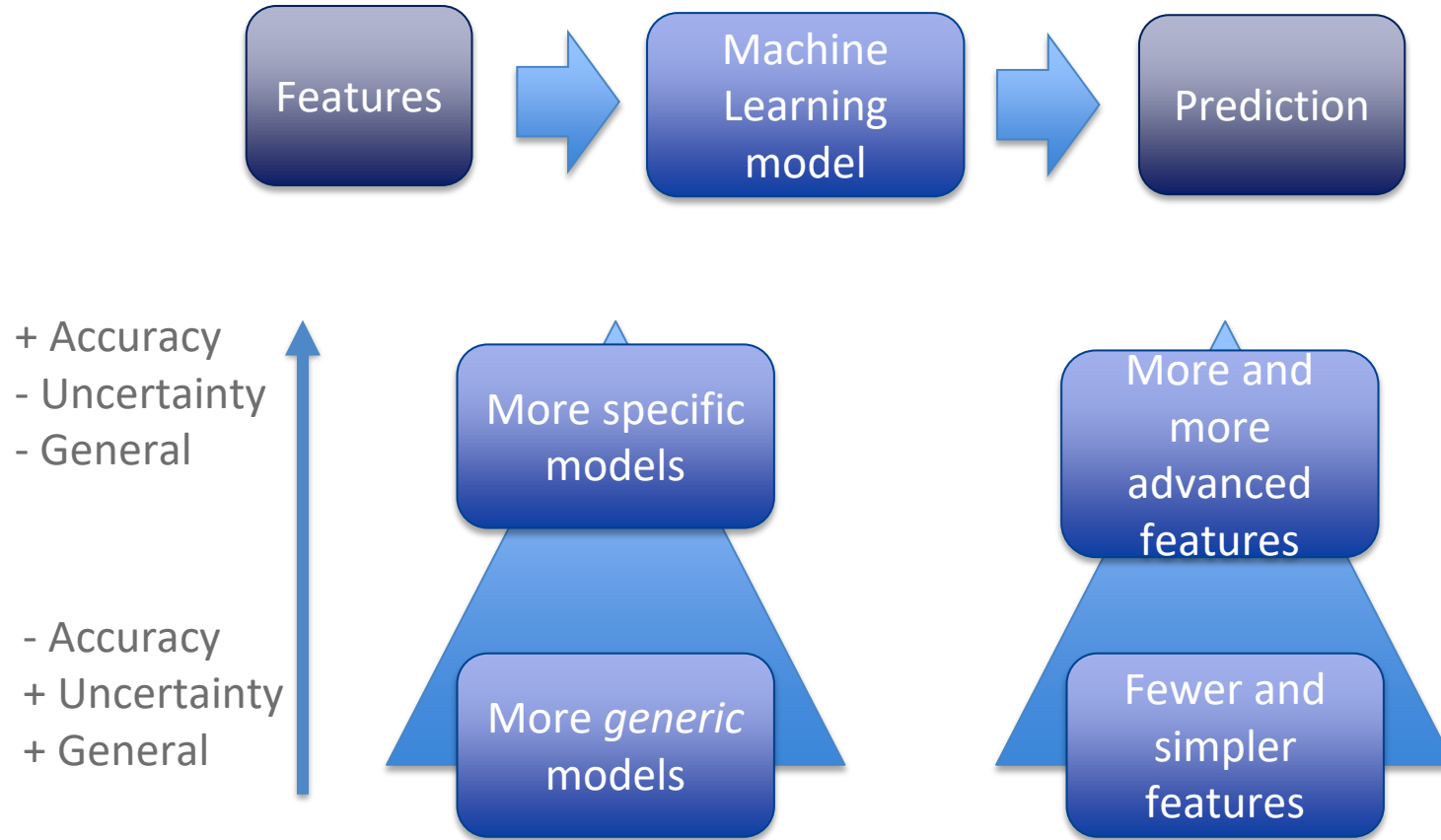
Transportation  
Research  
Laboratory



# Pilot3 – Machine learning models challenges

## Challenges

### 3. Multi-model approach



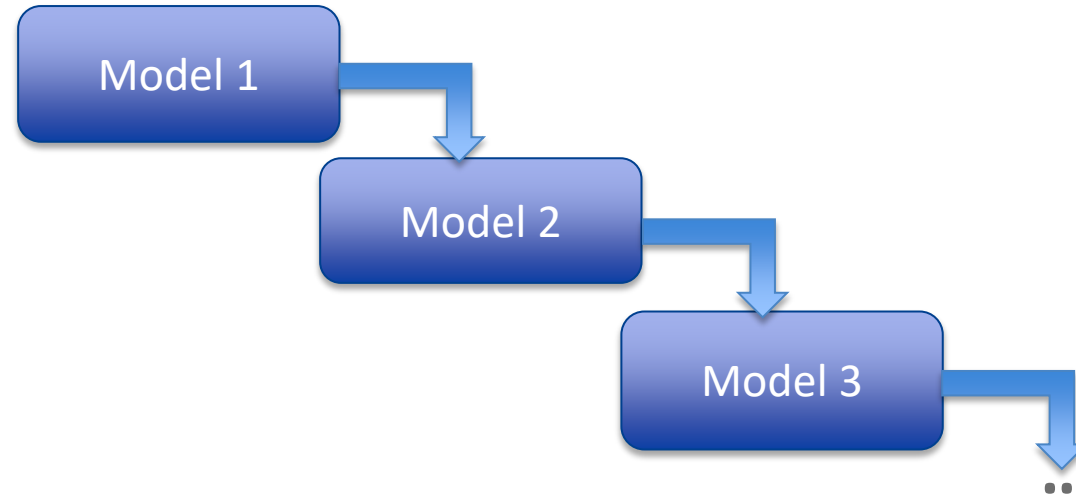


# Pilot3 – Machine learning models challenges

## Challenges

### 3. Multi-model approach

Specific  
More advanced features  
Ground-updated data



Which model to use for  
a given prediction:  
trade-off  
accuracy/uncertainty/  
genericity

Generic  
Fewer features  
Air-available data

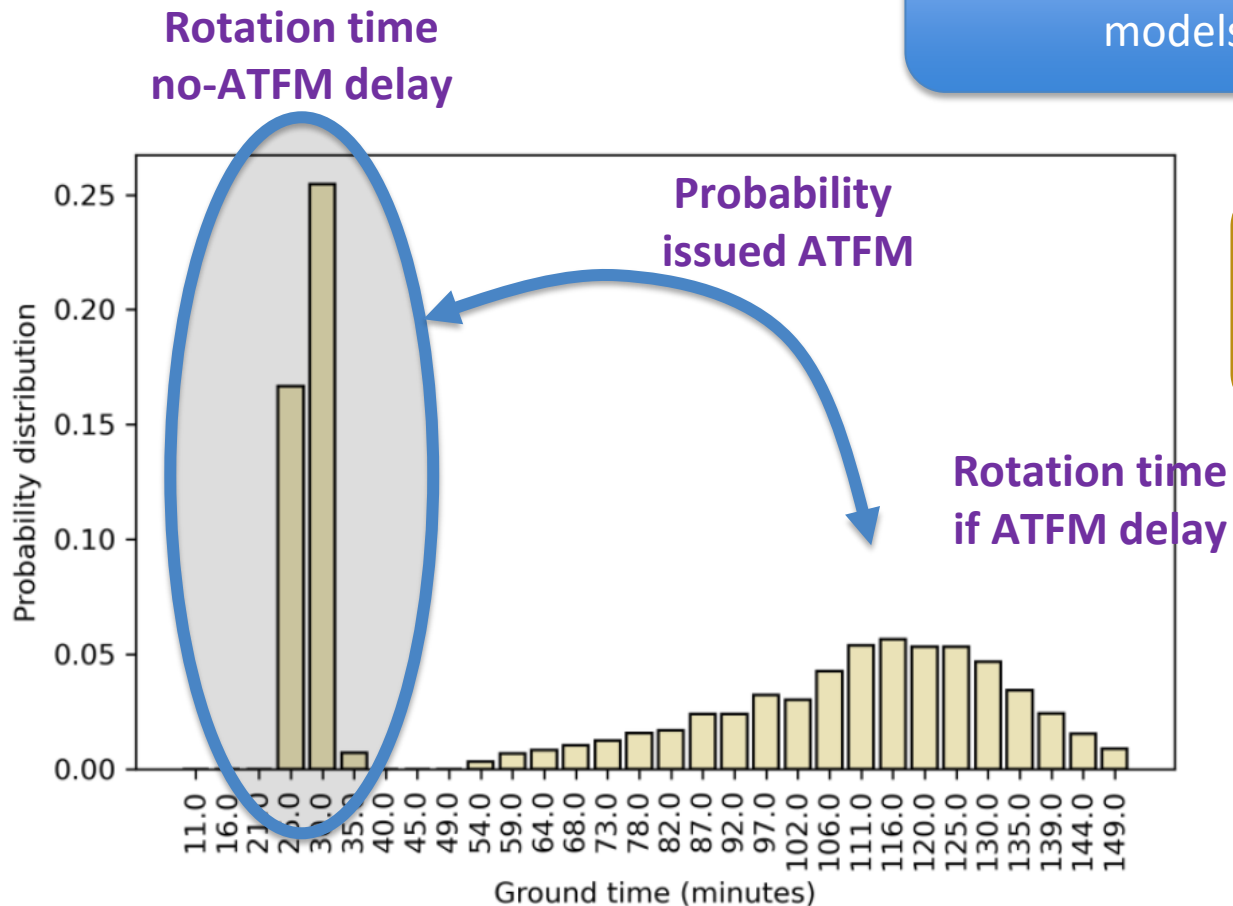
# Pilot3 – Machine learning models challenges

## Challenges

### 3. Multi-model approach

Estimation of complex  
interactions integrating  
models

Propagation  
of errors  
across  
models?

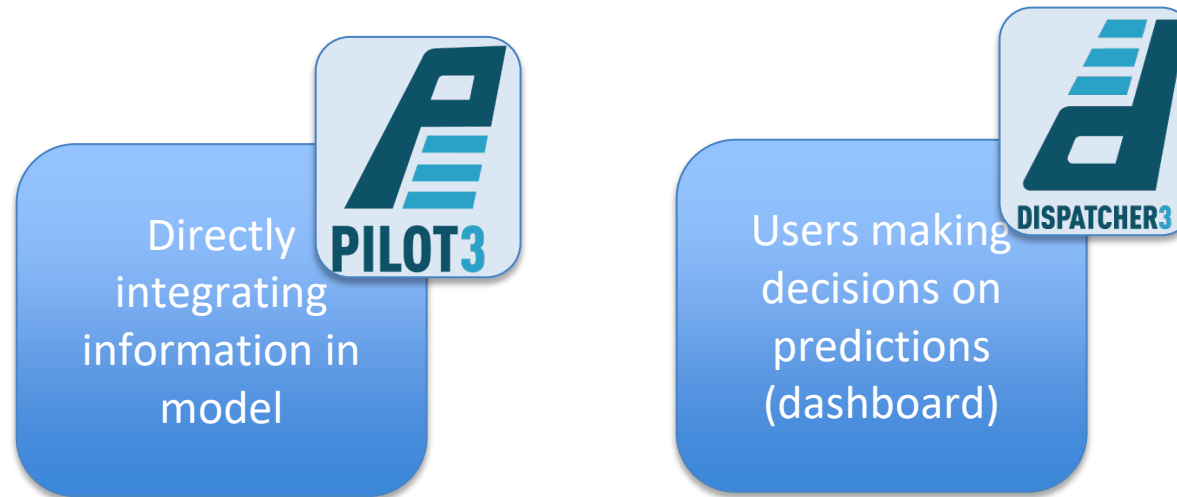


# Pilot3 – Machine learning models challenges

## Challenges

### 4. Visualisation – interpretability

Using predictions with uncertainties → How to present the information to users?



# Pilot3 – Machine learning models challenges

## Challenges

### 5. Data!

Transportation  
Research  
Laboratory



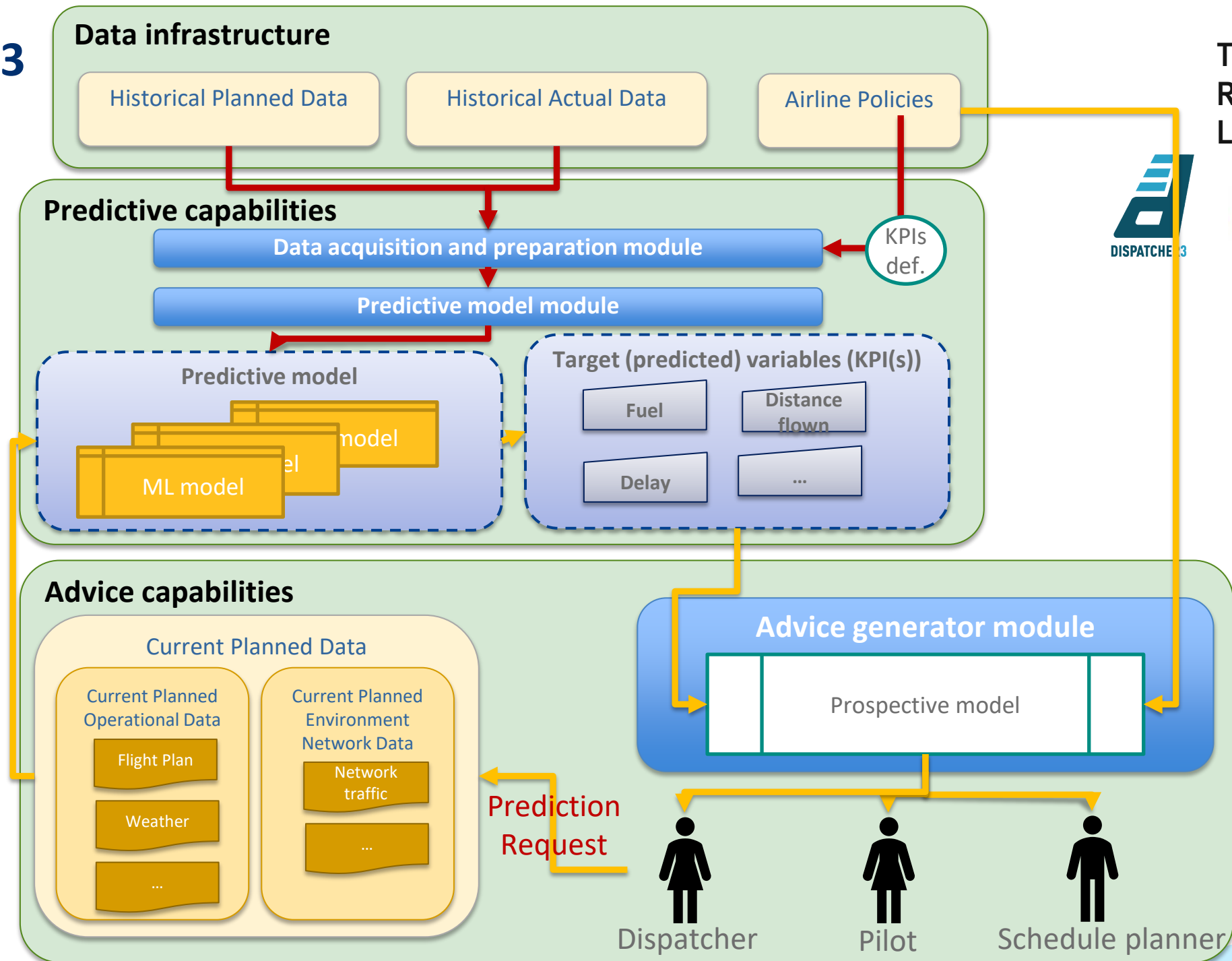
# Pilot3 – Machine learning models challenges



- **Machine learning models in Pilot3**
  - Used to compute cost function components
  - Used to estimate uncertainties that affect operations:
    - Distance at arrival (FL100-runway)
    - Taxi-in time
    - Reactionary delay
      - Block time
      - Rotation time
      - ATFM delay
- **Challenges (most of them applicable to ATM in general)**
  1. Need of distribution not only average expected value
  2. Prediction-horizon
  3. Multi-model approach
  4. Visualisation – interpretability
  5. Data!



# Dispatcher3

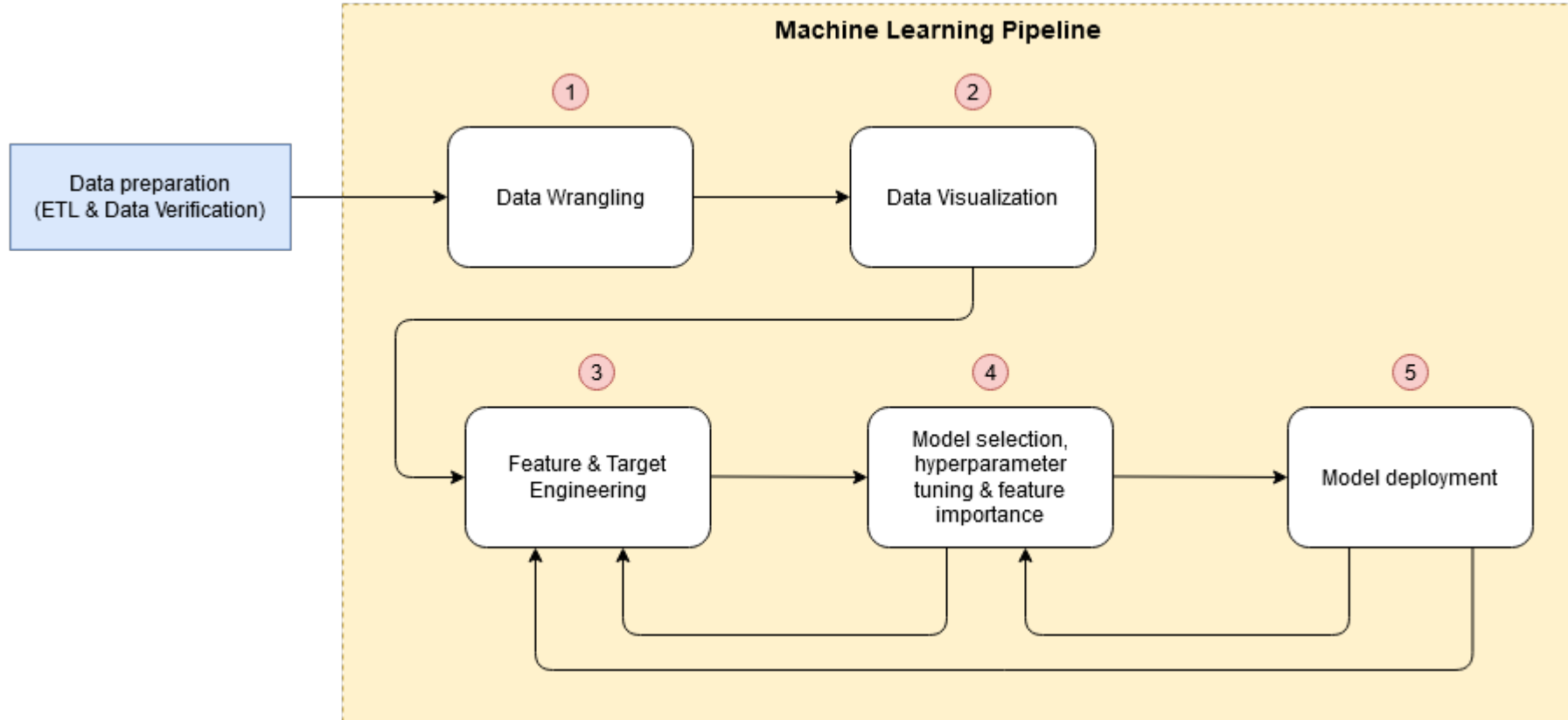


# Dispatcher3 – Data sources



| # | Data source                          | Description  | Used in  |
|---|--------------------------------------|--|--|
| 1 | FDM provided by Vueling              | Contains performance data and static data directly collected by the airlines                                       | Fuel deviation (pre-departure)                   |
| 2 | ADS-B data by OpenSky                | Radar data, contains trajectories  | Holding scenarios (pre-departure)                |
| 3 | METAR                                | Contains weather information at different European Airports  | Pre-departure and pre-tactical scenarios         |
| 4 | NOAA weather forecasts               | GRIB file containing meteorological information in a 4D space (latitude, longitude, altitude and time)             | Fuel deviation and runway in use (pre-departure) |
| 5 | ALLFT+                               | Network and flight plan and trajectories information from DDR2 by EUROCONTROL                                      | Pre-departure and pre-tactical scenarios         |
| 6 | ECTL R&D Archive                     | Network information and flight information provided by EUROCONTROL   | Pre-departure scenarios                          |
| 7 | Vueling flight data and flight plans | Flight information and flight plans from Vueling flights.  | Pre-departure and pre-tactical scenarios         |
| 8 | Flight Plans by PREDICT              | Simulated data using PREDICT software which estimates pre-tactical flight plans based on historical data (ALLFT+ ) | Pre-tactical scenarios                           |

# Dispatcher3





## Predictive capabilities

### Pre-departure models (using data available 3h prior SOBT)

- Holdings at arrival at EGLL
- Fuel usage
- Runway at arrival at LEIB

### Planned models (using data available at D-1)

- Probability being regulated by ATFM
- Location of regulation if flight regulated (airspace/airport)
- Probability regulated flight has zero minutes of delay assigned
- Distribution of delay if delay assigned is greater than zero

## Advice capabilities

- Rotation / reactionary delay propagation
- Integrated interactive flight plan visualisation

## Static vs dynamic features

### Static features

| Operational time       |
|------------------------|
| Hour of departure      |
| Day of the week        |
| Month                  |
| Airport information    |
| Size departure airport |
| Size arrival airport   |
| Departure hub airline  |
| Arrival hub airline    |

### Dynamic features

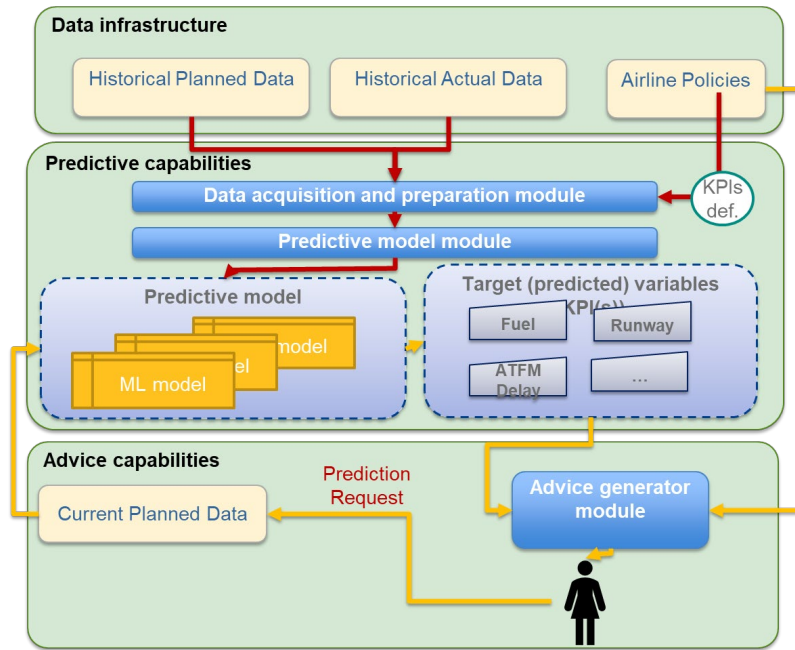
| Airport demand             |
|----------------------------|
| Normalized num. departures |
| Normalized num. arrival    |
| Network demand             |
| Normalized OC              |
| Normalized EC              |
| Weather                    |
| ATMAP score                |
| Temperature                |
| Wind speed                 |
| Visibility                 |
| u/v wind                   |
| Geopotential               |
| Relative humidity          |
| Ventilation rate           |

METAR

NOAA



# Pilot3 and Dispatcher3



## Predictive capabilities

### Pre-departure models (using data available 3h prior SOBT)

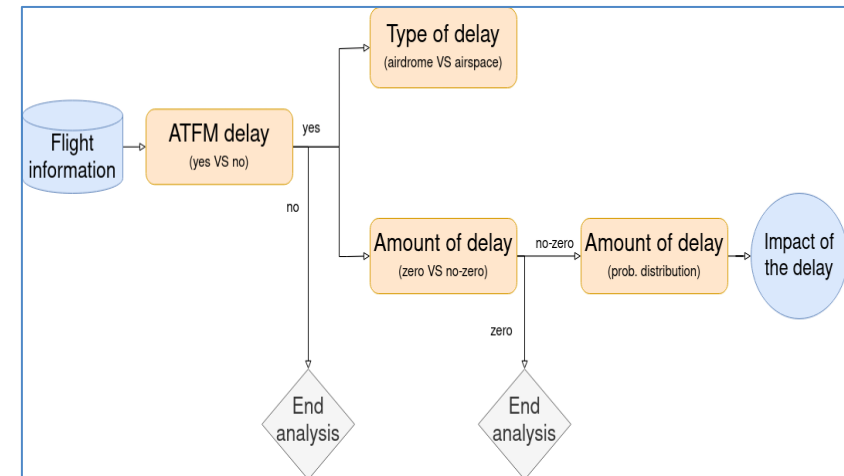
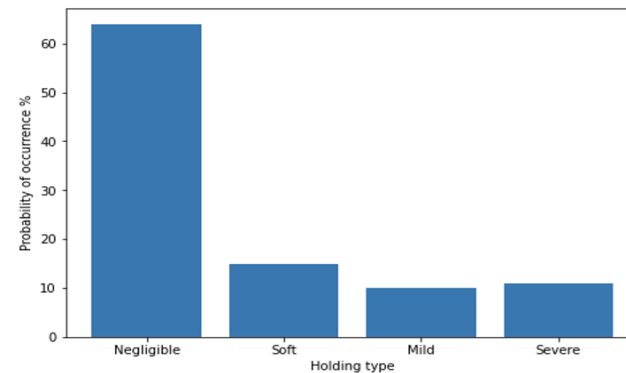
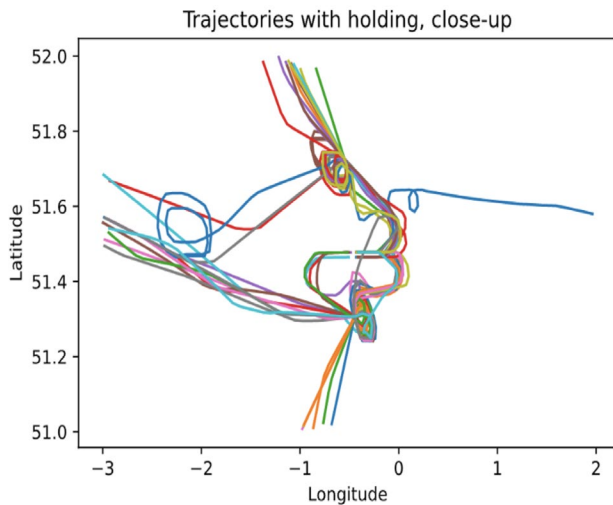
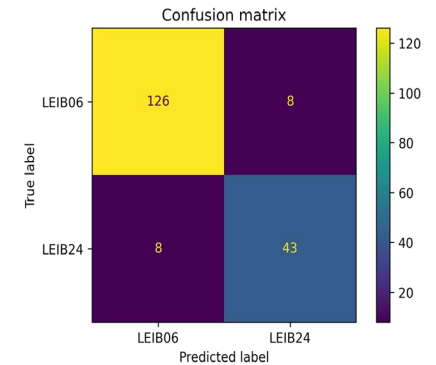
- Holdings at arrival at EGLL
- Fuel usage
- Runway at arrival at LEIB

### Planned models (using data available at D-1)

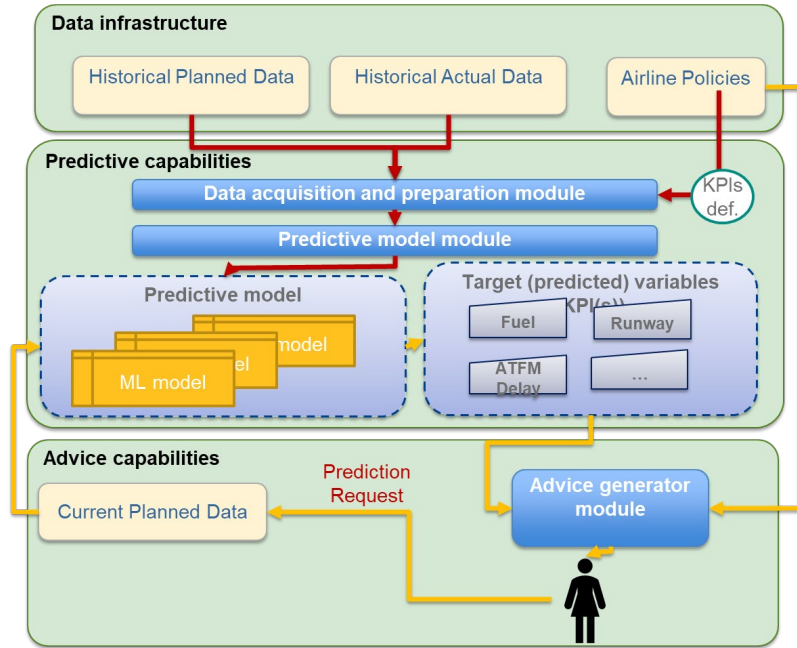
- Probability being regulated by ATFM
- Location of regulation if flight regulated (airspace/airport)
- Probability regulated flight has zero minutes of delay assigned
- Distribution of delay if delay assigned is greater than zero

## Advice capabilities

### Rotation / reactionary delay propagation Integrated interactive flight plan visualisation



# Pilot3 and Dispatcher3

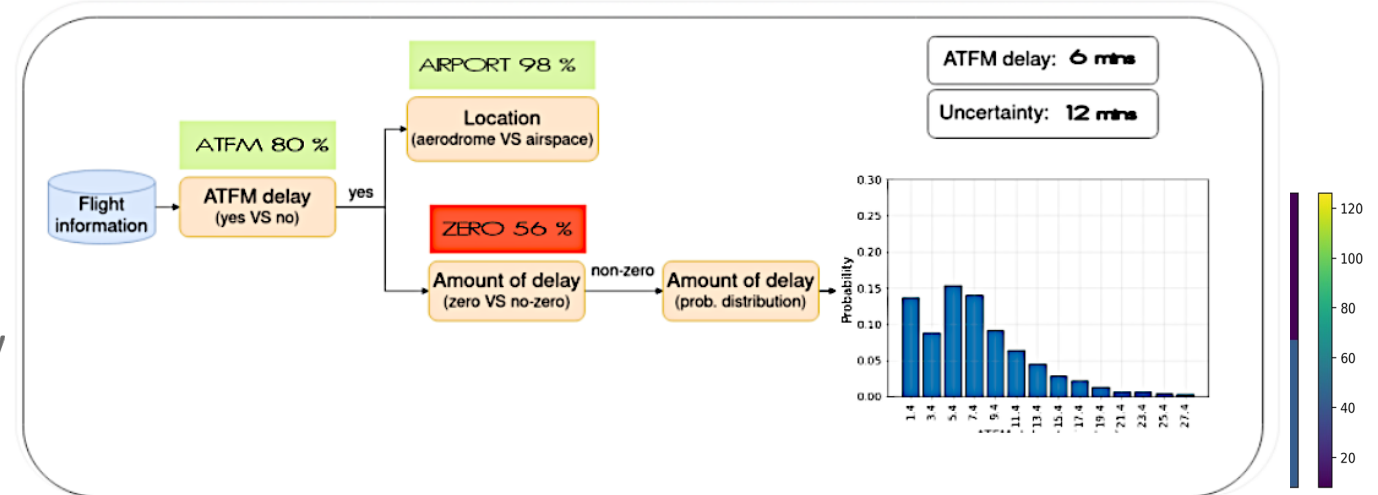


## Predictive capabilities

### Pre-departure models (using data available 3h prior SOBT)

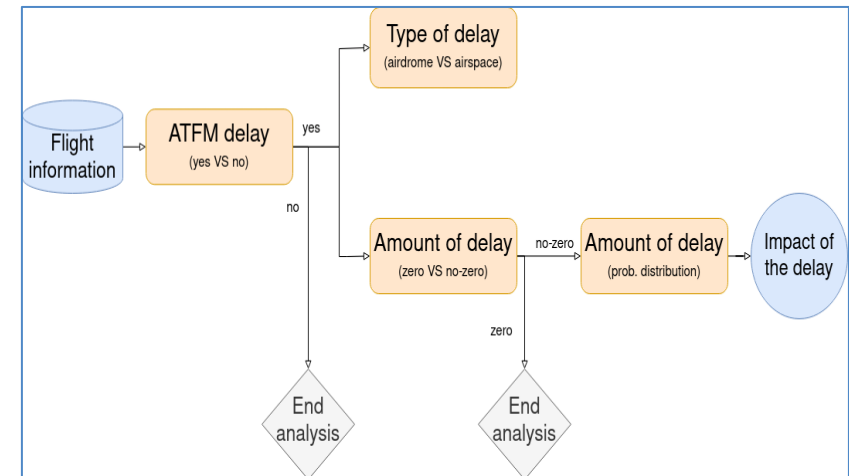
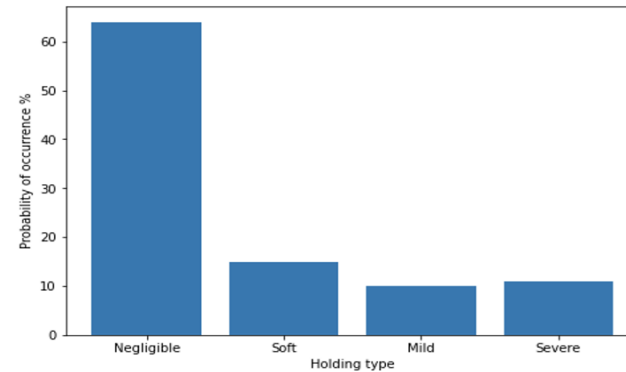
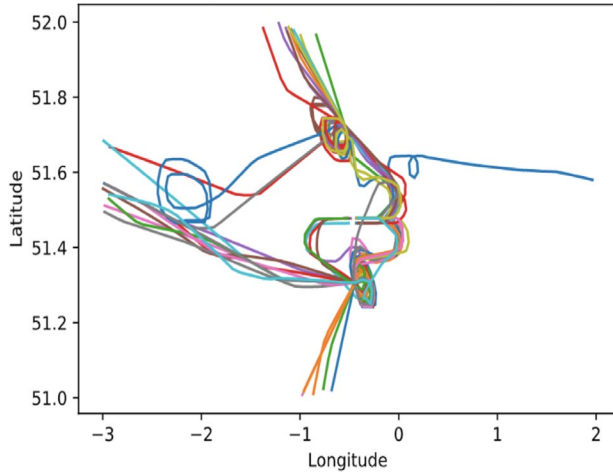
- Holdings at arrival at EGLL
- Fuel usage

Adv

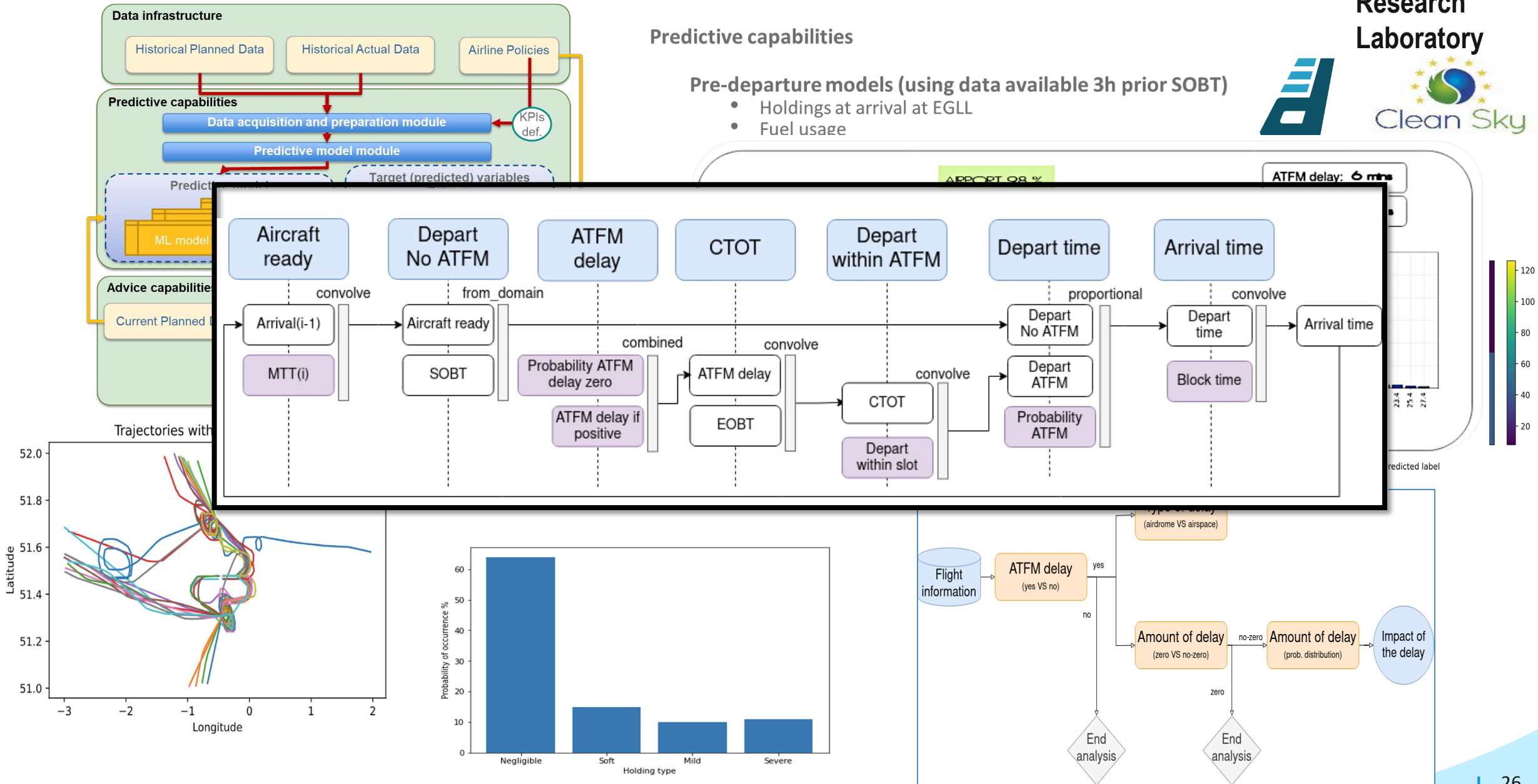


Predicted label

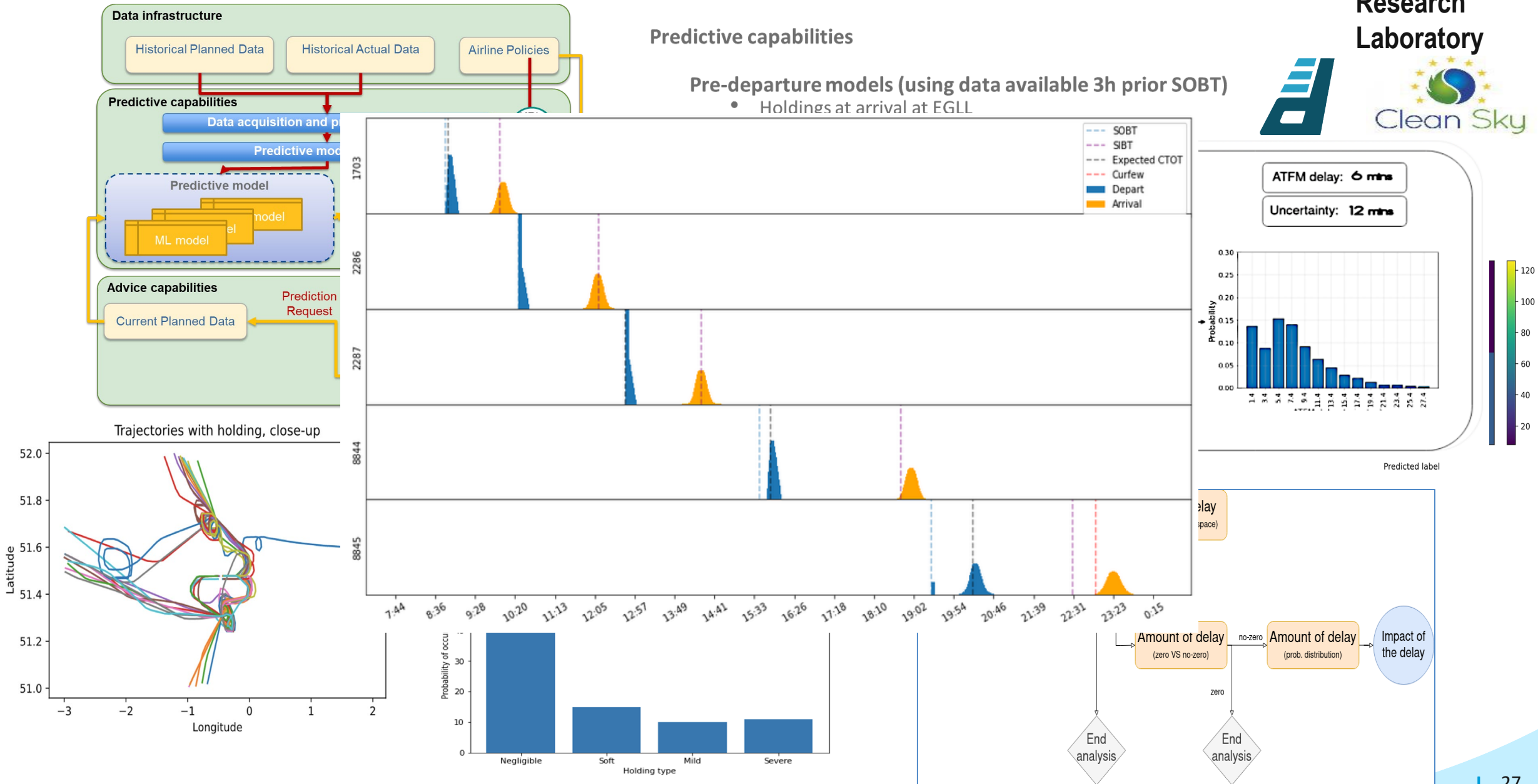
Trajectories with holding, close-up



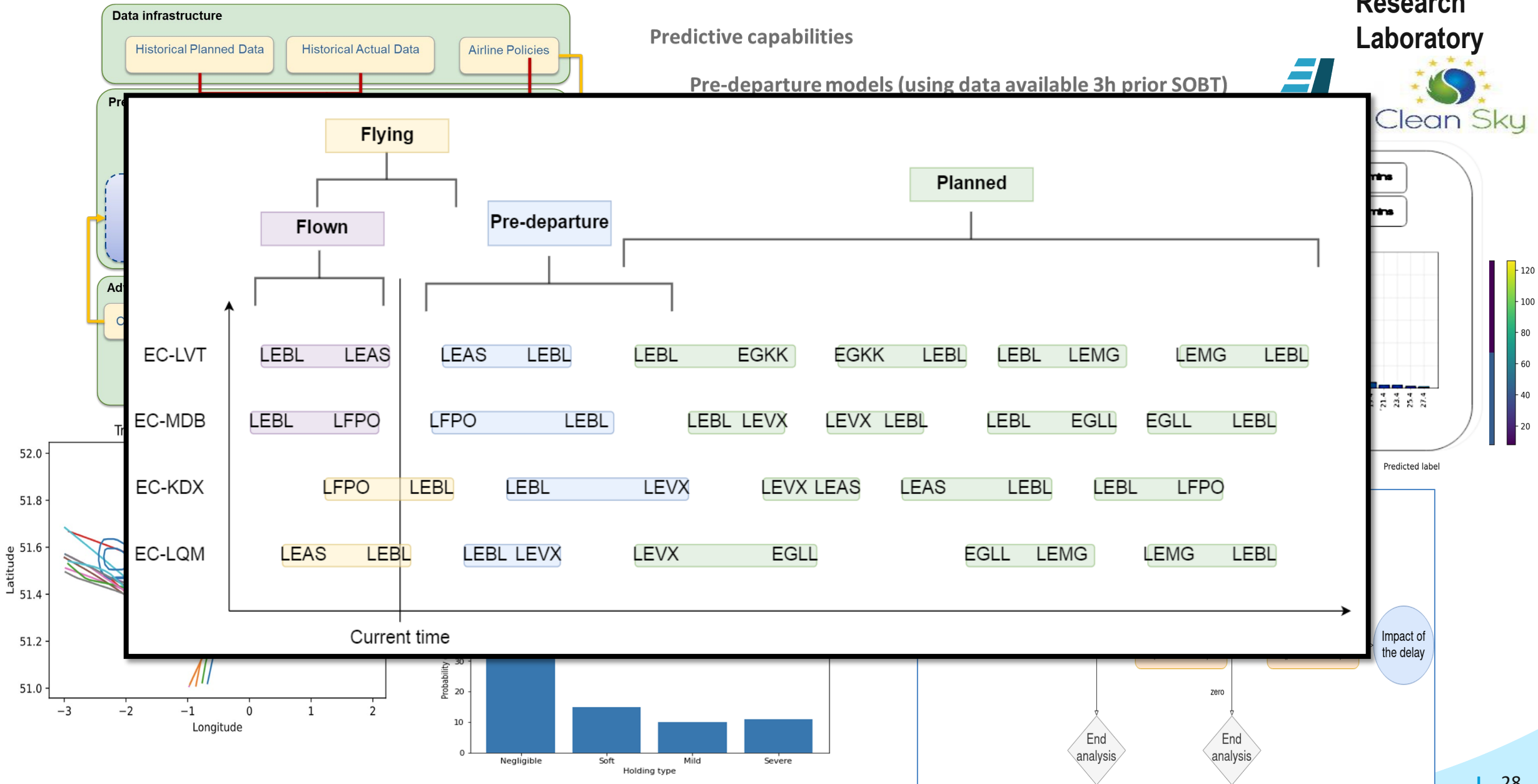
# Pilot3 and Dispatcher3



# Pilot3 and Dispatcher3



# Pilot3 and Dispatcher3





# Conclusions

**Models are useful but not on their own**

**Consider data availability**

**Interpretability**

Transportation  
Research  
Laboratory



**Autum  
School  
2025**

**Thank you!**

Transportation  
Research  
Laboratory

Quintin Hogg  
Trust