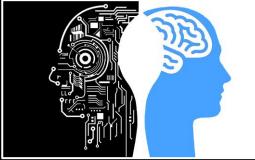


ARTIMATION Transparent ARTIficial Intelligence and AutoMATION to Air Traffic Management Systems



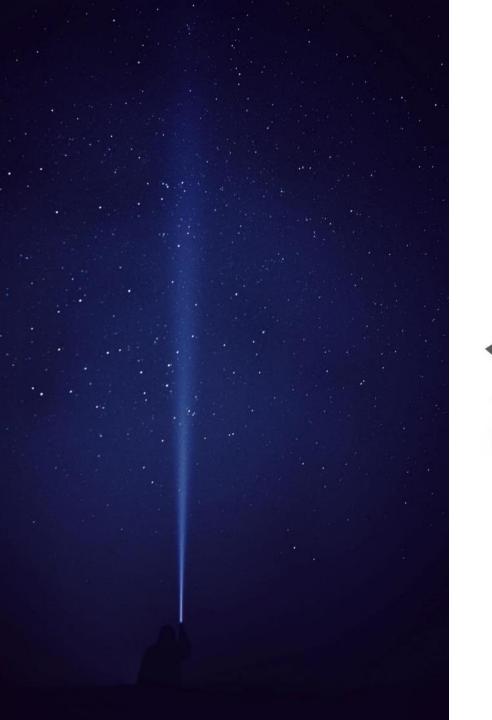
4<sup>th</sup> November 2022

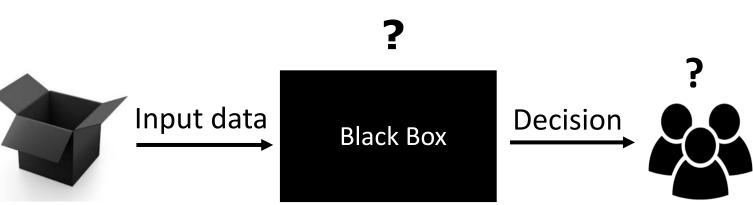




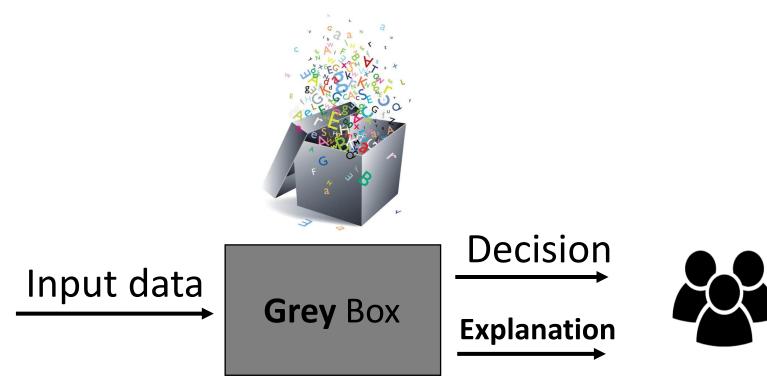
# Toward

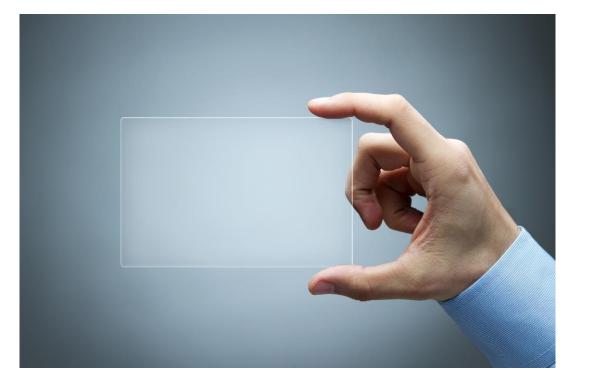
# Automatization Paradigm ?









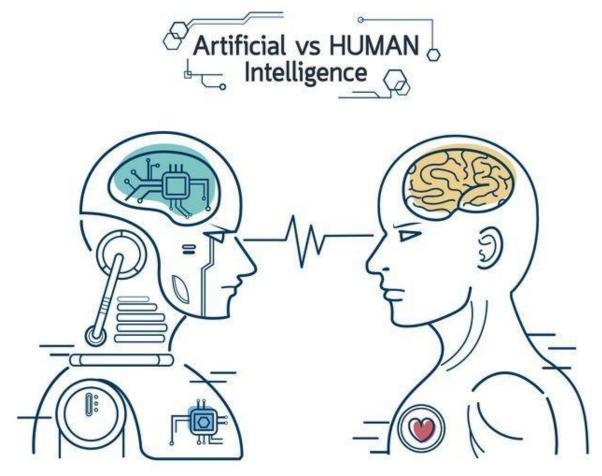


# Sometimes, the rationale behind the decision is more important than the decision itself

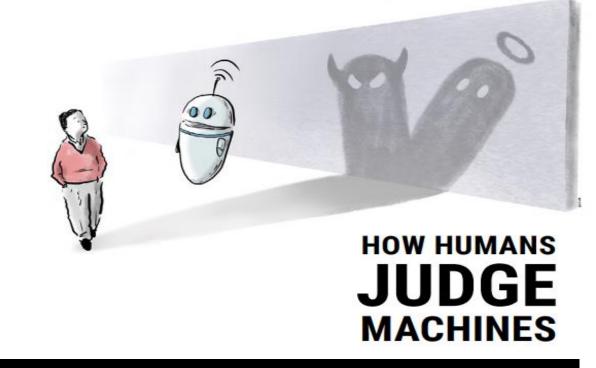


### How to open black boxes?

# The key to supporting this task is not only to visualize data, but also to allow users to interact with it

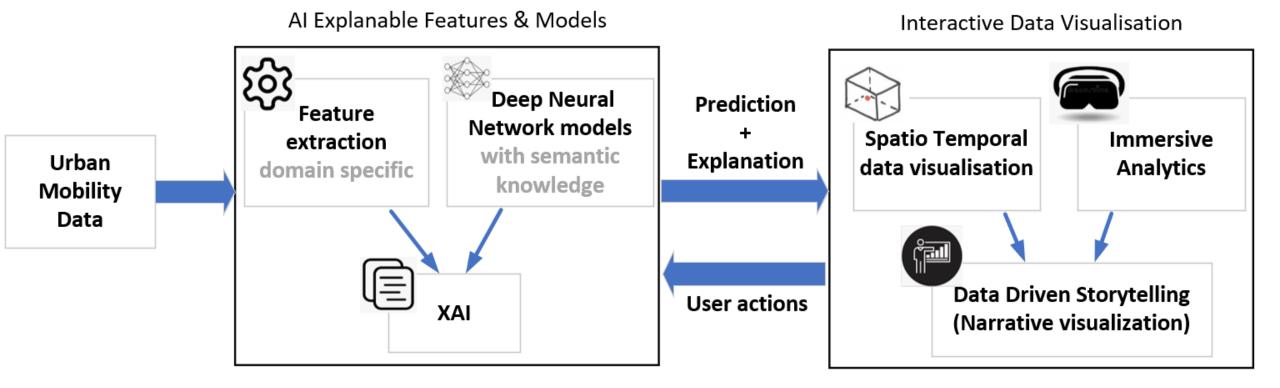


https://www.judgingmachines.com/



Cesar A. Hidalgo, Diana Orghian, Jordi Albo Canals, Filipa De Almeida. How Humans Judge Machines Relié – 2 février 2021.The MIT Press

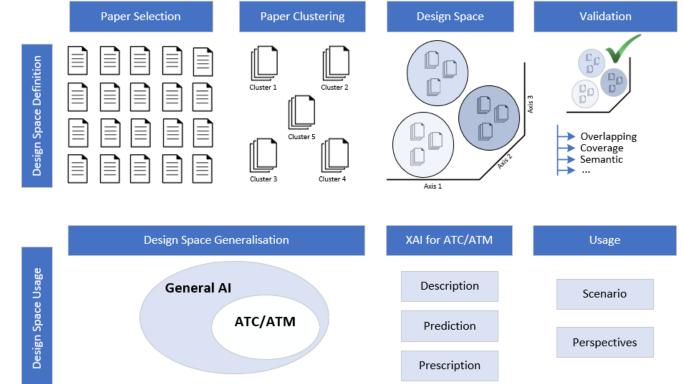


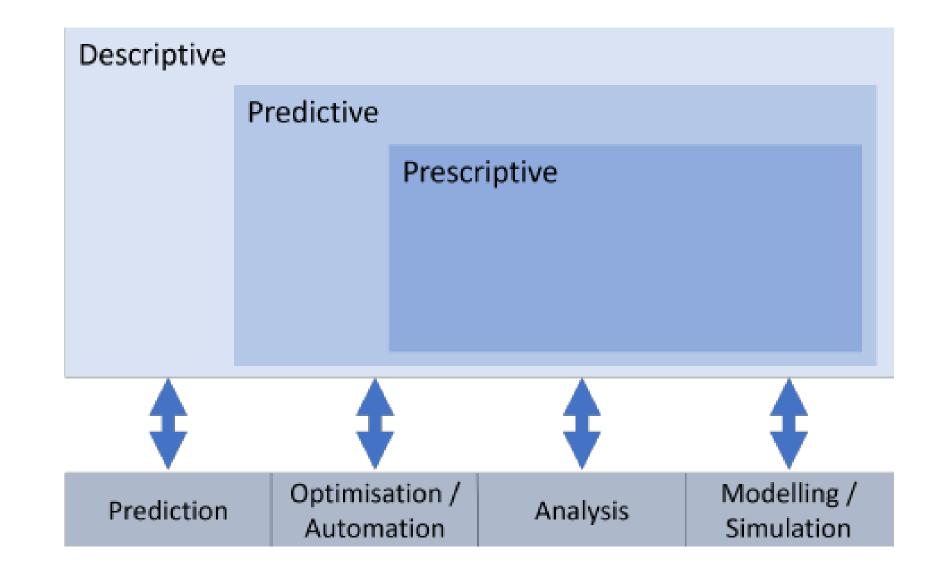


Systematic Review

### A Survey on Artificial Intelligence (AI) and eXplainable AI in Air Traffic Management: Current Trends and Development with Future Research Trajectory

Augustin Degas <sup>1,\*</sup>, Mir Riyanul Islam <sup>2,\*,</sup>, Christophe Hurter <sup>1,</sup>, Shaibal Barua <sup>2,</sup>, Hamidur Rahman <sup>2,</sup>, Minesh Poudel <sup>1,</sup>, Daniele Ruscio <sup>3,</sup>, Mobyen Uddin Ahmed <sup>2,</sup>, Shahina Begum <sup>2,</sup>, Md Aquif Rahman <sup>2,</sup>, Stefano Bonelli <sup>3,</sup>, Giulia Cartocci <sup>4,</sup>, Gianluca Di Flumeri <sup>4,</sup>, Gianluca Borghini <sup>4,</sup>, Fabio Babiloni <sup>4,</sup> and Pietro Aricó <sup>4,</sup>





Degas, A.; Islam, M.R.; Hurter, C.; Barua, S.; Rahman, H.; Poudel, M.; Ruscio, D.; Ahmed, M.U.; Begum, S.; Rahman, M.A.; Bonelli, S.; Cartocci, G.; Di Flumeri, G.; Borghini, G.; Babiloni, F.; Aricó, P. A Survey on Artificial Intelligence (AI) and eXplainable AI in Air Traffic Management: Current Trends and Development with Future Research Trajectory. *Appl. Sci.* 2022, *12*, 1295. https://doi.org/10.3390/app12031295



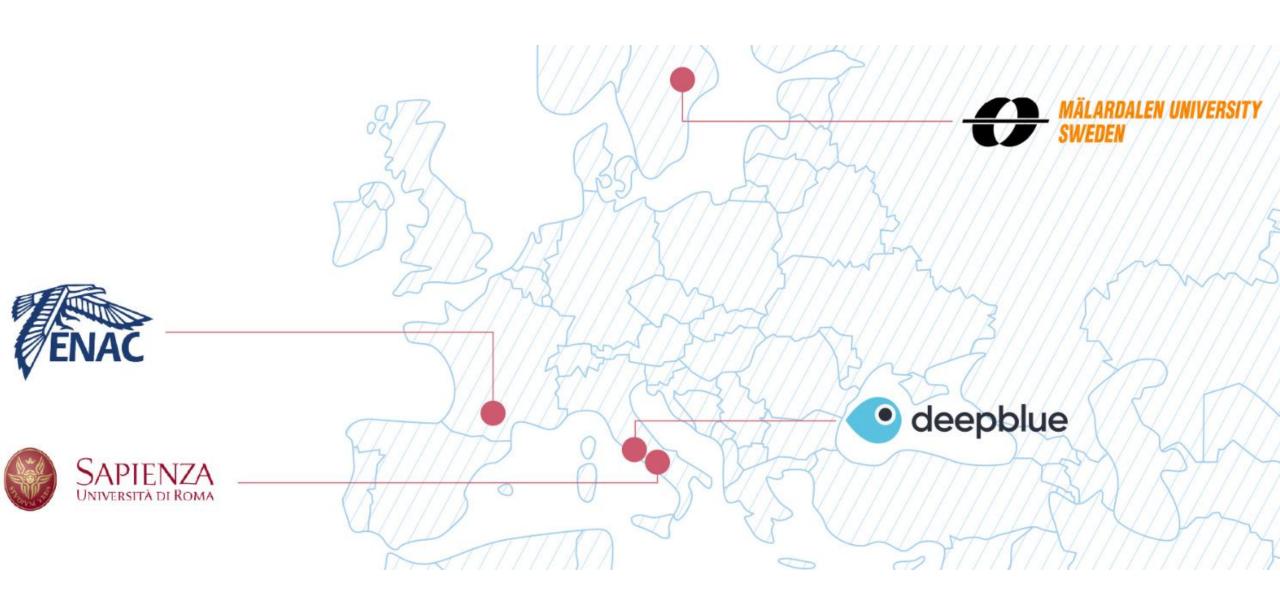




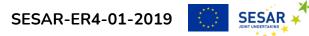
https://www.artimation.eu/

Transparent Artificial intelligence and Automation to Air Traffic Management

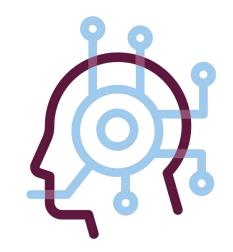
**Systems** 







# Conflict Detection and Resolution visualisation



 Assumés

 RAMS61
 IL76
 340
 1340

 BAW510
 E145
 360
 1560

 AZA564
 A320
 360
 1560

 RVR1784
 A319
 390
 1590

 BER531
 B734
 340
 1540

 AFR448
 B736
 370
 1370

 RVR464
 B736
 350
 1540

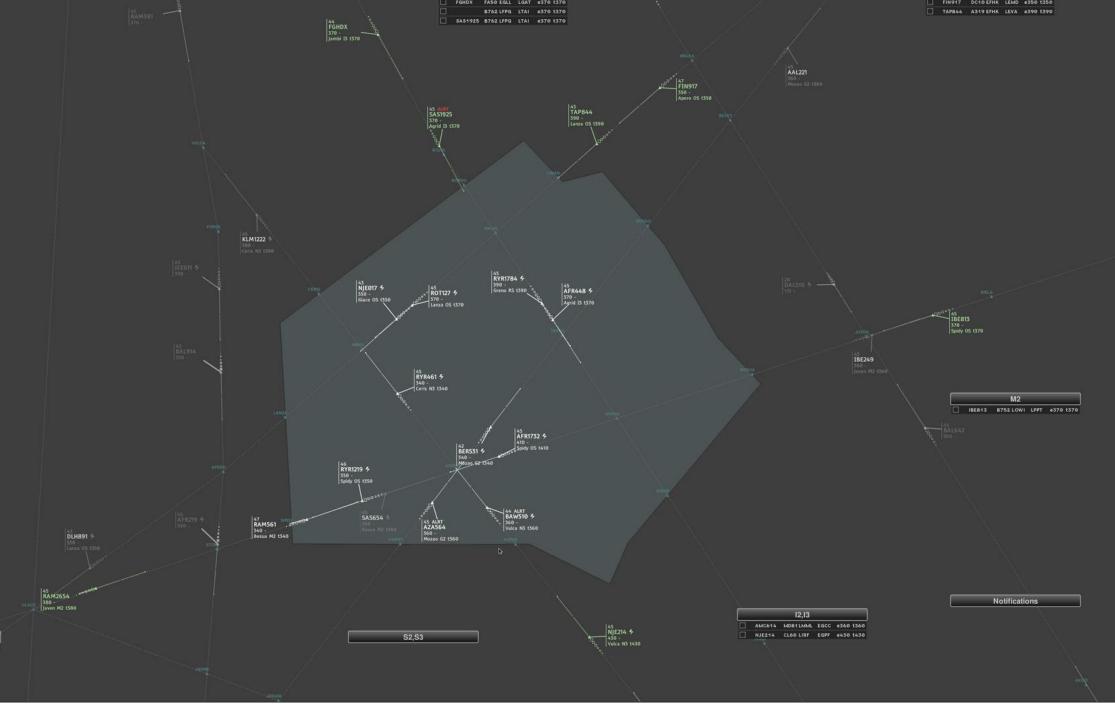
 AFR448
 B736
 350
 1570

 RVF464
 B736
 350
 1530

 AFR175
 MD88
 350
 1530

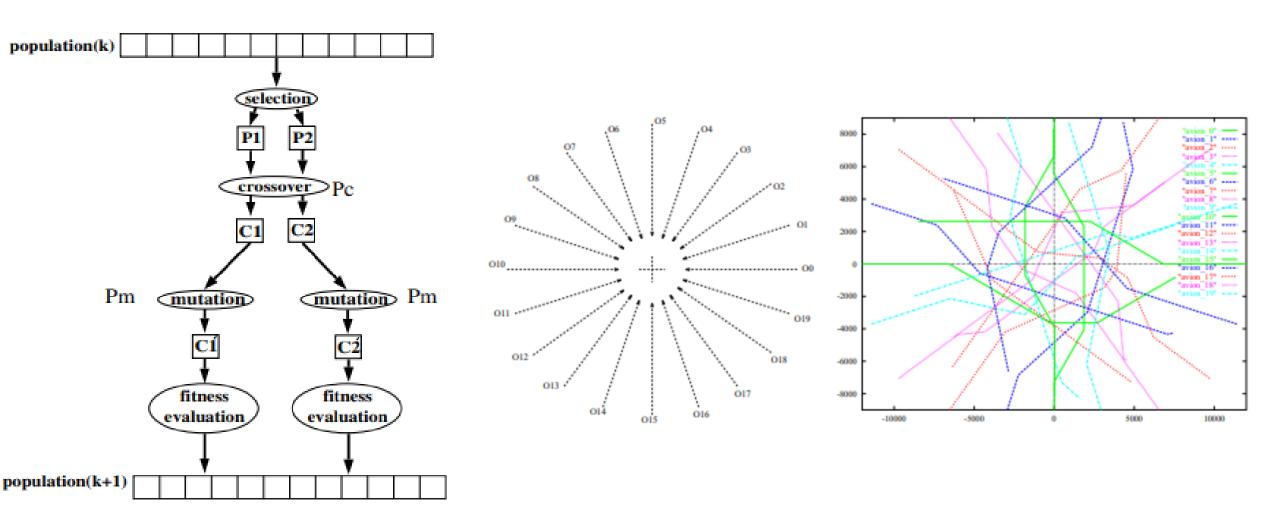
 AFR1752
 B738
 410
 410

 RVR12419
 B764
 350
 1530



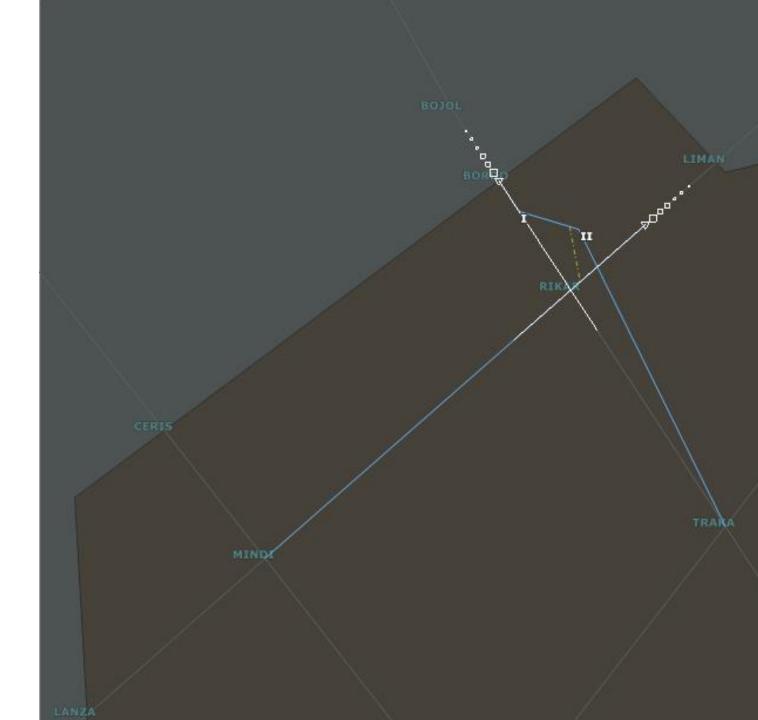
OS

 RAM2654 B753 GMMA LROP e380 t380

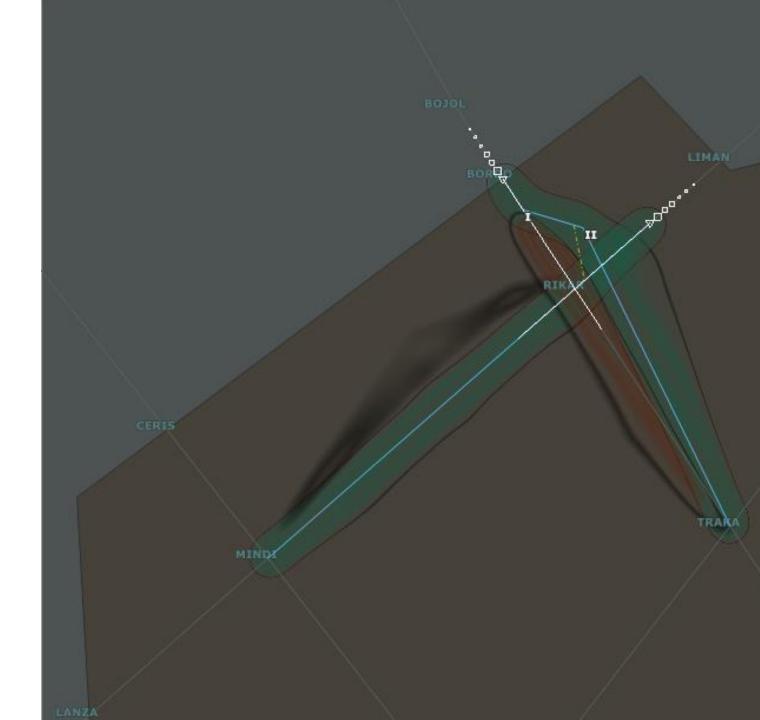


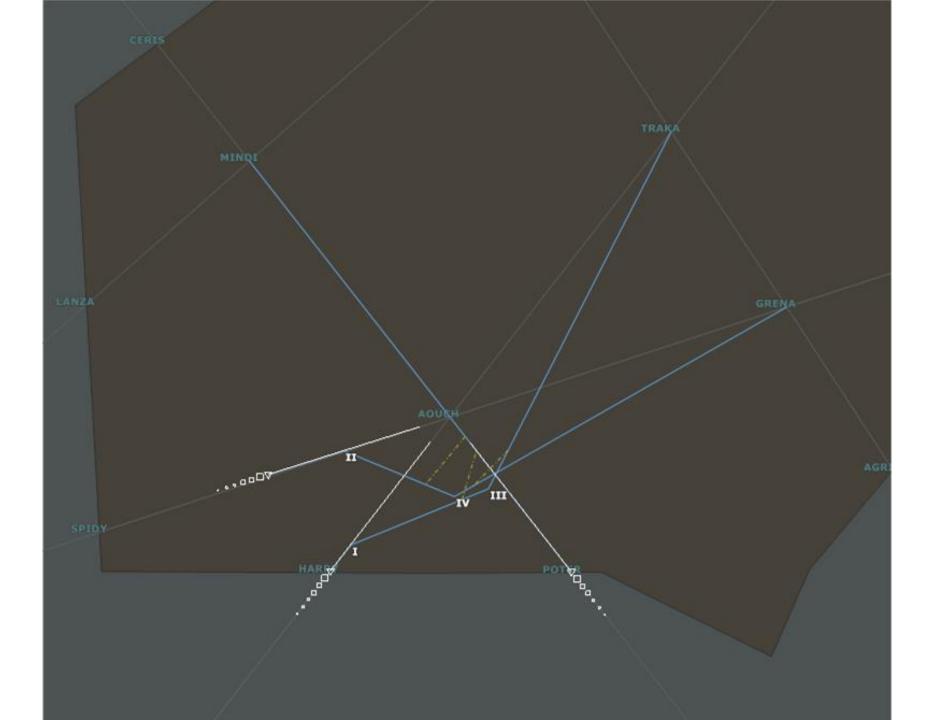
Durand, N., Alliot, J. M., & Noailles, J. (1996, February). Automatic aircraft conflict resolution using genetic algorithms. In *Proceedings of the 1996 ACM symposium on Applied Computing* (pp. 289-298).

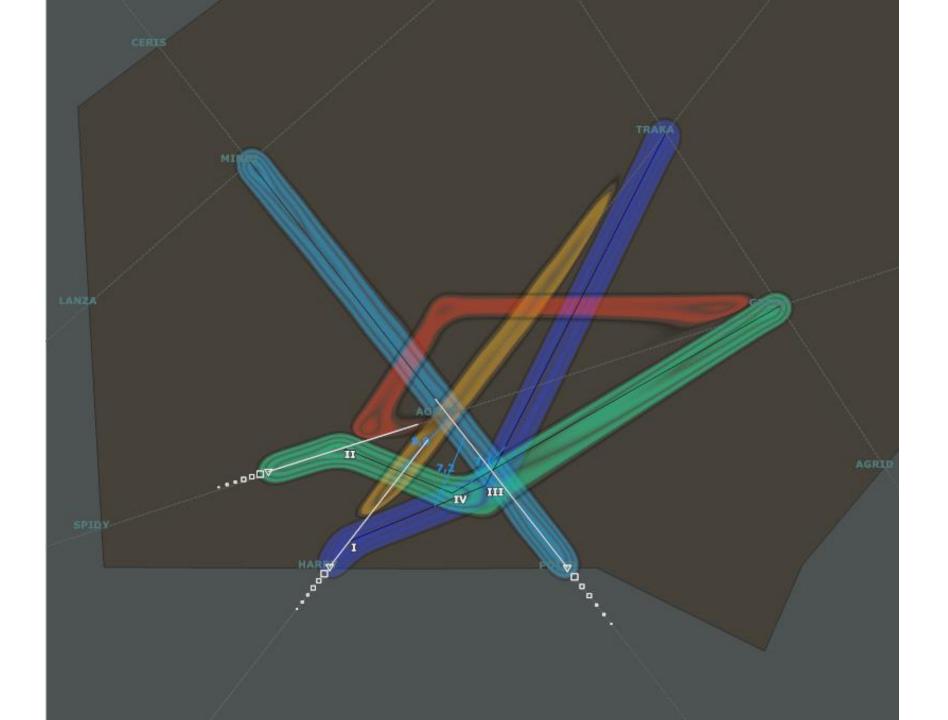
# Two aircraft crossing

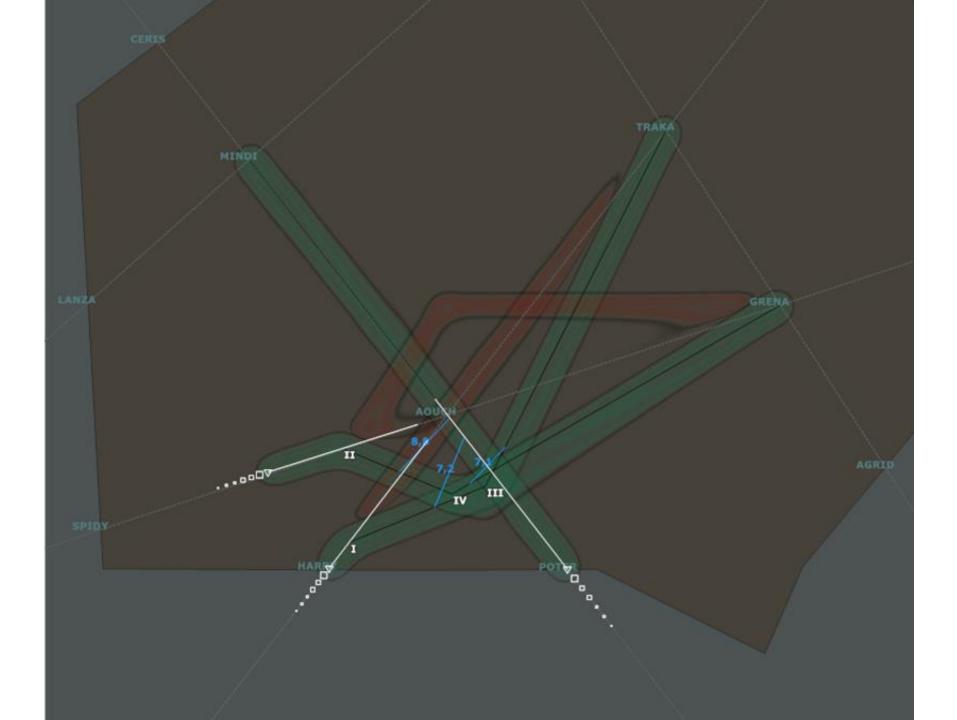


# Two aircraft crossing



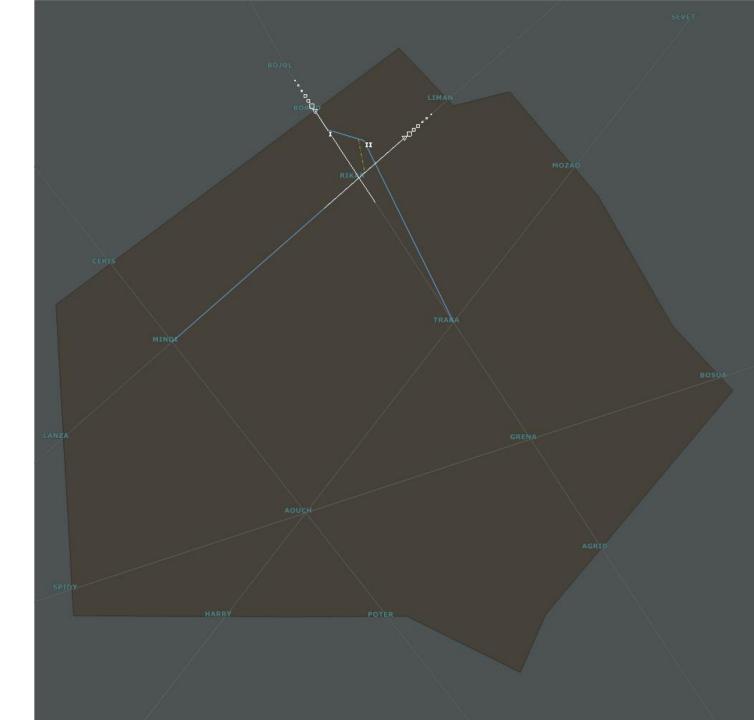




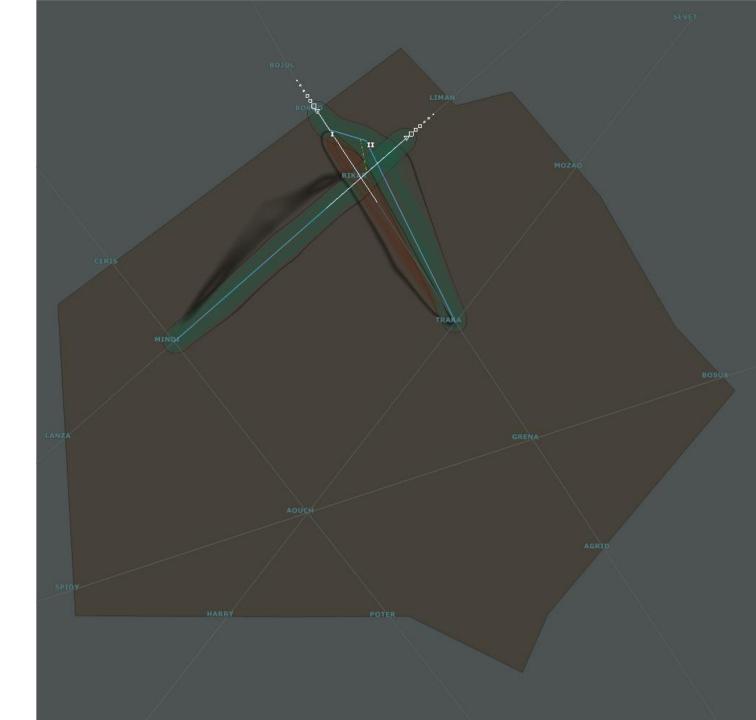


# Three levels of algorithm transparency

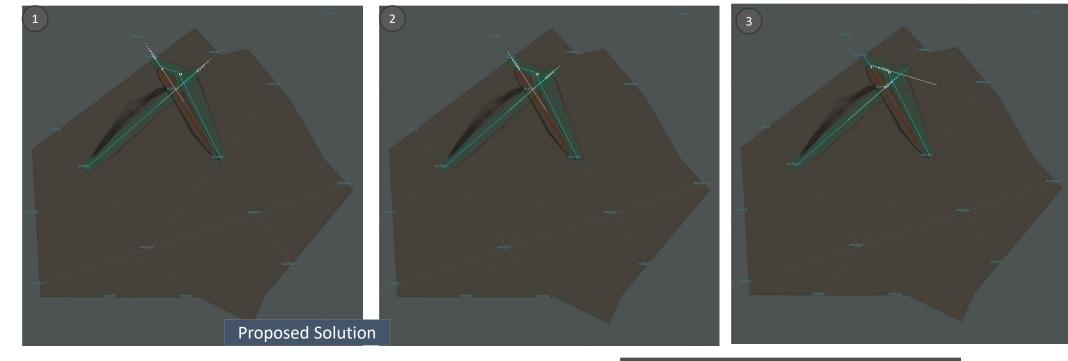
## **BB:** Black Box



# HM: Heatmap



# SB : Storyboard





# **Experimentation Setting**

21 Air Traffic Controllers (11 students, 10 experts)

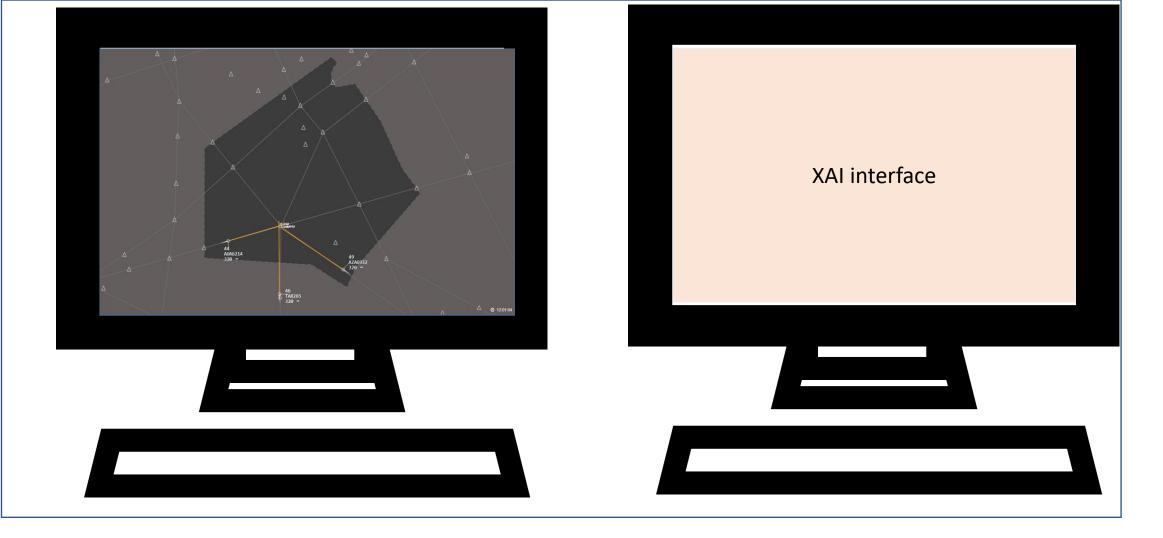
2 scenarios for each condition (Easy Hard) Black Box, Heat Map, StoryBoard

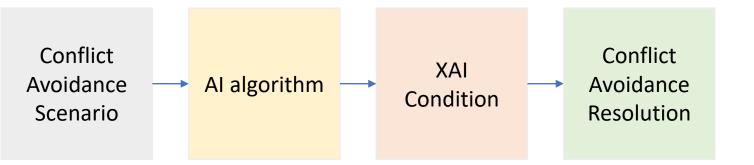
Assess the impact of each visualisation on the dimensions of acceptance, usability, human performance, work performance

Tower, En Route, Approach, Cockpit, Pseudo pilot working positions

## Simulation Running 2 minutes

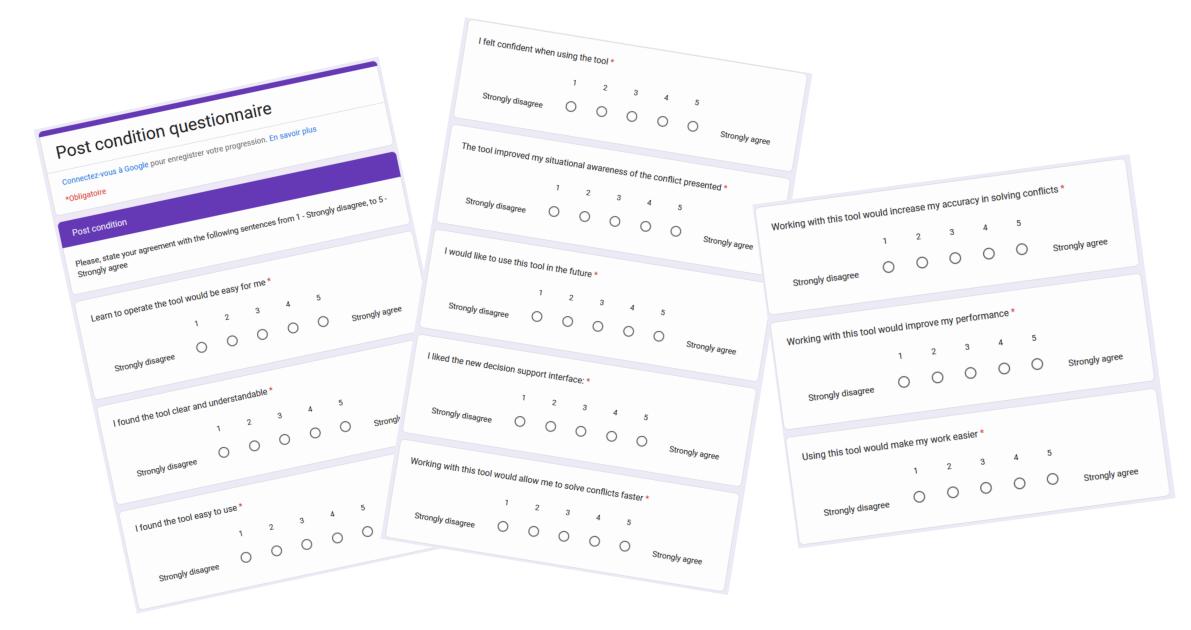


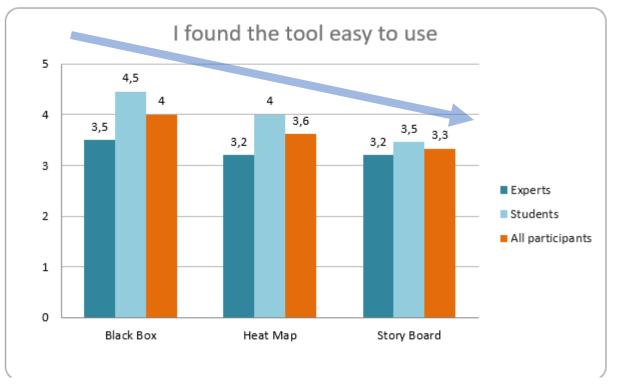


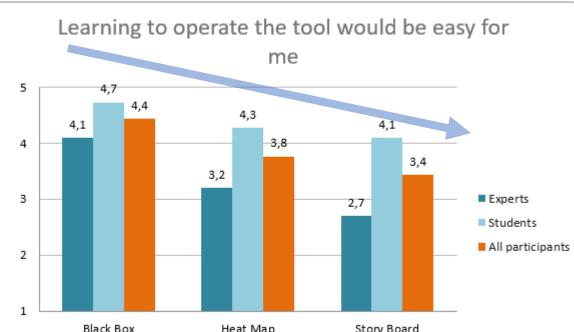


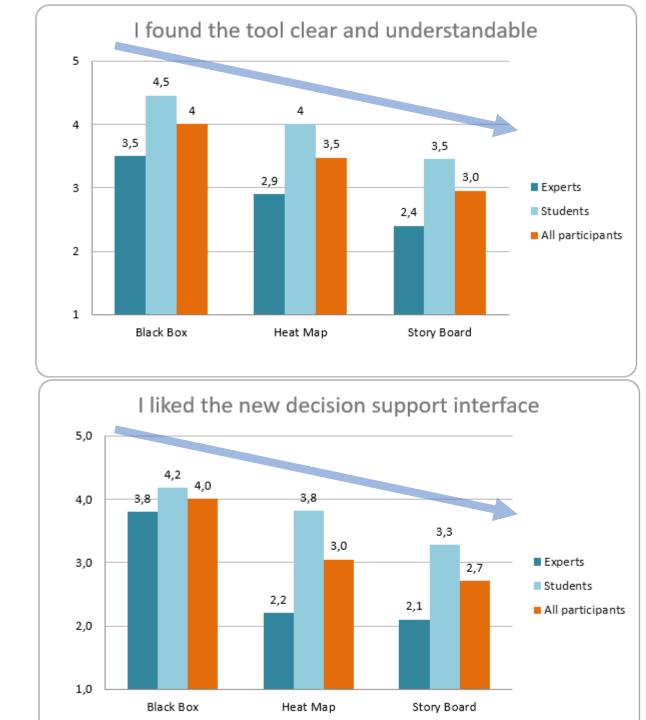
Condition 0: **BB** - Black Box Condition 1: **HM** - Heatmap Condition 3: **SB** - Storyboard

## **Preliminary Results**

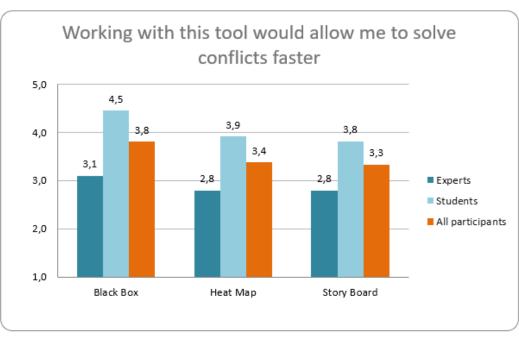


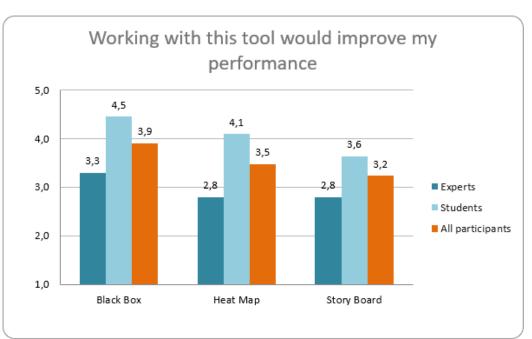


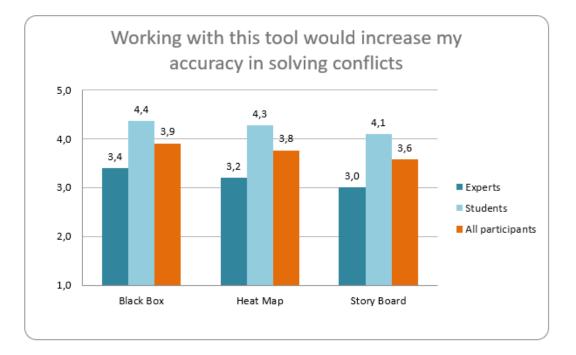


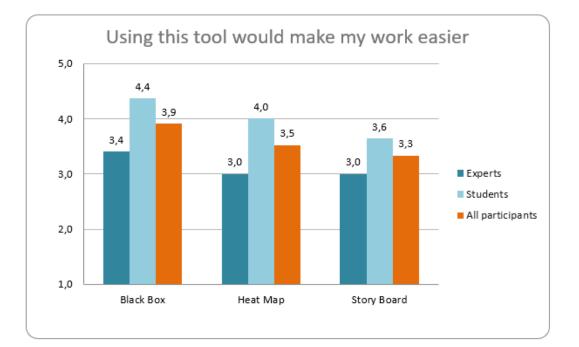


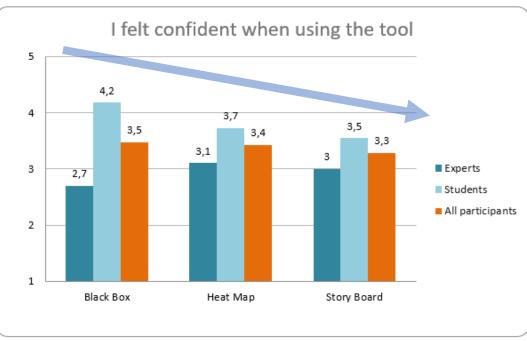
#### Impact on performance and operational support



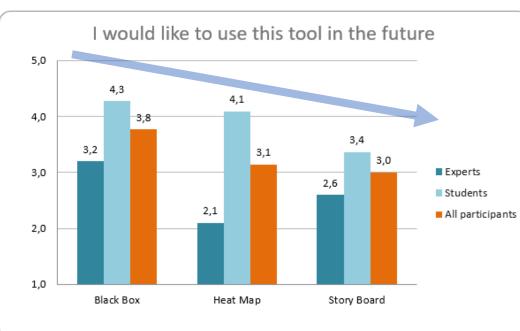




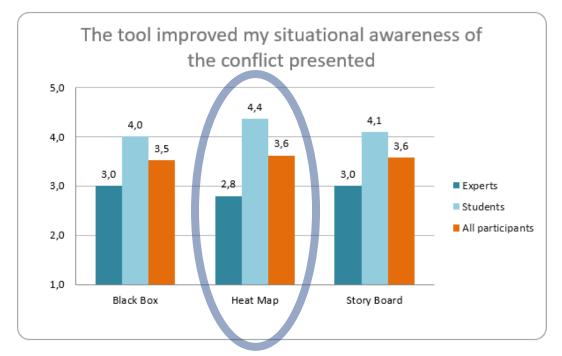




#### Acceptability



#### Situation Awareness



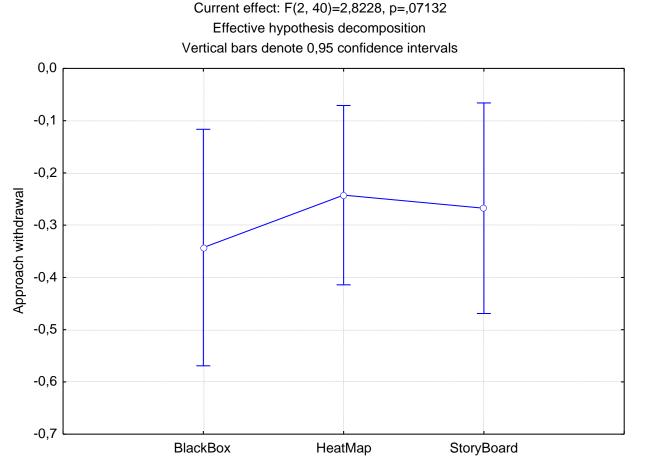


## Preliminary Results



#### https://www.brainsigns.com

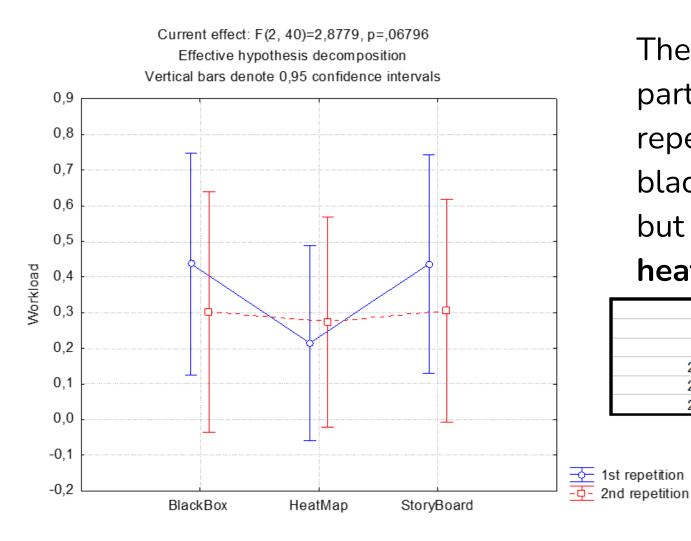
### Some preliminary results: Approach-Withdrawal



The **acceptability** of the two solutions with XAI was significantly higher with respect to the black box condition

BlackBox		0,035657	0,094591
HeatMap	0,035657		0,572333
StoryBoard	0,094591	0,572333	

### Some preliminary results: Workload



The Workload experienced by the participants among the two repetitions, decreased during the blackbox and storyboard conditions, but was already lower during the heatmap condition.

1st	BlackBox		0,003739	0,998869	0,065581	0,030196	0,064383
1st	HeatMap	0,003739		0,003313	0,217483	0,369798	0,214711
1st	StoryBoard	0,998869	0,003313		0,057580	0,026965	0,052227
2nd	BlackBox	0,065581	0,217483	0,057580		0,675436	0,955646
2nd	HeatMap	0,030196	0,369798	0,026965	0,675436		0,656614
2nd	StoryBoard	0,064383	0,214711	0,052227	0,955646	0,656614	



#### AI support and types of conflicts: the proposed AI solutions were not useful for conflicts with only two aircraft.



Human Machine Interface: Heat Maps were not straightforward to understand by ATCOs.

**Trust and XAI:** the main outcome from the collected feedback is that more trust is provided with transparent solution.

Safety: improved safety with user 'in the decision loop'.

**Training:** five ATCOs mentioned that it would be interesting to explore and understand better the advantages of the XAI solutions for training.



### Visual / Immersive Analytics



#### **RenderLine** Next-Generation of Data

Visualisation Tool







# Renderline

Moteur de rendu de trajectoires en 3D

Un outil pour la communication et l'analyse





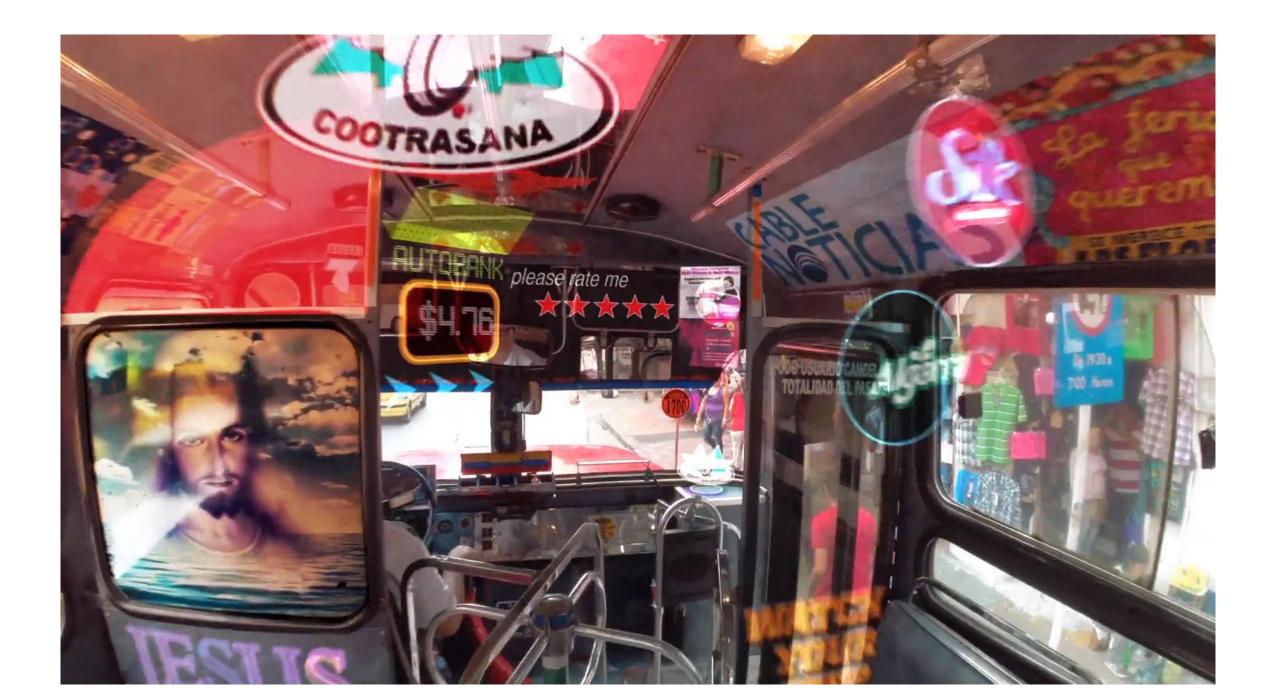
# Mai 2019

#### **Immersive Analytics**

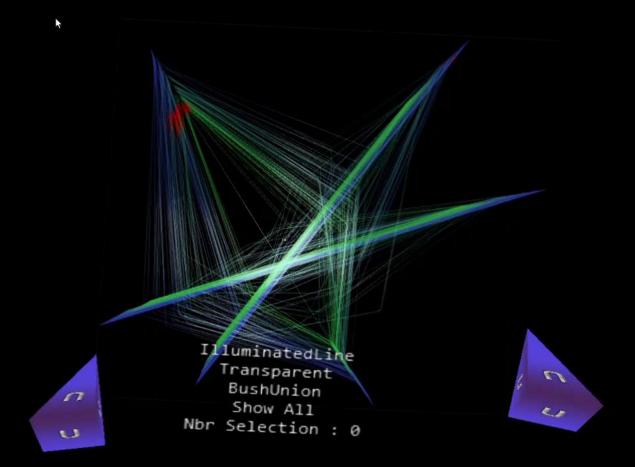




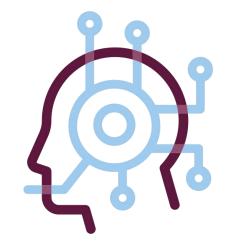




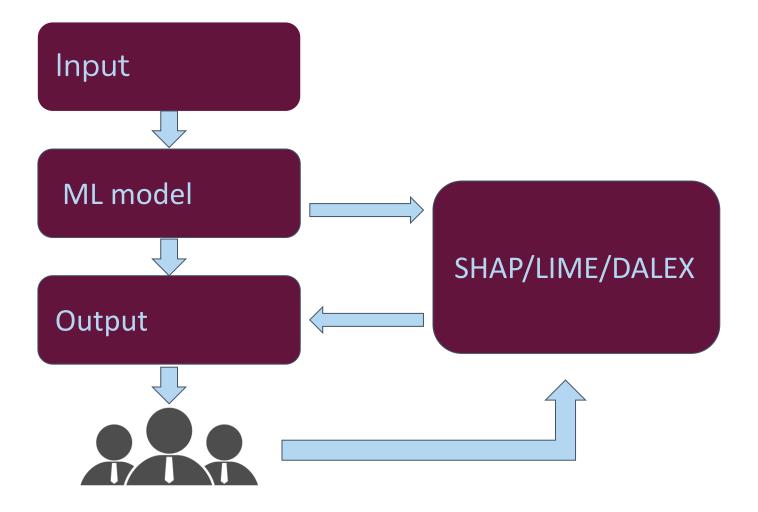
Hurter, C., Riche, N. H., Drucker, S. M., Cordeil, M., Alligier, R., & Vuillemot, R. **Fiberclay**: Sculpting three dimensional trajectories to reveal structural insights. *IEEE TVCG 2019* 



# Delay Prediction



### The underlying idea



SHAP: How much has each feature value contributed to the prediction compared to the average prediction?

The SHAP model provides explainability showing the parameters that best answer this question about a predicted delay

LIME: provides explanation by training surrogate models to approximate the predictions of the underlying black box model. It tests what happens to the predictions when variations of data is given into the machine learning model. LIME generates a new dataset consisting of permuted samples and the corresponding predictions of the black box model.

DALEX assumes that prediction is an approximation of the expected value of the dependent variable given values of different parameters. DALEX captures the contribution of a parameter to the model's prediction by computing the shift in the expected value of the prediction, while fixing the values of other parameters.



## Thank You for Your Attention

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