

Regression Methods Applied to Flight Variables for Situational Awareness Estimation Using Dynamic Bayesian Networks

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***Armilla Air Force Base
hosts the Spanish military
helicopter school***



***Eurocopter EC120:
Basic training***



***Sikorsky S76:
Instrumental Flight Rules
Search & Rescue***

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MOTIVATION

- SA is widely recognized as a key indicator of crew performance in flight safety and aviation human factors.
- SA measurements are often applied to:
 - Aircraft systems design.
 - Flight procedures & training design.
 - No notice of crew performance rating during the flight.
- SA depends on multiple variables with complex relationships and dynamic behaviour.
 - Variables related to information management are specially relevant.
 - Some of these variables can be monitored using an Electronic Flight Bag (EFB).
- DBN are a good candidate to provide a measurement of certain SA-related aspects in real time.

MOTIVATION

PhD Thesis: Implement a simulation environment to:

Perform simulated flights

Emulate an EFB used by the pilot

Use DBN to measure SA

WORKING HYPOTHESES

1. **Modern Information Management standards have a key role in maintaining high levels of SA during the flight.**
2. **Computational power of EFB can monitor pilot activities.**
3. **We seek reflections of SA in aircraft control actions and information management performed by the pilot.**

SA, IM & DBN IN CONTEXT

Definition and measurement of SA

“Situational Awareness is the field of study concerned with quantifying the perception of the environment critical to decision-makers in complex, dynamic areas”.

SA levels (Endsley, 1995)

Level 1 SA: Perception of the elements in the environment

- Monitor information & aircraft position queries performed by the pilot.
- Correlate them with pilot actions.

Level 2 SA: Comprehension of the current situation

- Level 1 + Analyze coherence of pilot actions with ideal flight parameters.
 - Expert knowledge required.

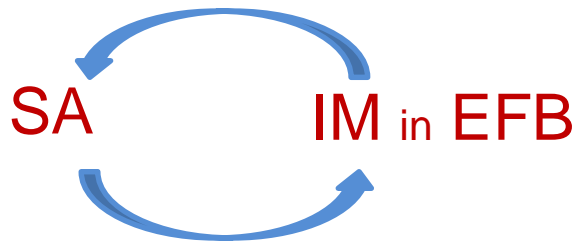
Level 3 SA: Projection of the near future

- Monitor if pilot actions (aircraft control and position & information queries) are timely.
 - Need to be careful with individual-dependent hypotheses.
- Detect when the pilot is “surprised” by flight events.
 - e.g. rough or wrong-sequenced corrections.

SA, IM & DBN IN CONTEXT

Estimation of SA based related to cockpit information management:

When a pilot operates a EFB to obtain information that should affect SA, there should be a relationship between SA and the way pilot makes use of information.



Consistency of the Aeronautical Information data model is enabled by SWIM

SWIM (System Wide Information Management):

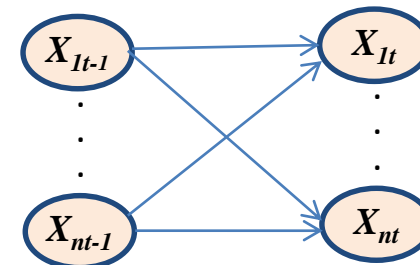
Standards, infrastructure and governance enabling the management of Air Traffic Management (ATM) information and its exchange between qualified parties via interoperable services. Under development by Eurocontrol, FAA, etc.

- * We are not expecting to perform a global assessment of SA:
 - e.g. No biometric sensors used at the current stage of research.
- * Estimation is based on a “additional pilot workload = 0” principle:
 - Applicability to real flights.
 - Reduced bias compared to classical SA rating techniques (SAGAT, Endsley et al., 1998).

SA, IM & DBN IN CONTEXT

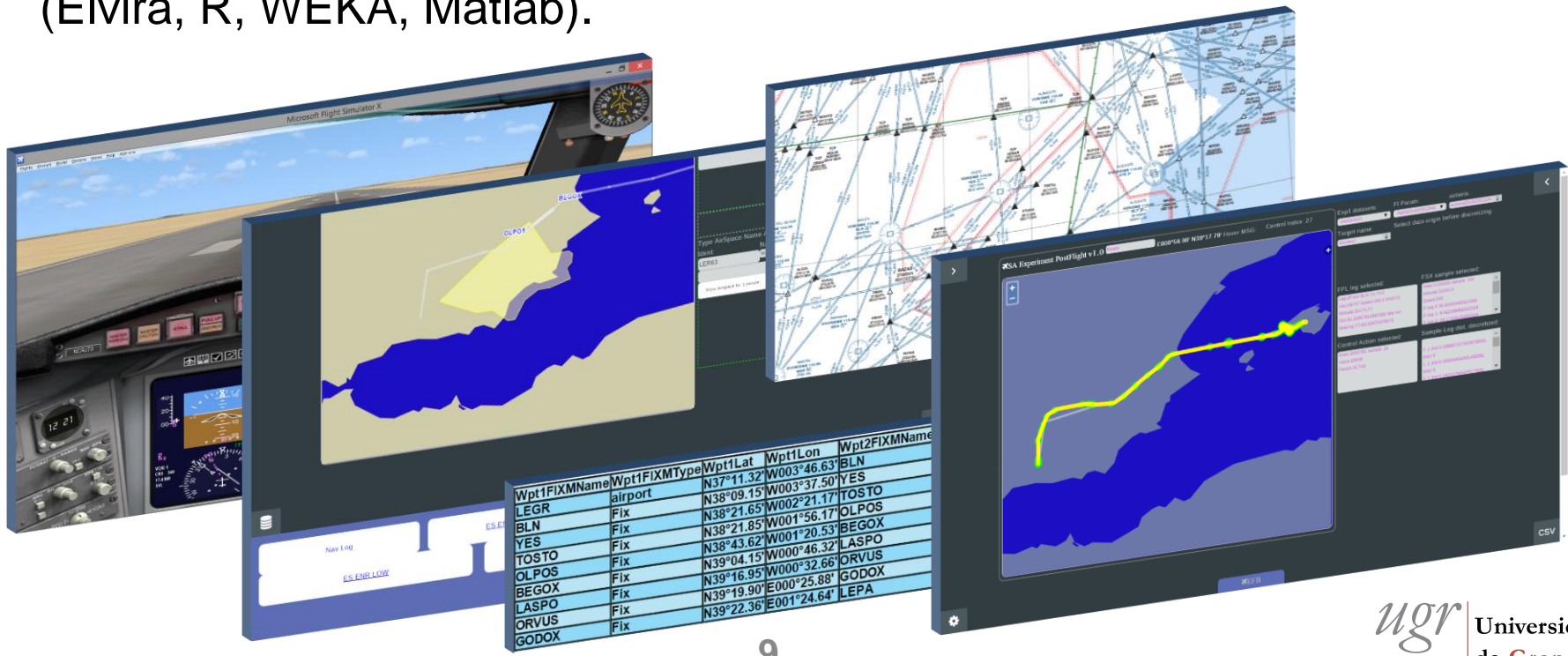
Applicability of DBN

- Bayesian networks are specially appropriate when we have a large number of variables with non-deterministic dependencies among them.
- In real-time applications, DBN can explicitly represent temporal relations between the measured variables.
- DBNs can be learned from data and, at the same time, include expert human knowledge when available:
 - **Combine effectively expert & data-inferred knowledge.**
- Inference in DBNs can be done in a short time, even in the presence of a large number of variables and observations.
- But it is necessary to adapt to DBN restrictions:
 - Only consider 1 former time step.
 - Need to discretize continuous variables
 - Without losing information.
 - Avoid EK dependency.



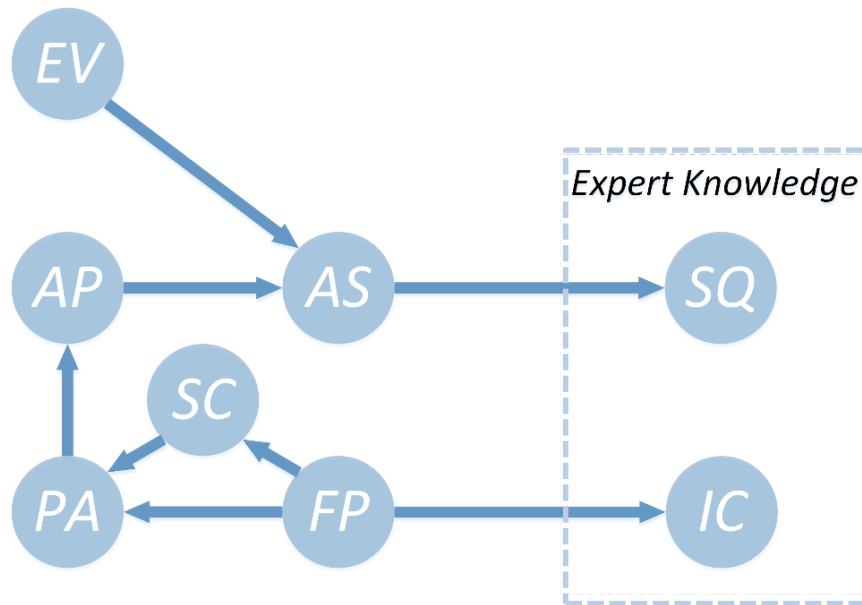
THE SIMULATION ENVIRONMENT

- Enables the performance of simulated flights.
- Provides the pilot with access to aeronautical information during the flight.
- Stores a dataset with the data collected during every experiment repetition.
- Performs certain post-flight transformations to the stored dataset and obtains data samples for exporting to external data processing tools (Elvira, R, WEKA, Matlab).



THE SIMULATION ENVIRONMENT

- The result is a dataset with different categories of variables, with an this expected simplified dependency model:



AS	Aircraft Situation	19
AP	Aircraft Parameters	4
EV	Environment Variable	4
PA	Pilot Action	21
IC	Information Check	37
SC	Situation Check	14
FP	Flight Plan	9
SQ	Situation Accuracy or Quality	11
119 variables, sampled every 7 seconds, 10 flights with durations from 54 to 64 minutes.		

- Additional expert knowledge can be added to the dataset.

SA ESTIMATION MODEL

- **10-fold cross validation:**

- In this experiment, in order to learn a model, 10 flights were performed.
- All are “correct” flights, without big deviations of the expected flight path and no “strange” pilot actions.
 - Different pilots have participated, different “flying styles”.
 - Same autopilot profile for all the flights. Limited automation.
- The DBN probabilities for each flight are learned individually using the 9 remaining flights, using Elvira.
 - Further details: Morales and Moral, 2015

- Each flight contains the variables X_i (currently 119).

- Each variable X_i has a set of parents Π_{it}

- **Assumptions:**

- Markovian process: Π_{it} is included X_{t-1}
- Stationary: $P(X_{it}/\Pi_{it})$ is time independent

SA ESTIMATION MODEL

- Markovian assumption requires the creation of **summary variables**, some of them based on expert knowledge, that avoid loss of relevant information when only the previous time step is used to predict variables.
 - We are trying that these summary variables are route-independent.
- In future stages we plan to train the model to identify a “**low awareness**” **probability** that depends on how timely & accurate corrective actions are performed, in comparison with the learned “correct” pilot reaction.
- But before addressing this, it is necessary to optimize the discretization of continuous variables...

DISCRETIZATION & REGRESSION

1. Direct discretization approach:

- Continuous variables discretized with thresholds provided with expert knowledge.
- Most of these continuous variables are such that their value on a time t is a small variation of the value on previous time $t-1$.
- Model for the DBN learned using the discretization and an optimized set of parents for each variable, according to a comparison of different metrics: BIC, Akaike, K2.

DISCRETIZATION & REGRESSION

2. Learning a discretization from data:

- Based on a two step optimization:
 - First we compute an optimal discretization for each variable with the empty graph (no parents for any variable).
 - Then we learn the optimal DBN structure as in the previous method.
- To find an optimal discretization we use a greedy algorithm:
 - The possible interval limits are the middle points between the actual points of the variable in the learned data.
- Solves the problem of expert knowledge dependency.
- Performance not promising enough.

DISCRETIZATION & REGRESSION

3. Learning the discretization + Linear Regression:

- Initial numerical estimation of each continuous variable X_{it} using linear regression and with variables in previous time as predictors Y_{t-1}^i .
 - Using **R** package *bestglm*
 - Number of predictors is limited in order to avoid overfitting:
 - number chosen: 4 predictors.
 - the best model is chosen using **BIC** criterion.
 - deterministic variable based on known values of variables at time step $t-1$:

$$Z_{i(t-1)} = a + b_1 Y_{t-1}^1 + \dots + b_k Y_{t-1}^k$$

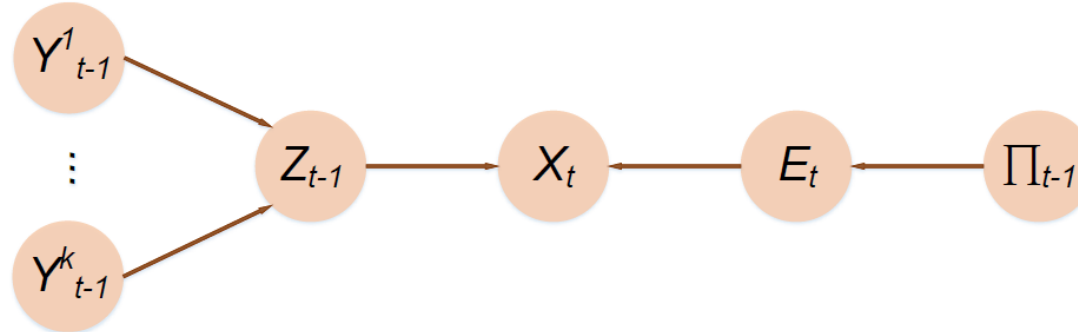
- error in the prediction:

$$E_{it} = X_{it} - Z_{i(t-1)}$$

DISCRETIZATION & REGRESSION

3. Learning the discretization + Linear Regression:

- Once the regression is computed, the dataset is expanded by including the artificial variables Z_{it} and E_{it} .
- The full model is represented by:



Where Π_{t-1} are the parents of variable X_t .

RESULTS

Different discretization strategies are compared using the Logarithm of Probability (LP):

$$LP = \log(P((X_t)_{t \in T} | M)) = \log(P(X_{t_0})) + \sum_{t \in T \setminus \{t_0\}} \log(P(X_t | M, X_{t-1}))$$

Where M is the learned model and $t_0 \in T$ is the initial time.

Higher values of LP are better. The best result corresponds to K2 score using summary variables.

Using regression always provides a higher LP:

	min Thr	10 Thr	20 Thr	50 Thr	100 Thr	regression
BIC Score	$-9.300e5$	$-6.690e5$	$-5.699e5$	$-6.360e5$	$-5.676e5$	$-2.475e5$
Akaike Score	$-9.310e5$	$-6.682e5$	$-5.655e5$	$-4.271e5$	$-5.075e5$	$-2.450e5$
K2 Score	$-9.305e5$	$-6.694e5$	$-5.653e5$	$-4.188e5$	$-3.602e5$	$-1.989e5$

Table 1: LP values for the estimation of all variables and using summary variables.

CONCLUSSIONS

- SA estimation is a relevant topic in flight safety & aviation human factors.
- We have implemented a simulation environment to collect relevant variables related with SA.
 - In accordance with state of the art aviation information management standards.
- Due to the limitations of DBN, it was necessary to perform a careful approach to the discretization of continuous variables and the creation of summary variables.
 - Regression methods allow us to obtain an accurate discretization that does not require specific expert knowledge.
- In return, we obtain a powerful tool to perform dynamic assessments based in large number of variables with complex relationships.
 - Future research to find a relevant measurement of SA.

Thanks for your attention

Questions?