

Using the Decision Ladder to Understand Human Decision-Making Processes During a UAV-Equipped Search and Rescue Mission

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ABSTRACT

Uncrewed Aerial Vehicles (UAVs) have the potential to be valuable technological additions to Search and Rescue (SAR) units. Advancements in the development of algorithmic functionality capable of detecting and classifying targets in the terrain continue to enhance the capability of UAV technology. However, prior to incorporating such technology within a SAR unit, it is important to consider the decision making that takes place during SAR missions in order to ensure that the functionality is embedded successfully. Here, we explore the utility of the Decision Ladder for characterising the decision making processes of SAR personnel when utilising a UAV during a hypothetical SAR mission. Five operator interviews with SAR personnel from across the United Kingdom were used to populate a Decision Ladder model. Analysis of the model suggests design recommendations for a future decision aid that could improve UAV team performance during a UAV-equipped SAR mission.

Keywords: Uncrewed aerial vehicles, Search and rescue, Decision ladder, Human factors integration

INTRODUCTION

The operation of Uncrewed Aerial Vehicles (UAVs) in goal-orientated contexts such as Search and Rescue is complex. Throughout the UAV operation, the human operators are separated from the air vehicle but must maintain visual line of sight. As a result, the Human Machine Interface (HMI) serves as the primary information resource for ensuring the UAV is operating safely, efficiently, and in compliance with regulations that set operational boundaries for the airborne parameters of the vehicle (e.g., Civil Aviation Authority; CAA, 2023). The HMI must present the required information to the human operators at the right time to ensure decision-making does not become erroneous or ineffective (Riley *et al.*, 2010; Lynch *et al.*, 2022). However, all too often there is a tendency to design UAV systems using technical requirements that fail to consider who is interacting with the system, and more importantly how (Steane *et al.*, 2023 *in press*). This tendency has given rise to HMIs that are complex and rely on the skill and technical knowledge of the human

operator to exploit the capability afforded by the UAV technology (Golightly *et al.*, 2022). In doing so, a high reliance is placed on the human operators responsible for controlling the UAV, and any benefits of autonomous functionality may be lost. To manage this, novel algorithmic developments and improved sensor technologies are being designed to improve the level and quality of information available to the human-UAV team (Goodrich *et al.*, 2007; Parnell *et al.*, 2022). The goal of adding this functionality is to assist users in reducing workload, maintaining situation awareness and reducing the risk of errors.

It is during these early design stages that Human Factors Integration (HFI) methodologies can be leveraged to transfer the user's needs into system requirements for use by system architects and software engineers (Bruseberg, 2008). The current work aims to demonstrate the utility of exploiting Human Factors (HF) methodologies to extrapolate such insight. Specifically, the use of the Decision Ladder was examined. We used the case study of SAR due to the recent interest in integrating image classification algorithms within UAV systems to detect and label targets within the terrain. It is anticipated that such functionality would reduce the workload placed on the human operator responsible for interpreting the payload data received from the UAV (i.e., the Payload Operator). However, when designing these support mechanisms, it is important to consider where they can be extended to present the information needed to manage the UAV and support the goal of the wider SAR team. In other words, the goals and tasks encompassed within the work domain should be considered when designing automated decision aids (O'Neill *et al.*, 2020). The Decision Ladder investigates key tasks in complex sociotechnical systems (Vicente, 1999) whilst also analysing the activities that take place in "decision-making terms" (Rasmussen *et al.*, 1994, p. 58). As such, the Decision Ladder is employed in the current study to support the integration of automated functionality by eliciting insight into why, who, when, where, what and how a decision aid would be used in the SAR application (Steane *et al.*, 2023 *in press*).

A template of the decision ladder is shown in Figure 1. The model is comprised of two types of node: rectangle nodes represent the information processing activities conducted by the decision-maker and circular nodes represent the resultant states of knowledge (McIlroy and Stanton, 2015). The Decision Ladder begins with an alert in the environment which prompts the decision-makers to assess and respond to a situation. The left-hand side of the decision ladder shows the subsequent situational assessment made by the human actors through a means of gathering information and identifying the options available to respond to the initial alert (Banks *et al.*, 2020). Once the optimal option is identified through the application of knowledge-based processing, the right-hand side of the ladder displays the planning and execution of this response. Although the Decision Ladder is shown as a sequential framework, a decision-maker equipped with a high level of expertise could shortcut aspects of the ladder to enable faster decision-making and more decisive action (McIlroy and Stanton, 2015). This is represented using two cognitive shortcuts. Firstly, a *leap* is a shortcut between two states of knowledge and is shown by the dashed line (see Figure 1). This shortcut occurs when a

state of knowledge triggers another; for example, by understanding the system's state, the decision-maker may already know what task to perform based on past experiences. The second shortcut is referred to as a *shunt* and occurs when the decision-maker requires further information from the environment in order to carry out a task. It is shown as a solid line between an information processing activity and state of knowledge (see Figure 1).

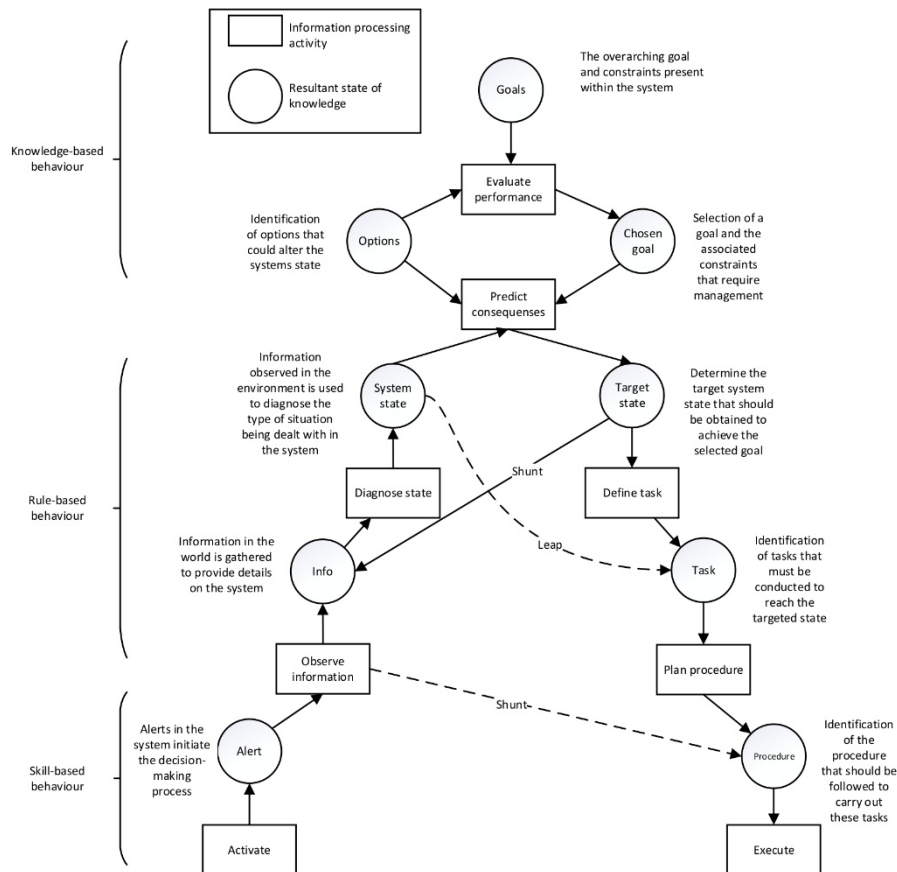


Figure 1: Decision Ladder template adapted from Parnell et al. (2021).

METHOD

In order to populate the Decision Ladder, semi-structured interviews were conducted with operationally active members of SAR units across England, Scotland and Wales. The structure of the interview was designed using the Schema World Action Research Method (SWARM; Plant and Stanton, 2016). This methodology was developed to measure the perceptual processes of aeronautical pilots. However, Plant and Stanton (2016) envisaged its use being extended beyond the domain of crewed aviation. The full SWARM repository comprises 95 cognitive prompts. Each prompt was designed based on the sub-types of the Perceptual Cycle Model (PCM; Neisser, 1976). Plant and Stanton (2016) recommended down-selecting and adapting the SWARM prompts based on the objectives of the study. As such, 30 and 33 SWARM

prompts were selected for use in the current work for the traditional ground search and UAV-equipped interviews, respectively.

Participants

A total of five participants were recruited from Mountain Rescue Teams across England, Scotland and Wales to take part in the operator interviews. Participants were currently operational SAR volunteers with varying amounts of operational experience. Within the participant sample were individuals with experience of working as a ground searcher, UAV Pilot and/or Search Manager. Those who had experienced UAV piloting held the relevant qualifications required by the CAA for operating UAVs within civil operations. The study received ethical approval from the University of Bristol Ethics Committee (Reference 10785).

Procedure

Prior to beginning the interview, participants read through the participant information sheet and, once they were happy to take part, signed the consent form. At the beginning of the interview, participants were shown a hypothetical SAR scenario involving an individual reported as missing in a National Park. Once the scenario was presented, the interviewee went through the set of SWARM prompts selected for inclusion. The participants described the approach that would be taken to respond to the missing person case when equipped with or without a UAV. In order to understand how the decision-making processes of the UAV team could be supported by a decision aid that displays image classification information, at the end of the interview, the following questions were asked:

“If the UAV could provide a confidence estimate for the image classification...

- 1. How would you interpret this information?*
- 2. Would it be a useful piece of information?*
- 3. Is there any other information that would be useful in assisting with image classification.”*

Each interview was recorded using a Dictaphone to enable for data transcription and subsequent analysis.

DATA ANALYSIS AND MODELLING

The Decision Ladder was developed iteratively following a period of data collection. Two Human Factors researchers were involved during the development of the Decision Ladder. Once the researchers reached a point of agreement, the final amalgamated model was sent to an independent Subject Matter Expert which served as a validation measure to ensure the models accurately depicted the practice of SAR teams.

Decision Ladder for a UAV-equipped SAR Response

The final amalgamation of the Decision Ladder for a UAV-equipped SAR response is shown in Figure 2 and begins when the Payload Operator identifies a potential sighting on their display. It was recognised that during both

a UAV-equipped response and a traditional ground search, the goal of the system would always be to exploit any new information received from the environment to adapt the current plan in a way that increases the likelihood of locating the missing person (MISPER). Indeed, any information presented from the UAV would likely lead to further human involvement through the re-allocation of resources by the Search Manager responsible for controlling and directing the search effort from a control vehicle. Yet, prior to re-allocating any assets or teams, it is important that the finding identified on the display is subject to a validation process to establish its reliability and relevance to the search. Whilst an object found on the ground would also be subject to a similar assessment, the validation process used during a UAV-equipped response was more complex due to the limited availability of information on the display.

The sparsity of resources in the SAR environment implicates the need to gather evidence to facilitate informed decision-making by ensuring the data collected by the UAV corroborates the information already known about the MISPER. For example, the last known location of the MISPER and the current location of the UAV may be reviewed in tandem to calculate the distance travelled between sightings. If the MISPER had travelled a respectable distance but was an elderly individual, the sighting may be ruled as irrelevant under the assumption that they would be incapable of carrying out such a journey due to their physical condition. Other information aspects that could be attended to include the environmental surroundings shown in the imagery. Both the terrain and its environmental stimuli (e.g., trees) provide a basis on which to infer the sighting's size, shape and movement. These characteristics provide further contextual information that aid in the classification and initial validation of the sighting.

It was also found that the Payload Operator could support the Pilot by reviewing the status and health of the UAV to ensure the vehicle is safe during its deployment and can continue to be used as part of the response. This involves monitoring parameters such as the battery life to determine the time left until the battery reaches a critical level. Finally, information would be reviewed to understand the availability of resources in the SAR environment that could be used to aid the response. The combined understanding of the sighting's relevancy, the state of the SAR system, and the health of the UAV would be used to identify a set of options that could be used to respond to the finding. This decision falls to the Search Manager; an individual considered to hold a significant amount of experience to effectively coordinate the search effort. For this reason, the Search Manager may intrinsically recognise the optimal course of action to validate the UAV data once the system state is diagnosed. This shortcut is displayed as a leap from the 'system state' node to the 'task' node that resides on the right-hand side of the Decision Ladder (see Figure 2).

The procedure shown on the right-hand side of the ladder outlines the processes conducted when re-tasking the UAV to investigate a sighting. Ordinarily, the UAV would land or be manoeuvred closer to the location where the sighting was originally identified. Subsequently, the Pilot would safely navigate the UAV within the search region whilst the Payload Operator monitors the display for the initial sighting. During the vehicle's deployment the

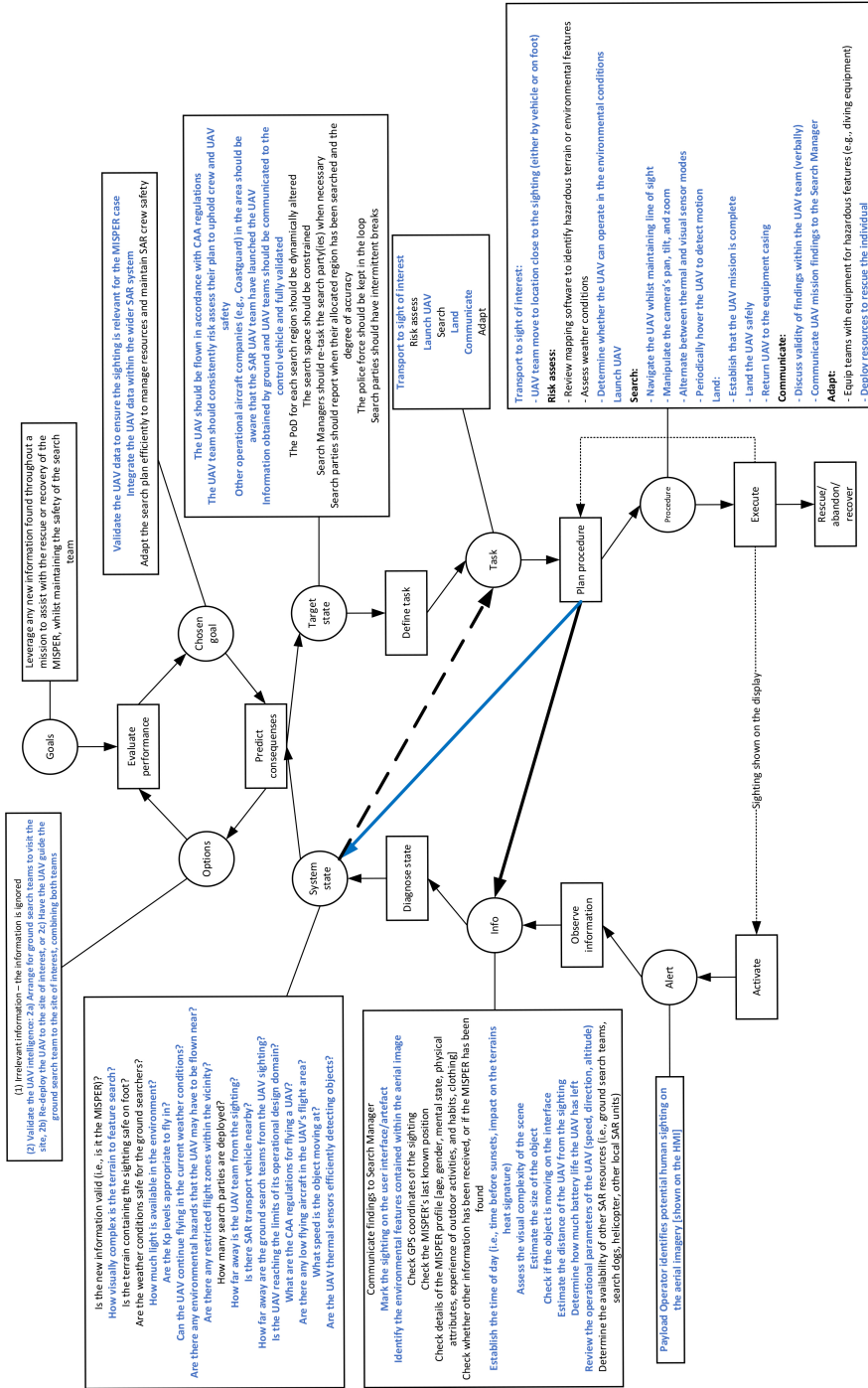


Figure 2. Condensed version of the Decision Ladder. Note. Text in blue indicates information used specifically for UAV-equipped missions.

UAV team could refer back to information on the left-hand side of the ladder. For example, the location of the sighting may be reviewed to maintain awareness of where to navigate the UAV and allocate attention when looking at the aerial imagery. This referral is depicted as a shunt between the ‘plan procedure’ processing activity and the ‘information’.

It is also critical for the UAV team to maintain awareness of the systems state, such as its parameters, to ensure the vehicle is adhering to regulations stipulated by the CAA. These regulations currently prevent the UAV being flown beyond visual line of sight and more than 120 metres from the earth’s surface (CAA, 2023). The continuous monitoring of these parameters is shown as an additional shunt between the ‘plan procedure’ activity and ‘system state’ information node.

In the context of UAV-equipped SAR missions, it is important to demonstrate the flexibility of SAR responses. This is because new information is used to shape and guide the direction of a search effort, meaning the planned procedure is always subject to change. To encapsulate this within the Decision Ladder, a link was placed between the ‘plan procedure’ and ‘execution’ nodes to highlight the way in which a plan could be altered following the reception of new intelligence or changing circumstances in the environment (see Figure 2). An additional link was also embedded between the ‘execution’ and ‘activation’ nodes to demonstrate the cyclical nature of SAR (see Figure 2). Fundamentally, SAR involves continuously adapting and carrying out the plan using information from the search teams until an outcome is reached. Such outcomes could see the MISPER being rescued or the mission being abandoned when the safety of the SAR team is at risk.

DISCUSSION

The design recommendations identified using the Decision Ladder are shown in Figure 3 with the potential cognitive shortcuts that could be enabled following their integration. First, the implementation of automatic functionality for imagery analysis (i.e., object detection, classification) could manifest in a shortcut between the ‘alert’ to ‘task’ node. This is because the presentation of a detected sighting may trigger an intrinsic understanding of the task needed to respond to the potential finding. Nevertheless, SAR operators indicated that the findings of an automated system would likely not be taken on face value. Instead, further information would be gathered to corroborate the relevance of the sighting and determine an appropriate response. In order to support these information gathering activities, the decision aid could utilise the information aspects used to review a sighting’s relevancy (e.g., size, motion, speed, distance travelled) by presenting these to the Payload Operator alongside the automatic detection notification. This support may facilitate the second shortcut between the ‘information’ and ‘task’ nodes.

The support mechanisms presented thus far would not enable an intrinsic recognition of the required procedure as no SAR mission involves the same type of MISPER. For this reason, the procedure used to respond to a sighting must be reviewed on a case-by-case basis to determine the most suitable response. For instance, it may not be appropriate to navigate a UAV within the vicinity of a vulnerable individual as the noise generated from the vehicle

may invoke feelings of fear in the MISPER, causing them to conceal themselves from the SAR team. However, a decision aid presenting the options available to the Search Manager could help determine which resource to allocate by displaying the set of resources available for re-tasking. These options could be displayed in order of their appropriateness based on factors such as the remaining UAV battery, mission timing requirements and the type of MISPER profile being dealt with during a mission. In doing so, a shortcut may be enabled between the ‘options’ and ‘procedure’ nodes.

CONCLUSION

The uptake of ‘off-the-shelf’ UAVs for SAR missions has provided an invaluable technological tool for aiding human responders operating on the ground. Nevertheless, the generic design of these systems has resulted in UAV teams adopting strategies to manage the technological constraints associated with the technology. These strategies extend to the analysis of the aerial imagery, navigation management, status and health monitoring, and the communication of information with search management personnel. It should also be recognised that these strategies evolve quickly, with operators finding adaptive techniques and refining their Standard Operating Procedures to manage emergent issues.

The design recommendations proposed in the current work aim to provide support mechanisms that could help manage each of these aspects. By identifying how work is currently done using the Decision Ladder, it became possible to consider: (i) why intervention is need; (ii) who would utilise the support; (iii) when the support mechanism would be used; (iv) what it would be used to achieve; and (v) how this would be embedded or modified to fit current practice (Steane et al., 2023 *in press*). Even so, it is important that these models are viewed as representations of SAR practice as opposed to a fully fledged descriptions of UAV-equipped SAR missions. For this reason, further validation of the Decision Ladder is required using empirical methods to ensure the assumptions from this work accurately represent SAR practice (Banks et al., 2020; Parnell et al., 2021). Moreover, future work could extend the current research effort by modelling the practice of SAR teams for more specific UAV mission vignettes to encapsulate the broad range of techniques used to embed UAVs within SAR operations (see Table 1 for a set of mission vignettes developed using insight from UAV operators).

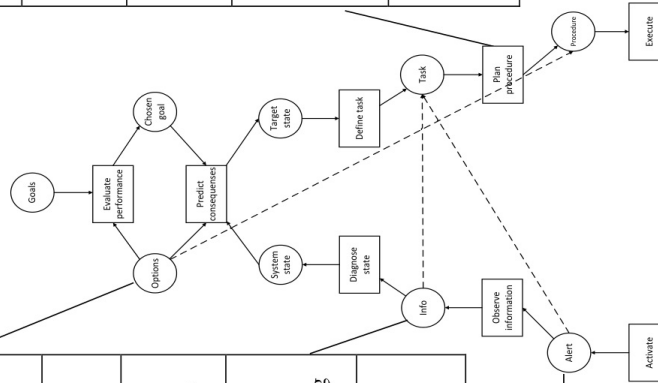
Table 1. UAV mission vignettes.

UAV mission vignette	Description
Sound sweep	Use an onboard speaker to broadcast messages to possible victims on the ground.
Region search	Investigate a region of land using a formalised search strategy.
Localised search	Investigate a localised geographical feature (e.g., gorge) using a formalised search strategy.
Parallel search	Use the UAV to guide ground search teams to a potential sighting on the ground.
Terrain mapping	Use the UAV to supplement knowledge of the terrain and inform the development of a search plan.

Current practise Determine the most appropriate resource to re-task to validate a sighting	Design recommendation Recommend whether to re-task the UAV team or ground search team based on their location, remaining battery life of the UAV and the MISPER profile Estimate the time each resource would take to be re-tasked
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Current practise Check for motion	Design recommendation Automatically detect motion in the terrain
Current practise Check GPS coordinates of the sighting	Design recommendation Automatically calculate the speed of the object Automatically calculate the GPS coordinates of a sighting
Current practise Identify the environmental features contained in the aerial image	Design recommendation Clearly display the GPS location of the sighting landmark and the sighting Automatically display the distance between the size of an object in the terrain
Current practise Check the missing person's last known location	Design recommendation Display the last known location of the sighting on the display Calculate the distance between the sighting's location and the last known location of the missing person
Current practise Check environmental conditions	Design recommendation Automatically notify the UAV team when the environmental conditions are beginning to breach the Operational Design Domain of the UAV (e.g., wind speed exceeds capabilities)

Current practise Periodically hover the UAV	Design recommendation Provide automatic reminders to hover the UAV periodically Provide a control input that automatically hovers the UAV upon request
Current practise Alternate between sensor modes	Design recommendation Automatically recommend the sensor mode to activate Enable the display to show both visual and thermal modes concurrently on request
Current practise Monitor the parameters of the UAV	Design recommendation Measure the distance of the UAV from the human operators Display the distance metric using a traffic colour coding scheme to indicate the compliance with CAA regulations (e.g., green indicates compliance)
Current practise Communicate the UAV mission findings to the Search Manager	Design recommendation Enable the UAV mission findings (e.g., sightings) to be transmitted to any mission planning applications used in the control vehicle



Current practise Identify an object/sighting that looks out of place	Design recommendation Automatic object detection Automatic object classification Confidence intervals
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