Can Structured Analytic Techniques Enhance Intelligence, Surveillance, and Reconnaissance (ISR) Assessments When Providing Incremental Information

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ABSTRACT

Structured Analytic Techniques (SATs) have been incorporated within academia over the past several decades in an effort to enhance decision-making outcomes. Although, little research has focused on the effect of SATs within an operational environment such as Intelligence, Surveillance, and Reconnaissance (ISR) Mission Sets. The objective of this study was to evaluate the effectiveness of SATs when providing a vague narrative to human operators. Differences in response accuracy were investigated when a SAT was coupled with a narrative based on two different information workflow methodologies (i.e., complete narrative all at once or the narrative divided into incremental sections). Moreover, six groups of 25 participants each (N = 150) were randomly assigned to one of three analytic techniques and provided a narrative with all the information all at once or the information provided in incremental sections. The SATs implemented were the Method for Defining Analytical Questions (MDAQ), which was developed in-house by our ISR subject matter experts (SMEs), a Scaffolding approach, and a Control approach. The findings provided evidence that implementing the MDAQ approach when given incremental information significantly improved performance compared to all other group configuration (p<0.05). This discovery will not only support ISR tool development, training exercises, and technology transition but could be beneficial in enhancing human computer interactions and human decision-making across other domains within academia and industry.

Keywords: Structured analytic techniques (SATS), Intelligence, surveillance, and reconnaissance (ISR), Method for defining analytical questions (MDAQ)

INTRODUCTION

ISR Background

The ability of military operators to effectively collect, interpret, and respond to adversary ground movement patterns relies heavily within the Intelligence, Surveillance, and Reconnaissance (ISR) domain. Decisions made within an ISR environment can influence the success or failure of a mission, the safety and well-being of military personnel, and the ability to command and maintain military superiority. However, with the continual advancements in adversary disruptive technology, deceptive functionality, and imagery jamming technologies, ISR data collections may not capture or contain the critical information needed to inform strategic planners on future planning guidance.

ISR data collections from systems of electro-optical/infrared (EO/IR) sensors can range from full-motion video (FMV) coverage, synthetic aperture radar imagery (SAR), and distributed heatmapping which all play a critical factor in supporting strategic future decision-making capabilities. During these various data collection procedures, there are critical variables that can provide additional information to support the operators' decision which are known as Essential Elements of Information (EEI). If EEIs are correctly identified by the operators, they can provide underlying evidence and justification to inform future decisions with non-biased rationale. Moreover, military EEIs can provide valuable information on assets, vulnerabilities, current and future threats, and controls. EEIs can be assessed based on the reliability and creditability of the information gathered to determine if additional collections are required to make an informed decision with high confidence. However, there are specific instances where additional data collections are not feasible or a decision needs to be made within a small time window. At these instances, Structured Analytic Techniques (SATs) and cognitive reasoning becomes the foundation that support decisions moving forward (Coulthart, 2017).

SATS: Supporting Decision-Making

SATs have been widely used within academia over the past several decades as a support system to enhance critical thinking and logical reasoning when confronted with complex situations (Gioia and Thomas, 1996). Early methodologies discovered that following a structured approach could significantly improve detection of relational associations between variables of interest, augment logical methods of reasoning, and support empirical observations and testing (Pennington and Hastie, 1998). Although, within the literature, there have been discoveries that challenge the effectiveness of SATs and their benefit to improve decision-making competency (Chang et al., 2018). Nevertheless, the implementation of SATs has gained interest across a multitude of domains. More recently, SATs has become an area of interest for strategic military operations in an effort to support warfighters decision-making and planning capabilities (Borg, 2017). Heuer, a retired CIA agent, discovered through his operational and analytical experience that the Analysis of Competing Hypotheses (ACH) could be leveraged as a cornerstone for intelligence assessments and reduce confirmation bias (Heuer, 1999).

ACH is a process where analysts can identify a set of hypotheses for a given problem statement, evaluate each hypothesis by developing consistent and inconsistent evidence, reassess the hypotheses based solely on the evidence obtained, and reject of accept the hypothesis (Dhami et al., 2019). Although incorporating ACH in a military setting has been shown to improve the current analytic process of intelligence and counterintelligence assessments, there are limitations associated with ACH. One of the biggest limitation is that analysts generally do not obtain situational awareness across the Joint All-Domain Command and Control (JADC2) environment which significantly hinders their ability to develop adequate hypotheses to support the question at hand. Another limitation, with respect to the ISR domain, is that the development of hypotheses does not support the breakdown of high-level Priority Intelligence Requirements (PIRs) into Essential Elements of Information (EEI) and further into smaller subcomponents. For these reasons, our research objective was to determine if a specific SAT or a hybrid-approach could enhance the detection of EEIs relating to a high-level question resulting in improved performance accuracy.

Finding the appropriate SAT that specifically supports ISR analysts breakdown PIRs into EEIs and its subcomponents is no easy task. However, one technique within the literature stood out more than others. Scaffolding encourages operators to divide a given task into manageable and achievable subtasks (Sidky, 2019). This is accomplished by partitioning the task into a problem statement, generating and selecting possible solutions, justifying the selection, and testing and evaluating the results which is comparable to the ACH approach (Xun and Land, 2004). Scaffolding has been shown to enhance an operators' performance when apportioned ill-defined and complex analytic problem sets, similar to those ISR analysts must evaluate and decipher, by inciting the operator to create theoretical outcomes based upon a solution framework (Stender and Kaiser, 2015). This strategy augments critical thinking, inductive reasoning, and metacognition which may translate into ISR operations when considering a problem statement, generating EEIs, justifying the reasoning, and providing solution sets with associated confidence ratings (Warli et al., 2021). Based on this information from the literature, Scaffolding appears to be an appropriate SAT to support ISR analysts identify EEIs leading to greater efficacy in prediction assessments.

Although there is strong evidence that Scaffolding appears to be the best 'fit' for ISR analysts focusing on EEI identification for prediction assessments, developing a unique hybrid-SAT may provide the greatest performance enhancement. Therefore, ISR subject matter expert (SMEs) developed the Method for Defining Analytical Questions (MDAQ). MDAQ is depicted as a structured approach to assess, process, hypothesize, and deliver a decision with high confidence. MDAQ was formulated based on ISR SMEs experience and knowledge, ACH, and Scaffolding with the premise of gather EEIs within a problem set guided by key indicators. MDAQ is a continual process allowing the operator to deep-dive into the problem statement and produce several EEIs and indicators.

Information Workflow: How Information Is Perceived

It has been widely publicized that SATs have been shown to improve cognitive processes and decision-making competency when evaluating complex problem statements. However, another issue that ISR analysts face is how collected intelligence is obtained. When information is obtained all at once, it can be extremely challenging for an operator to remain vigilant and focused while interpreting and analyzing the content. Obtaining all the information at once can result in cognitive impediment which occurs when an operator reaches their information throughput capacity also known as information overload (Nelson et al., 2015). At this point, the operator can no longer process and effectively respond to the critical information provided (Nelson et al., 2016). This can lead to EEIs being inadvertently overlooked and poor decision-making recommendations. On the other hand, when information is obtained in incremental sections, operators become more focused on parsing and detecting subtle descriptors and indicators (Ade and Deshmukh, 2013). Therefore, providing content incrementally has the opportunity to improve information throughput, attention to detail, and EEI detection.

The psychological process of sorting through and identifying critical components of collected intelligence within a military setting has similar characteristics to a vague or ambiguous narrative where there is missing or incomplete information. In both instances, an operator will subconsciously follow the four phases of the decision-making process (i.e., intelligence phase, design phase, choice phase, and implementation phase) (Chiheb et al., 2019). For a military setting, these phases correlate to the operator determining the EEI, identifying indicators and observables to support the EEI, and providing an educated recommendation and justification for future military direction with high confidence to their leadership. For an ambiguous narrative, these phases correlate to defining the problem statement, collecting and identifying supporting components to the problem statement, and providing a solution. Based on the review of the literature, there is a knowledge gap in understanding if SATs, based upon information workflow methodologies, can enhance detection of EEIs resulting in improved performance. Therefore, the objective of this study was to evaluate the effectiveness of SATs on performance accuracy when providing an ambiguous narrative in incremental or complete sections.

METHODS AND MATERIALS

Study Participants

In order to adequately evaluate SATs and information workflow, an appropriate sample size must be represented. A power analysis was conducted to determine the sample size for the study based on a desired significant level, effect size, and statistical power. Previous literature has discovered a significant performance improvement when implementing a Scaffolding approach during cognitive assessments (Belland et al., 2007). In this study, a pairwise comparison of 0.27 for pooled standard deviation of subjects and a mean difference of 0.23 was identified in the analysis. Using a standard power of 0.80, a sample calculation was performed resulting in a sample size of 25.

Participants were recruited from Amazon's Mechanical Turk (Mturk) which is a crowdsourcing platform where surveys and research studies can be hosted and completed for human intelligence tasks. Participants were required to be 18 years old or older, speak fluent English, be located within the United States, possess basic computer skills, have no prior experience with SAT research studies, and obtain a minimum of a 95% accuracy rate on Mturk. As well, additional demographic and screening questionnaires were

provided following the study to gather insight for statistical and correlation analysis with respect to age, gender, education, and video game experience.

Experimental Design

A between-subjects experimental design was selected in order to mitigate learning effects from the task. Based on the power analysis of 25 participants per condition, a total of 150 participants were randomly assigned to one of three SATs (Scaffolding, MDAQ, Control) and obtained the information all at once or in incremental sections (i.e., all content divided into 5 subsections) as shown below:

- Scaffolding with Incremental Sections (Scaffolding_{INC})
- Scaffolding with Complete Sections (Scaffolding_{ALL})
- MDAQ with Incremental Sections (MDAQ_{INC})
- MDAQ with Complete Sections (MDAQ_{ALL})
- Control with Incremental Sections (Control_{INC})
- Control with Complete Sections (Control_{ALL})

Each of the SATs followed a specific structured approach with the premise of enhancing cognitive decision-making capabilities resulting in higher solution accuracy. Participants assigned to the Scaffolding group were required to read through the content and develop a problem statement, generate solutions, provide justification and reasoning, and select the most promising solution to the problem statement. Participants assigned to the MDAQ group were required to read through the content and develop an indicator and/or observable to support their assessment. Indicators and observables are bits of information that provide insight and relational associations to a person, place, or event. The development of indicators and observables can strengthen the justification and rational of the participants final decision. MDAQ is a continual process where participants have the ability to develop and associate several indicators and observables until they suspect they have identified all the critical bits of information. Lastly, if participants were assigned to the Control group, they were required to read through the content and provide a single solution (see Figure 1).



Figure 1: Schematic information workflow process for each SAT.

In addition to the assigned SAT conditions, participants were coupled with two information workflow methodologies (i.e., all or incremental sections). If participants were given all the content all at once, they would read the information and follow the structured approach assigned (i.e., Scaffolding, MDAQ, Control). Following this process, the participants would provide a single solution. If participants were given the content in incremental sections, they would read through the first section and follow the structured approach assigned. Following the first increment, the participants would provide a single solution. This process was repeated for each of the five sections. The purpose of the incremental sections was to enhance the detection of subtle indicators and observables to support EEI identification and solution accuracy.

Performance Task: ISR Adjacent

The task was selected and reformatted from an online mystery-solving game 'The Haunted Portrait' (5minutemystery.com). The task provides an opportunity for online participants to engage in an immersive storyline that requires critical thinking to detect and decipher subtle clues such as EEIs and indicators. The participants' objective was to provide the correct solution to the storyline. Within the story, there are main characters. For each main character, there was only one EEI/indicator. Therefore, depending on the users' analytic approach, they could easily become overwhelmed or distracted resulting in overlooking these critical clues. The narrative was selected and supported by our in-house ISR SMEs as being similar to actual ISR analysts work in regards to identifying information as to be used as a proxy. The narrative displays similar characteristics that are essential to complex ISR mission environments such as critical thinking, future decision-making, logical reasoning, and problem solving.

Data Analysis

Statistical analysis was conducted using R Statistical Analysis software (R version 4.1.2). An analysis of variance (ANOVA) was performed evaluating performance based on (1) Structured Analytic Techniques and information workflow methodologies (2) Structured Analytic Techniques and incremental EEI. A pairwise comparison of information workflow was performed for each SAT to evaluate response time using two-tailed two-sample t-tests.

RESULTS

The results for the study were analyzed focusing on three sections: performance accuracy based on Structured Analytic Techniques and information workflow methodologies, performance accuracy based on Structured Analytic Techniques and incremental EEI, and pairwise comparison of information workflow for each Structured Analytic Technique in relation to response time.

First, we will discuss the findings for performance accuracy based on SATs and information workflow methodologies. As shown in Figure 2, there was

a statistically significant difference in performance accuracy for MDAQ_{INC} compared to all other conditions ($F_{1,48}$ =4.11, p<0.05). MDAQ_{INC} resulted in 68% solution accuracy (17/25), Control_{INC} resulted in 40% solution accuracy (10/25), Scaffolding_{INC} and MDAQ_{ALL} resulted in 28% solution accuracy (7/25), Scaffolding_{ALL} resulted in 24% solution accuracy (6/25), and Control_{ALL} resulted in 16% solution accuracy (4/25) (see Table 1).

With respect to confidence, there was a statistically significant difference between conditions ($F_{5,144}$ =5.20, p<0.01). Scaffolding_{INC} and MDAQ_{ALL} displayed the highest confidence in their prediction solution (3.7 out of 5.0). This was followed by Control_{ALL} (3.2 out of 5.0), MDAQ_{INC} and Scaffolding_{ALL}.

ANOVA comparing Analytic Technique and Information Workflow						
Condition Comparison	Performance	F	р			
Control _{ALL} vs MDAQ _{All}	4(16%) vs 7(28%)	1.03	0.32			
Control _{ALL} vs Scaffolding _{ALL}	4(16%) vs 6(24%)	0.48	0.49			
Control _{ALL} vs Control _{INC}	4(16%) vs 10(40%)	3.69	0.06			
Control _{ALL} vs MDAQ _{INC}	4(16%) vs 17(68%)	18.43	< 0.01			
Control _{ALL} vs Scaffolding _{INC}	4(16%) vs 7(28%)	1.03	0.32			
MDAQ _{ALL} vs Scaffolding _{ALL}	7(28%) vs 6(24%)	0.10	0.75			
MDAQ _{ALL} vs Control _{INC}	7(28%) vs 10(40%)	0.78	0.38			
MDAQ _{ALL} vs MDAQ _{INC}	7(28%) vs 17(68%)	9.16	< 0.01			
MDAQ _{ALL} vs Scaffolding _{INC}	7(28%) vs 7(28%)	0.00	1.00			
Scaffolding _{ALL} vs Control _{INC}	6(24%) vs 10(40%)	1.45	0.23			
Scaffolding _{ALL} vs MDAQ _{INC}	6(24%) vs 17(68%)	11.62	< 0.01			
Scaffolding _{ALL} vs Scaffolding _{INC}	6(24%) vs 7(28%)	0.10	0.75			
Control _{INC} vs MDAQ _{INC}	10(40%) vs 17(68%)	4.11	0.05			
Control _{INC} vs Scaffolding _{INC}	10(40%) vs 7(28%)	0.78	0.38			
MDAQ _{INC} vs Scaffolding _{INC}	17(68%) vs 7(28%)	9.16	< 0.01			

 Table 1. ANOVA depicting correct responses for SATs and information workflow methodologies. Statistical significance alpha level of 0.05.

An Analysis of Variance (ANOVA) was conducted comparing Structured Analytic Techniques (SATs) and Information Workflow for Solution Accuracy. Statistical Significance at an alpha level of 0.05.

(3.0 out of 5.0), and Control_{INC} (2.6 out of 5.0). It is important to note that the confidence scale was from 0 to 5 where 0 represents low confidence and 5 represents high confidence.

Next, we will discuss the findings for performance accuracy based on SATs and incremental EEI. Increment 1 displayed a statistically significant difference in performance accuracy with 8 out of the 25 participants providing the correct solution for MDAQ compared to 1 out of 25 participants providing the correct solution for Control ($F_{1,48}=7.4$, p<0.01). Increment 2 displayed a statistically significant difference in performance accuracy with 15 out of the 25 participants providing the correct solution for MDAQ compared to 5 out of 25 participants providing the correct solution for MDAQ compared to 5 out of 25 participants providing the correct solution for Scaffolding ($F_{1,48}=9.6$, p<0.01). Increment 3 displayed a statistically significant difference in performance accuracy with 16 out of the 25 participants providing the correct solution for Scaffolding ($F_{1,48}=9.6$, p<0.01). Increment 3 displayed a statistically significant difference in performance accuracy with 16 out of the 25 participants providing the correct solution for Scaffolding ($F_{1,48}=9.6$, p<0.01). Increment 3 displayed a statistically significant difference in performance accuracy with 16 out of the 25 participants providing the correct

solution for MDAQ compared to 7 out of 25 participants providing the correct solution for Scaffolding ($F_{1,48}=7.2$, p<0.01). Increment 4 displayed a statistically significant difference in performance accuracy with 14 of the 25 participants providing the correct solution for MDAQ compared to 7 out of the 25 participants providing the correct solution for Scaffolding ($F_{1,48}7.2$, p<0.01). Increment 5 displayed a statistically significant difference in performance accuracy with 17 out of the 25 participants providing the correct solution for Scaffolding the correct solution for MDAQ compared to 7 out of the 25 participants providing the correct solution for MDAQ compared to 7 out of the 25 participants providing the correct solution for MDAQ compared to 7 out of the 25 participants providing the correct solution for MDAQ compared to 7 out of the 25 participants providing the correct solution for MDAQ compared to 7 out of the 25 participants providing the correct solution for MDAQ compared to 7 out of the 25 participants providing the correct solution for Scaffolding ($F_{1,48}=9.2$, p<0.01) (see Table 2). It is important to note that both groups (i.e., all, incremental) received the same content, however the incremental group received the content in 5 subsections.



Figure 2: Correct responses for SATs and information workflow methodologies. Statistical significance alpha level of 0.05.

With respect to confidence, there was a statistically significant difference between MDAQ compared to Control ($F_{1,48}$ =4.3, p = 0.04) and MDAQ compared to Scaffolding ($F_{1,48}$ =4.2, p = 0.05) for Increment 2. There was a statistically significant difference between MDAQ compared to Control ($F_{1,48}$ =4.3, p = 0.04) and MDAQ compared to Scaffolding ($F_{1,48}$ =6.2, p = 0.02) for Increment 3. There was a statistically significant difference between MDAQ compared to Control ($F_{1,48}$ =7.3, p = 0.01) and MDAQ compared to Scaffolding ($F_{1,48}$ =4.0, p = 0.05) for Increment 4. There was a statistically significant difference between MDAQ compared to Control ($F_{1,48}$ =13.1, p<0.01) and MDAQ compared to Scaffolding ($F_{1,48}$ =5.0, p = 0.03) for Increment 5. There was not a significant difference in confidence ratings between groups for Increment 1.

Lastly, we will discuss the findings for response time based on Structured Analytic Techniques and information workflow methodologies. An analysis of variance (ANOVA) was conducted and displayed a statistically significant difference for response times between condition ($F_{5,144}=7.6$,

Table 2. ANOVA	comparing	SATs	by	increments	for	solution	accuracy.	Statistical
significa	nce alpha le	evel of (0.05	5.				

ANOVA comparing Analytic Technique by Increments							
Increment 1							
Condition Comparison	Performance	F	р				
Control vs MDÂQ	1(4%) vs 8(32%)	7.35	0.01				
Control vs Scaffolding	1(4%) vs 4(16%)	2.00	0.16				
MDAQ vs Scaffolding	8(32%) vs 4(16%)	1.75	0.19				
Increment 2							
Condition Comparison	Performance	F	р				
Control vs MDAQ	9(36%) vs 15(60%)	2.94	0.09				
Control vs Scaffolding	9(36%) vs 5(20%)	1.57	0.22				
MDAQ vs Scaffolding	15(60%) vs 5(20%)	9.60	< 0.01				
Increment 3							
Condition Comparison	Performance	F	р				
Control vs MDAQ	12(48%) vs 16(64%)	1.28	0.26				
Control vs Scaffolding	12(48%) vs 7(28%)	2.13	0.15				
MDAQ vs Scaffolding	16(64%) vs 7(28%)	7.20	0.01				
Increment 4							
Condition Comparison	Performance	F	р				
Control vs MDAQ	9(36%) vs 14(56%)	2.01	0.16				
Control vs Scaffolding	9(36%) vs 7(28%)	0.35	0.55				
MDAQ vs Scaffolding	14(56%) vs 7(28%)	4.20	0.05				
Increment 5							
Condition Comparison	Performance	F	р				
Control vs MDAQ	10(40%) vs 17(68%)	4.11	0.05				
Control vs Scaffolding	10(40%) vs 7(28%)	0.78	0.38				
MDAQ vs Scaffolding	17(68%) vs 7(28%)	9.16	< 0.01				

An Analysis of Variance (ANOVA) was conducted comparing Structured Analytic Techniques (SATs) by Increments for Solution Accuracy. Statistical Significance at an alpha level of 0.05.

p<0.01). As a result, a two-tailed two-sample t-test was performed comparing Structured Analytic Techniques and information workflow for a pairwise comparison. Control_{ALL} was statistically significant compared to MDAQ_{INC} (F_{1,48}=16.2, p<0.01), MDAQ_{ALL} (F_{1,48}=24.1, p<0.01), Scaffolding_{INC} $(F_{1,48}=23.1, p.01)$, and Control_{INC} $(F_{1,48}=5.9, p = 0.02)$ for response time. Scaffolding_{ALL} was statistically significant compared to MDAQ_{INC} (F_{1,48}=9.3, p<0.01), MDAQ_{ALL} (F_{1,48}=9.6, p<0.01), and Scaffolding_{INC} $(F_{1,48}=9.0, p<0.01)$ for response time. Control_{INC} was statistically significant compared to MDAQ_{INC} (F_{1,48}=7.0, p<0.01), MDAQ_{ALL} (F_{1,48}=5.3, p = 0,02), and Scaffolding_{INC} (F_{1,48}=4.9, p = 0.03). Moreover, the Control_{ALL} condition spent the least amount of time on the task at approximately 14.5 minutes followed by Scaffolding_{ALL} at approximately 19.5 minutes, Control_{INC} at approximately 21.5 minutes, Scaffolding_{INC} at approximately 29 minutes, MDAQ_{ALL} at approximately 30 minutes, and MDAQ_{INC} at approximately 36.5 minutes. Although there was no statically significant difference, almost every condition showed shorter response time when the participants' prediction was incorrect compared to when their prediction was correct. Therese factors are important to take into consideration when involving real-world ISR situations because the decisions and recommendations may have serious consequences on future military direction. In addition, there was no statistically significant difference with respect to gender, age, education, and video game experience between Structured Analytic Techniques and solution accuracy.

CONCLUSION

Intelligence, Surveillance, and Reconnaissance (ISR) operators are being continually tasked to identify and evaluate key indicators and Essential Elements of Information (EEIs) to support leadership-driven Priority Intelligence Requirements (PIRs). These EEIs support operational decision-making capabilities by providing the commander with supportive information in order to make a high confident decision in a timely manner. As you could image, ISR operators are under a tremendous amount of stress to filter through vague datasets and ensure accurate and relational intel is being disseminated to leadership.

This research study focused on probing SATs coupled with information workflow methodologies to enhance rational and analytical thinking when providing a vague narrative. The findings provided underlying evidence that MDAQ coupled with incremental sections of information could significantly improve detection of EEIs leading to improved solution accuracy. Moreover, previous literature has shown that the Scaffolding approach can enhance the decision-making process resulting in improved decision accuracy. Yet, we did not observe these enhancements in this study. The reason may be that the information provided in the storyline was extremely vague and ill-defined in order to replicate tasks similar to those an ISR analyst may observe. Future research should include multiple dialogues with varying readability and comprehension difficulty levels to get a better understanding on the efficacy of SATs across complex tasks.

Nevertheless, we are advancing in the right direction in an effort to support and improve operational training and education, structured analytic techniques and methodologies, and accurately detecting EEIs in situations related to ISR processing, exploitation, and dissemination (PED). It is vital that we continue to refine and develop tools that can support our analysts and optimize performance in order to protect national security.

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