

The role of artificial intelligence (AI) in improving technical and managerial cybersecurity tasks' efficiency

Managerial
cybersecurity
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efficiency

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Abstract

Purpose – Artificial intelligence (AI) can assist in the worldwide shortage of cybersecurity workers in technical and managerial roles. Thus, the purpose of this study was to investigate the role of AI in automating many of the routine tasks associated with cybersecurity. As such, AI enables cybersecurity personnel to reduce their workloads and focus on more strategic aspects of their work.

Design/methodology/approach – This study is an exploratory field study. The authors started by conducting a literature review to assess the possibility that AI tools can provide and how they can improve cybersecurity efficacy. Following this, the authors identified the specific core tasks for two cybersecurity work roles (technical and managerial) and searched for specific commercial tools that can perform each of the tasks. Then, the authors used the free ChatGPT 3.5 to list the current cybersecurity systems that use AI for the associated tasks, which the authors then reviewed with the tools' documentation and websites to confirm these tasks were conducted or assisted by AI.

Findings – Results indicated that all 14 cybersecurity tasks of the technical work role are currently noted to be performed by commercial cybersecurity systems with AI-integrated capabilities, while only 11 of the 17 managerial work role tasks currently appear to be performed by AI.

Practical implications – The rapid integration of AI capabilities into commercial cybersecurity systems may suggest that the cybersecurity workforce must be currently trained on how to use AI tools in their daily operations, especially as it pertains to technical cybersecurity work roles.

Social implications – The cybersecurity workforce shortage is reported to exceed four million cybersecurity workers worldwide in 2023. Thus, further understanding of the role of AI in improving the efficiency of technical and managerial cybersecurity tasks is significant.

Originality/value – The value of this research lies in the initial assessment of the current AI capabilities of commercial cybersecurity systems, which will ultimately provide the “super-human” performances resulting from human-AI teaming.

Keywords Cybersecurity tasks efficiency, Artificial Intelligence (AI) for cybersecurity, Cybersecurity workforce productivity, Cybersecurity management

Paper type Research paper



1. Introduction

There is an estimated 625,000 cybersecurity employee shortage in the United States (U.S.) (Cyberseek.org, 2023). Moreover, it is estimated that around 200,000 to 250,000 additional

cybersecurity vacancies are also available at the U.S. Department of Defense (DoD) (DCWF, 2023). Furthermore, according to the Cybersecurity Workforce study of the International Systems Security Certification Consortium, in 2023, there is a shortage of about four million cybersecurity workers worldwide, which increased by 12.6% from the previous year (ISC2, 2023). The recent Whitehouse (2023) National Cyber Workforce and Education Strategy called for significant investments in national programs to increase the production of qualified cybersecurity employees to fill these vacancies. However, such an approach appears to be based on current technologies rather than focusing efforts on research and development to automate many of the current cybersecurity tasks using advanced technologies such as artificial intelligence (AI). In a keynote provided at the Knowledge Management Conference in Geneva, Switzerland, Dr Brian Buckles (2023) stated, “When it comes to cybersecurity workforce shortage, instead of asking for more horses and faster horses, we should ask for a John Deere.”

History has provided us with an interesting perspective on how technological advances can greatly enhance productivity. Such technological development takes time and effort. Over the past two centuries, we experienced shorter and shorter times between manual, time-consuming tasks to technologies automating these tasks while doing them faster, cheaper and more effective. This concept can be demonstrated, for example, by agricultural innovations developed by John Deere, such as introducing their tractors to replace manual and horse-powered agricultural processes in cotton picking. Till the mid-19th century, about two-thirds of all worldwide cotton production was produced in the U.S. totally manually. From the early 1920s, John Deere, the U.S. agriculture technology company, has been working on a design of a multirow cotton-picking combine that can collect cotton 525 times faster (or 52,500% more effectively) than handpicking (Bennett *et al.*, 2015; Deere.com, 2023). It took John Deere about 25 years to enhance their combine, whereas in the 1960s, an estimated 96% of all cotton picking in the U.S. was done by machine (Businessmodulehub.com, 2020). According to Bennett *et al.* (2015), “the time required to pick a bale of cotton . . . has decreased from approximately 50 to 70 man-hours [hand-picking] down to eight minutes [machine-based picking]” (p. 226). This type of increased efficiency and time savings is needed in the cybersecurity arena to offset labor shortages.

In 2008, The National Initiative of Cybersecurity Education (NICE) Framework, also known as the Workforce Framework for Cybersecurity, was originally created to help both industry and government to synchronize their cybersecurity workforce by providing a common list of cybersecurity knowledge, skills and abilities that are needed to perform a specific list of cybersecurity tasks associated with the defined work roles. In early 2023, the DoD published the DoD Cyber Workforce Framework (DCWF, 2023), including the 52 NICE Framework work roles and 23 additional work roles in cybersecurity and other cyber-related domains such as software engineering, data analytics and AI. Additionally, the U.S. National Security Agency (NSA) has recognized that the field of cybersecurity can be categorized by both technical and managerial work roles (NSA, 2022). Moreover, according to CyberSeek.org (2023), as of October 2023, there were over 88,000 job openings for the Information Systems Security Manager (OPM ID: 722; NICE ID: OV-MGT-001) and Cyber Defense Analyst (OPM ID: 511; NICE ID: PR-CDA-001), representing about 13.3% of all cybersecurity job openings in the U.S. combined. With the rise and wide use of generative AI, along with technological innovation in the integration of AI into the workforce, especially commercial cybersecurity tools, our initial assumption was that most, if not all, of the core cybersecurity tasks for the two aforementioned work roles, could be automated by AI. Our initial assumption was based on the significant technological innovation that Generative AI such as Chat Generative Pretrained Transformer (ChatGPT) by OpenAITM

and Google Bard™ have demonstrated in recent months, along with the indications provided by many commercial cybersecurity tools that started advertising the integration of Generative AI into their software to aid cybersecurity professionals improved security outcomes, enhanced work role task efficiency and ultimately provide organizational cost-effectiveness. Thus, in this research study, we have started by reviewing prior scientific literature about AI in the context of these two cybersecurity work roles, followed by using the free version of ChatGPT (Ver. 3.5) to assist us with the initial identification of as many commercially available cybersecurity tools as possible that claim to use AI to automate each of the cybersecurity tasks associated with the two work roles. We aimed to provide a current assessment of the role of AI in improving cybersecurity tasks efficiency and to help close the cybersecurity workforce gap. Specifically, this research addressed the following research questions:

- RQ1. What key technical and managerial cybersecurity core work role tasks can be automated by AI?
- RQ2. What commercial AI vendors, platforms or tools are currently available to automate key technical and managerial cybersecurity core work role tasks?

2. Background

2.1 Cybersecurity and artificial intelligence

Cybersecurity is defined as a “computing-based discipline involving technology, people, information, and processes to enable assured operations. It involves the creation, operation, analysis, and testing of secure computer systems” (CSEC, 2017, p. 16). Cybersecurity also aims to reduce the risks of cyber threats, including malware, phishing, carrier denial attacks, social engineering attacks and man-in-the-middle attacks (Funk, 2022). Cybersecurity and AI must work together to increase cybersecurity awareness within organizations. While cybersecurity threats faced by global businesses are rising in number and sophistication, integrating AI into cybersecurity systems can help reduce such challenges and the burden on cybersecurity personnel (Funk, 2022; Kadel *et al.*, 2022). AI-based approaches for cybersecurity applications appear to be already applied to various tasks, such as access control or user authentication, flexible and continuous network situation awareness, dangerous behavior monitoring, detection of misleading information, data loss prevention, fraud detection, abnormal traffic identification, risk and compliance management, to name a few (Kadel *et al.*, 2022; Krishnappa, 2023; Zhang *et al.*, 2022). According to Kadel *et al.* (2022), businesses use AI to automate cyber protections against different kinds of threats, such as spam and phishing, and to identify malware, fraudulent payments and compromised computers, as well as in forensics investigations. AI's strength is its capability to process and analyze huge quantities of data efficiently, automatically classify it, recognize meaningful patterns and eliminate irrelevant noise (Krishnappa, 2023; Kshetri, 2021; Zeadally *et al.*, 2020; Zhang *et al.*, 2022). By implementing AI, cybersecurity specialists are enabled to respond quickly and mitigate potential threats (Funk, 2022). AI approaches can enhance the performance of conventional slow or insufficient cybersecurity systems to protect efficiently against an increasing range of intricate cyber threats (Kadel *et al.*, 2022). One of the major benefits of AI is that it can learn from experiences and recognize malware even if it was never seen before, enabling proactive protection against potential breaches, even before vulnerabilities are disclosed and rectified to the public, thus stopping them before causing damage. As a result, AI can help address and reduce serious threats such as zero-day vulnerabilities (Kshetri, 2021; Kumar *et al.*, 2023; Tojiboyev and Safoev, 2023).

Furthermore, according to [Krishnappa \(2023\)](#), by examining network traffic and evaluating the oddity of accessed data packets, AI can help identify patterns not easily discovered by humans, such as a potential cyber-attack. AI systems ascertain rules through their training or operation rather than leaning on subject matter experts to devise them, thereby necessitating engineers and programmers to encode the rules into the system ([Kroll et al., 2021](#)). Various AI tools can be used to identify and mitigate cyber threats, such as artificial neural networks, expert systems, intelligent agents, machine learning (ML), bio-inspired computing, deep learning and more ([Kadel et al., 2022](#); [Krishnappa, 2023](#)). These tools use different types of algorithms or statistical methods to perform their mission, such as decision trees, k-nearest-neighbor algorithms, support vector machines; self-organizing maps and naïve Bayes algorithms ([Zeadally et al., 2020](#)). ChatGPT is an advanced AI language model that interacts and communicates with users and generates human-like responses. According to [Jeyaraman et al. \(2023\)](#), ChatGPT can be used as a security assistant to investigate and develop security solutions, penetration testing, identify vulnerabilities and correct bugs. Moreover, [Al-Hawawreh et al. \(2023\)](#) claim that ChatGPT could be used as a brainstorming tool to generate new ideas related to cybersecurity policy reports, and [Ayinde et al. \(2023\)](#) include ChatGPT's capabilities to summarize text, answer questions and to create unbiased automated decision-making systems.

2.2 Artificial intelligence-enabled cybersecurity efficiency

Applying AI can help organizations' management to make better decisions about their business. By automating processes using AI technologies, organizations may save time and substantially boost efficiency while discovering more faults than they could manually, defending against attacks, remediating vulnerabilities that are not even identified or fixed ([Krishnappa, 2023](#); [Kumar et al., 2023](#)), assessing customer threats and deciding on customers' ongoing control ([Funk, 2022](#)). AI is better and faster than humans to perform repetitive and routine cybersecurity procedures that might fatigue the cybersecurity team ([Funk, 2022](#); [Kshetri, 2021](#)) or those requiring collecting and processing large amounts of information, resulting in superior outcomes ([Funk, 2022](#); [Kadel et al., 2022](#); [Krishnappa, 2023](#); [Kshetri, 2021](#); [Kumar et al., 2023](#)). The key determinants of AI's efficiency are the quality and quantity of the data the system was trained on ([Kshetri, 2021](#)). AI diminishes the cost of breach detection and response, saving between 1% and 15%, with an average of 12% ([Kadel et al., 2022](#)). AI thoroughly examines the organization's network to look for any cybersecurity breaches with greater speed and accuracy than human beings, halting cyber-attacks faster ([Kshetri, 2021](#)). Signature codes must be compared immediately to identify the cyberattack. Using an AI system, these are done faster, avoiding significant failures and losses ([Tojiboyev and Safoev, 2023](#)). ML analyzes existing data and improves over time. Its ability to recognize and detect very small changes from that pattern and to apply them immediately improves its methods and practices continuously ([Kadel et al., 2022](#); [Zeadally et al., 2020](#)). Real-time traffic monitoring enables AI technology to immediately detect and act on any unauthorized activity ([Tojiboyev and Safoev, 2023](#)). ChatGPT processes natural language prompts and generates responses customized according to the user's requirements and circumstances ([Sharma and Dash, 2023](#)). It can be used for searching for diverse, complex and flexible solutions more efficiently and effectively, automatically creating procedures and documents based on common knowledge ([Korzynski et al., 2023](#)). AI's efficacy, speed and accuracy allow cybersecurity teams to do their work faster and more effectively instead of spending significant time on manual detection, examining alerts and determining if they are benign or malicious to define their reaction ([Kadel et al., 2022](#)). AI optimizes cybersecurity tasks and frees up valuable resources for the organization ([Kadel](#)

et al., 2022; Krishnappa, 2023; Kumar *et al.*, 2023). By reducing the time and effort dedicated to policy creation and topography understanding, AI enables cybersecurity teams to focus more on strategic aspects of cybersecurity. This leads to a more robust cybersecurity posture, better threat identification and enhanced protection against cyber threats (Kumar *et al.*, 2023). AI has the potential to support cybersecurity professionals by identifying high-priority areas that need more attention and automating certain tasks that are currently carried out manually, thus reducing their workload and allowing them to focus on more complex and critical tasks. This is especially important when considering the current shortage of cybersecurity professionals worldwide (ISC2, 2023).

2.3 *The artificial intelligence cybersecurity workforce revolution*

According to the workforce study (ISC2, 2023), while the demand for AI skills is steadily increasing, it is not yet considered as one of the most important requirements for hiring. Skills related to AI are now considered one of the top five categories in terms of in-demand skills. This marks a significant change from the previous year when they were not as highly sought after. As AI technology continues to evolve and impact various areas of cybersecurity threats and defense, the demand for these skills is likely to rise even further in the coming years (ISC2, 2023). AI integration in cybersecurity has displaced workers, as computers' increased efficiency makes them preferable to human experts. Organizations adopting AI in their infrastructures experience increased efficiency and enhanced data security as the technology continuously learns (Tojiboyev and Safoev, 2023). Novel tools and analytical methods simplify the process of distinguishing unusual activities from the typical background. Automated tools can sift through system alerts, forwarding the most critical or actionable items to human operators. These tools realign human tasks, eliminating monotonous incident management and system-state tracking tasks and allowing more time for tool development and other responsibilities (Kroll *et al.*, 2021). AI implementations should be regarded as complex sociotechnical systems rather than simple technical tools. When evaluating their actions, it is essential to consider the human factors, policies and interactions that constitute a larger framework and position specific tools within that environment. Automated systems can exhibit high autonomy or work closely with humans. Certain automated systems are designed to assist human decision-making, like systems prioritizing detected alerts for a Security Operations Center. In contrast, some function independently and make decisions without human intervention, such as firewalls or spam filters (Kroll *et al.*, 2021). According to recent research, generative pretrained transformers (GPT) models are likely to have an impact on the work tasks of around 80% of workers (Ritala *et al.*, 2023). As automation takes over various tasks, there is a possibility that the operators may lose their familiarity with the intricate details of the process. They may become less aware of the specifics of their actions and less competent in understanding or supervising the internal functioning of the automation. This situation can lead to concerns related to the dependability and trustworthiness of the automated process (Kroll *et al.*, 2021). AI exhibits rapid data processing capabilities and excels in recognizing specific scenarios. However, it can be susceptible to disturbances and might not always provide accurate or ethical assessments of novel situations. In contrast, humans are more adaptable and can make swift judgments when confronted with new developments in a network, although they also benefit from machine support to play an auxiliary role (Kroll *et al.*, 2021; Zhang *et al.*, 2022). AI aims to support cybersecurity experts in this domain rather than replacing them entirely. Consequently, it remains essential for experts in the field to step in and apply their network expertise to make informed assessments of the existing network state. AI technology is becoming more significant in promoting automation in the field of

cybersecurity. By automating many of the routine tasks associated with cybersecurity, AI enables cybersecurity personnel to focus on more strategic aspects of their work.

3. Methodology

This study is categorized as a design science, also known as Design and Development research. According to [Hevner et al. \(2004\)](#), “Design science ... creates and evaluates [Information Technology] IT artifacts intended to solve identified organizational problems” (p. 77). [Ellis and Levy \(2010\)](#) indicated that design and development research is focused on developing an artifact “that can serve to strengthen the interaction in the conceptualization and evaluation cycle” (p. 108). They further indicated that developing “new methods and processes for implementing existing models or using existing tools” (p. 108) constitutes a viable design science artifact. The artifact proposed in this study is the categorization of commercial AI vendors, platforms or tools (*RQ2*) that are currently available to automate cybersecurity core work role tasks (*RQ1*).

To answer the research questions proposed, there is a need to understand the various tasks the cybersecurity workforce copes with. There are two main types of roles in the cybersecurity workforce, technical and managerial roles, which differ in the tasks they perform ([NSA, 2022](#)). Therefore, to address *RQ1*, the first step for this study was to identify the roles and specific core tasks for each work role category. Given the magnitude of over 1,000 cybersecurity tasks identified across the 52 work roles in the NICE Framework ([NICE Framework Reference Spreadsheet, 2017](#)) and 75 work roles in the DCWF ([2023](#)) compounded by the significant cybersecurity workforce shortage ([ISC2, 2023](#)), our focus in this study was on the core tasks that are specifically related to two main work roles within the cybersecurity domain. The first is from the managerial cybersecurity aspect, and the second is from the technical cybersecurity aspect. These two work roles are Information Systems Security Manager (OPM ID: 722; NICE ID: OV-MGT-001) and Cyber Defense Analyst (OPM ID: 511; NICE ID: PR-CDA-001), which are critical for any organization. We have identified and used these two work roles as examples to conduct the investigation. The associated core cybersecurity tasks were extracted from the [NICE Framework \(2017\)](#) and the [DCWF \(2023\)](#) to serve as the foundation for addressing *RQ1*. We have assumed at this point that all core cybersecurity tasks for both the technical and managerial work roles can be automated by AI.

The next step in the study addressed *RQ2*, which focused on uncovering the existing commercial AI vendors, platforms or tools that can assist in each technical and managerial cybersecurity work role task. Uncovering such existing commercial AI resources and assessing their ability to use AI to either fully perform, if properly configured, or assist in performing the tasks found in the first step can significantly contribute to the current cybersecurity workforce shortage ([ISC2, 2023](#)). To do that, we have unleashed the use of the free ChatGPT version 3.5 to help us generate the data for two tables that will summarize the results of our investigation. We have prompted ChatGPT to provide responses to the specific commercial vendors, platforms or tools that use AI to conduct each core task within the work role. Each of the tasks had its own prompt, following the prompt format: “Which commercial tools use AI to [tasks]?” Because of ChatGPT’s limitations ([Jeyaraman et al., 2023](#)), each commercial tool proposed by ChatGPT was further verified through the commercial company website to confirm its authenticity. The findings of the second step will be analyzed to answer *RQ2* and define what part of the cybersecurity workforce can be empowered by AI tools. Therefore, the significant shortage of human workers in cybersecurity may be reduced, as each worker may empower herself/himself with AI to

form human-AI teams that will ultimately be able to perform the work of a multitude of current cybersecurity workers.

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4. Results

The results of the first step, identifying the specific core tasks for both work role categories, are listed in Table 1 for the technical category and Table 2 for the managerial category. Table 1 provides a list of the 14 core cybersecurity tasks from the 2017 NICE Framework in alignment with the 2023 DCWF for the technical work role of Cyber Defense Analyst (OPM ID: 511; NICE ID: PR-CDA-001). Table 2 lists the 17 core cybersecurity tasks from the 2017 NICE Framework in alignment with the 2023 DCWF for the managerial work role. The specific commercial vendors, platforms or tools that use AI to conduct each core task within the identified work role were prompted using the free version of ChatGPT 3.5, which is based on the body of knowledge update as of January 2022.

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No.	Task ID		Task description
	NICE	DCWF	
1	T0023	433	Characterize and analyze network traffic to identify anomalous activity and potential threats to network resources.
2	T0043	472	Coordinate with enterprise-wide cyber defense staff to validate network alerts.
3	T0155	723	Document and escalate incidents (including the event’s history, status and potential impact for further action) that may cause ongoing and immediate impact to the environment.
4	T0164	745	Perform cyber defense trend analysis and reporting.
5	T0166	750	Perform event correlation using information gathered from a variety of sources within the enterprise to gain situational awareness and determine the effectiveness of an observed attack.
6	T0178	767	Perform security reviews and identify security gaps in security architecture, resulting in recommendations for inclusion in the risk mitigation strategy.
7	T0198	800	Provide daily summary reports of network events and activity relevant to cyber defense practices.
8	T0214	823	Receive and analyze network alerts from various sources within the enterprise and determine possible causes of such alerts.
9	T0258	956	Provide timely detection, identification and alerting of possible attacks/intrusions, anomalous activities and misuse activities and distinguish these incidents and events from benign activities.
10	T0259	958	Use cyber defense tools for continual monitoring and analysis of system activity to identify malicious activity.
11	T0260	959	Analyze identified malicious activity to determine weaknesses exploited, exploitation methods, effects on system and information.
12	T0293	1,107	Identify and analyze anomalies in network traffic using metadata (e.g. CENTAUR).
13	T0294	1,108	Conduct research, analysis and correlation across a wide variety of all-source data sets (indications and warnings).
14	T0297	1,111	Identify applications and operating systems of a network device based on network traffic.

Source: Created by authors

Table 1.
Cybersecurity tasks
of the cyber defense
analyst (OPM ID:
511; NICE ID: PR-
CDA-001)

Table 2.
Core cybersecurity
tasks of the
information systems
security manager
(OPM ID: 722; NICE
ID: OV-MGT-001)

No.	Task ID		Task description
	NICE	DCWF	
1	T0147	705	Manage the monitoring of information security data sources to maintain organizational situational awareness.
2	T0157	730	Oversee the information security training and awareness program.
3	T0158	731A*	Participate in an information security risk assessment during the security assessment and authorization process.
4	T0159	733	Participate in the development or modification of the computer environment cybersecurity program plans and requirements.
5	T0192	790	Prepare, distribute and maintain plans, instructions, guidance and standard operating procedures concerning the security of network system(s) operations.
6	T0211	816	Provide system-related input on cybersecurity requirements to be included in statements of work and other appropriate procurement documents.
7	T0215	824	Recognize a possible security violation and take appropriate action to report the incident, as required.
8	T0219	828	Recommend resource allocations required to securely operate and maintain an organization's cybersecurity requirements.
9	T0229	852	Supervise or manage protective or corrective measures when a cybersecurity incident or vulnerability is discovered.
10	T0234	862	Track audit findings and recommendations to ensure that appropriate mitigation actions are taken.
11	T0248	919	Promote awareness of security issues among management and ensure sound security principles are reflected in the organization's vision and goals.
12	T0254	947	Oversee policy standards and implementation strategies to ensure procedures and guidelines comply with cybersecurity policies.
13	T0263	962	Identify security requirements specific to an information technology (IT) system in all phases of the system life cycle.
14	T0264	963	Ensure that plans of actions and milestones or remediation plans are in place for vulnerabilities identified during risk assessments, audits, inspections, etc.
15	T0265	964	Assure successful implementation and functionality of security requirements and appropriate information technology (IT) policies and procedures that are consistent with the organization's mission and goals.
16	T0275	1,016	Support necessary compliance activities (e.g. ensure that system security configuration guidelines are followed, compliance monitoring occurs).
17	T0280	1,032	Continuously validate the organization against policies/guidelines/procedures/regulations/laws to ensure compliance.

Notes: *Task classified as Core Tasks per NICE Framework (2017), but additional per DCWF (2023)
Source: Created by authors

The core tasks from [Tables 1](#) and [2](#) were prompted to derive the two sets of results, one for each work role. The results for the second step in the study, which addressed *RQ2*, were provided in [Table 3](#) for Cyber Defense Analyst, a technical role, and [Table 4](#) for Information Systems Security Manager, a managerial role, with the names of the relevant platforms/tools that were identified to perform these indicated tasks, given the table size with the details of each platform/tool. All the tools provided by the different vendors were annotated for each task (the columns of the table). For each vendor, the percent of tasks that have AI tools to assist was calculated (last column). For each task, the number of AI tools that are currently provided was summarized (in the last row of the table).

Cyber Defense Analyst (OPM ID: 511; NICE ID: PR-CDA-001)													
Vendor No.	Name	Core Tasks and Commercial Tools That Claim To Use AI To Do The Tasks											
		T0023	T0043	T0155	T0164	T0186	T0178	T0198	T0214	T0258	T0259	T0260	T0293
1	Darktrace	Darktrace			Darktrace	Darktrace	Darktrace Cyber AI Platform	Darktrace	Darktrace	McAfee Enterprise Security platform	McAfee MVISION EDR	Darktrace Cyber AI Platform	Darktrace Enterprise Immune System
2	McAfee				McAfee Enterprise Security Manager	McAfee Enterprise Security Manager	McAfee Vulnerability Manage	McAfee Enterprise Security Manager	McAfee Enterprise Security Manager	McAfee Network Security platform	McAfee MVISION EDR	McAfee Threat Intelligence Exchange	McAfee Enterprise Security Manager (ESM)
3	Symantec (now NortonLifeLock)	Symantec (now NortonLifeLock) Network Security			Symantec (now NortonLifeLock) Security Analytics	Symantec (now NortonLifeLock) Security Manager	Symantec (now NortonLifeLock) Security Services	Symantec (now NortonLifeLock) Security Services	Symantec (now NortonLifeLock) Security Services	Symantec (now NortonLifeLock) Security Services	Symantec (now NortonLifeLock) Endpoint Security	Symantec (now NortonLifeLock) Security Services	Symantec (now NortonLifeLock) Network Security
4	FireEye	FireEye Network Security	FireEye Helix	FireEye Helix			FireEye Helix			FireEye Network Security	FireEye Endpoint Security	FireEye Threat Intelligence + Mandiant Threat Intelligence (formerly FireEye)	FireEye Network Security
5	Fortinet		Fortinet FortiSOAR		Fortinet FortiSEM	Fortinet FortiSEM	Fortinet FortiAI	Fortinet FortiSEM	Fortinet FortiSEM	Fortinet FortiGate			
6	Cisco	Cisco Stealthwatch			Fortinet FortiSEM Cisco SecureX				Cisco SecureX		Cisco SecureX Endpoint		Cisco Stealthwatch
7	Splunk		Splunk Phantom	Splunk Phantom	Splunk Enterprise Security	Splunk Enterprise Security	Splunk Enterprise Security	Splunk Enterprise Security	Splunk Enterprise Security				Splunk Enterprise Security
8	IBM		IBM Resilient	IBM Resilient	IBM Qradar	IBM Qradar		IBM Qradar	IBM Qradar				IBM Qradar
9	Palo Alto Networks	Palo Alto Networks Cortex XDR	Palo Alto Networks (acquired by Palo Alto Networks)	Palo Alto Networks (acquired by Palo Alto Networks)	Palo Alto Networks Cortex XDR	Palo Alto Networks Cortex XDR	Palo Alto Networks Cortex XDR	Palo Alto Networks Cortex XDR	Palo Alto Networks Cortex XDR	Palo Alto Networks Cortex XDR	Palo Alto Networks Cortex XDR	Palo Alto Networks Cortex XDR	Palo Alto Networks Cortex XDR
10	Vectra	Vectra Networks							Vectra Cognito				Vectra Cognito
11	Trend Micro				Trend Micro Deep Security	Trend Micro TippingPoint			Trend Micro Deep Discovery		Trend Micro Apex One		
		AI % of total Core Tasks											
		71.4%											
		64.3%											
		64.3%											
		57.1%											
		57.1%											
		50.0%											
		50.0%											
		50.0%											
		42.9%											
		35.7%											
		28.6%											

(continued)

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Table 3. Commercial tools that use AI for the 14 core tasks associated with cyber defense analyst (OPM ID: 511; NICE ID: PR-CDA-001)

Table 3.

12	Fidelis	Fidelis Cybersecurity							Fidelis Elevate		Fidelis Elevate			25.6%
13	Sumo Logic					Sumo Logic Cloud SIEM			Sumo Logic Cloud SIEM		Sumo Logic Cloud SIEM			21.4%
14	ExtraHop	ExtraHop Reveal(x)								ExtraHop Reveal(x)	ExtraHop Reveal(x)		ExtraHop Reveal(x)	21.4%
15	Forescout				Forescout Platform								Forescout CounterACT	21.4%
16	LogRhythm					LogRhythm Security Investigation and Event Management (SIEM)								14.3%
17	SolarWinds				SolarWinds Security Event Manager									14.3%
18	AT&T					AllenVault USM (now AT&T AllenVault USM Anywhere)			AllenVault USM (now AT&T AllenVault USM Anywhere)					14.3%
19	Netkope	Netkope Security Cloud				Netkope Security Cloud								14.3%
20	Siemplify (Now Google)		Siemplify				Siemplify							14.3%
21	Cybereason		Cybereason Business Platform D3 Security				Cybereason Business Platform D3 Security							14.3%
22	D3 Security													14.3%
23	CrowdStrike									CrowdStrike Falcon	CrowdStrike Falcon			14.3%
24	Recorded Future											Recorded Future		14.3%

(continued)

Table 4.
Commercial tools that
use AI for the 17 core
tasks associated with
information systems
security manager
(OPM ID: 722; NICE
ID: OV-MGT-001)

Information Systems Security Manager (OPM ID: 722; NICE ID: OV-MGT-001)																			
Vendor No.	Core Tasks and Commercial Tools That Claim To Use AI To Do The Tasks																		
Name	T0147	T0157	T0158	T0159	T0192	T0211	T0215	T0219	T0229	T0262	T0264	T0265	T0275	T0280	AI % of Total Core Tasks				
1 IBM	IBM Qradar			IBM Qradar	Symantec (now NortonLifeLock)	Symantec Endpoint Security	Symantec Endpoint Detection and Response (EDR)	McAfee MVISION EDR	McAfee MVISION EDR						IBM OpenPages	IBM OpenPages	IBM OpenPages	IBM OpenPages	47.1%
2 Symantec	Symantec (now NortonLifeLock)			Symantec (now NortonLifeLock)	Symantec Endpoint Security	Symantec Endpoint Detection and Response (EDR)	McAfee MVISION EDR	McAfee MVISION EDR	McAfee MVISION EDR										41.2%
3 McAfee	Enterprise Security			Enterprise Security	Enterprise Security	Enterprise Security	Enterprise Security	Enterprise Security	Enterprise Security										35.3%
4 FireEye	FireEye Helix			FireEye Helix	FireEye Helix	FireEye Helix	FireEye Helix	FireEye Helix	FireEye Helix										
5 ServiceNow																			35.3%
6 RSA																			35.3%
7 Trend Micro	Trend Micro Cloud One			Trend Micro Deep Security	Trend Micro Deep Security	Trend Micro Deep Security	Trend Micro Deep Security	Trend Micro Deep Security	Trend Micro Deep Security										29.4%
8 LogRhythm	LogRhythm Security			LogRhythm Security	LogRhythm Security	LogRhythm Security	LogRhythm Security	LogRhythm Security	LogRhythm Security										29.4%
9 MetricStream																			29.4%
10 Fortinet	Fortinet FortiSIEM			Fortinet FortiSIEM	Fortinet FortiSIEM	Fortinet FortiSIEM	Fortinet FortiSIEM	Fortinet FortiSIEM	Fortinet FortiSIEM										23.5%
11 Splunk	Splunk Enterprise Security			Splunk Enterprise Security	Splunk Enterprise Security	Splunk Enterprise Security	Splunk Enterprise Security	Splunk Enterprise Security	Splunk Enterprise Security										23.5%
12 Palo Alto Networks																			23.5%
13 Lockpath																			23.5%
14 Forescout																			17.6%
15 SolarWinds	SolarWinds Security Event Manager			SolarWinds Security Event Manager	SolarWinds Security Event Manager	SolarWinds Security Event Manager	SolarWinds Security Event Manager	SolarWinds Security Event Manager	SolarWinds Security Event Manager										17.6%
16 CrowdStrike																			17.6%
17 LogGate																			17.6%
18 ZenGRC																			17.6%
19 Resolver																			17.6%
20 Darktrace	Darktrace Immune System			Darktrace Immune System	Darktrace Immune System	Darktrace Immune System	Darktrace Immune System	Darktrace Immune System	Darktrace Immune System										11.8%

(continued)

Table 4.

41	ProcessUnity																		ProcessUnity	ProcessUnity	11.8%
42	Netwrix Auditor																		Netwrix Auditor	Netwrix Auditor	11.8%
43	Sumo Logic Cloud SIEM																				5.9%
44	Netkope Security Cloud																				5.9%
45	Rapid7 InsightVM																				5.9%
46	Carbon Black																				5.9%
47	Bitdefender GravityZone Ultra																				5.9%
48	LogPoint SIEM																				5.9%
49	Elastic Security																				5.9%
50	Check Point CloudGuard																				5.9%
51	Tufin																				5.9%
52	RiskSense																				5.9%
53	Trustwave																				5.9%
54	Imperva																				5.9%
55	Cynet																				5.9%
56	ACL																				5.9%
57	Galvanize																				5.9%
58	Comply Advantage																				5.9%
59	Ncontracts																				5.9%
60	Nlyte																				5.9%
61	Ansible																				5.9%
62	Open Policy Agent (OPA)																				5.9%
63	ArcherDX																				5.9%
64	Compliance.ai																				5.9%
65	LogicManager																				5.9%
66	Quantivate GRC																				5.9%
Current available		20	0	19	7	0	0	0	0	13	0	14	7	0	0	13	0	15	16	17	17

Source: Created by authors

Our results indicated that all 14 core cybersecurity tasks of the technical work role are currently indicated to be performed by at least 7 tools, while 6 of the 17 total core cybersecurity tasks of the managerial work role have no evidence of current AI tools that can perform these tasks. Specifically, for the 11 managerial tasks which are AI-assisted, 7 to 20 tools are provided. However, six tasks do not have even a single AI tool that can assist in performing these tasks. IBM is the vendor with the larger number of tasks covered (47.1%), and Symantec is the second with 41.2%. Nevertheless, most vendors cover less than 18% of the tasks. For all the 14 technical tasks, at least seven different AI tools currently exist. Darkface is the vendor with the largest number of tasks covered (71.4%), and 8 of the 35 vendors cover more than 50% of the tasks. It appears that commercial tools to perform highly managerial and organizational cybersecurity strategies are less common as these are typically the tasks that, until January 2022 (ChatGPT 3.5), appear to fall under the purview of cybersecurity professionals.

5. Conclusions

Automation has the potential to free up humans from monotonous, repetitive or high-attention-demanding tasks, thus providing them with more opportunities for strategic thinking, expediting decision-making, plan implementation and cybersecurity management. As a result, professionals can enhance their efficiency, completing existing tasks with greater precision while gaining the capacity to undertake new responsibilities. The potential implications of AI technologies for cybersecurity are immense. Using various AI tools based on different techniques, technologies and algorithms can assist with many of the different tasks of technical and managerial cybersecurity roles. Nonetheless, these are the first steps of evolution, while commercial companies are developing more and more specific tools, which appear to focus first on the technical cybersecurity tasks. Thus, it reduces overall organizational costs as it can take over repetitive tasks, resulting in using fewer cybersecurity professionals and enabling them to focus on other more strategic and managerial tasks that AI tools currently appear to shy away from. Combining human expertise and machine intelligence is important to leverage the synergies of both humans and machines. Cybersecurity professionals can be more effective in their work with AI-enabled cybersecurity platforms as it helps minimize the time and effort necessary to perform those tasks. We predict that the evolution of AI tools, especially those based on GPT, will simplify the tasks of managerial roles too. However, these platforms may introduce risks, such as biases, privacy breaches and more, that organizations using them must be aware of.

6. Future research

While our work here is preliminary, our study brings several interesting points that should be addressed in future research. Specifically, given the existing use of AI within the technical work role, future research should further assess how much efficiency is occurring. It is important to know that, as of the current state of AI integration, it is unclear to what extent AI and GPT are actually helping professionals become more efficient in what they do or if they cause individuals to actually spend more time trying to figure out how to seek the help of AI to do the tasks they wish it to do. Academic institutions must look closer at the current state of AI usage in cybersecurity and develop opportunities for students during their academic degrees to build their human-AI teaming skills, especially for those seeking to engage in future cybersecurity work

roles. One initial national effort in that direction is led by the NSA in collaboration with the National Science Foundation in establishing the criteria for what students should be required to learn and what competencies are expected from them when graduating from an NSA-designated CyberAI program (<https://www.towson.edu/cyberai>). The second interesting point where future research may help is to seek empirical results on how much of the labor deficiency is being offset by existing AI tools. Such results can provide predictions and trends in closing the gap for the cybersecurity work shortage. The third point for future research is to help assess the future labor demands, especially in the technical aspect of cybersecurity, given the evolution and trends in using AI tools, as found in our studies, primarily for technical tasks. In summary, future research should help us address the anticipated level of efficiency gained by the use of AI, what is the number of cybersecurity vacant jobs that will be reduced by AI and what will be the future workforce requirements.

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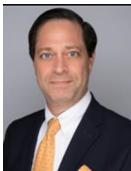
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Further reading

DCWF Tool, available at: <https://public.cyber.mil/wid/dcwf/>

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