

Open Source Intelligence - Development of a Trend Radar utilizing a Systematic Literature Review

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Abstract

Open Source Intelligence (OSINT) is currently experiencing an intensive discourse, heightened since the Russian invasion of Ukraine. However, despite numerous attempts at standardized definitions, the intelligence discipline remains ambiguous. This paper introduces a practice-validated OSINT trend radar, categorizing technologies by maturity, intelligence cycle phase, and use case. Serving as a profound knowledge base and tool for identifying research gaps, the radar emerges from a structured design process. Sixty studies underwent categorization and validation through expert interviews, revealing the absence of a comprehensive, automated third-generation OSINT system in Germany. Technological gaps, especially in the planning, direction, dissemination, and integration phases, are evident. Although intelligent support technologies were identified, practical implementation lags behind theory. The human factor therefore remains central to the OSINT process. Future research should thus prioritize developing applications for underserved phases, probing reasons for limited widespread implementation of proven applications, with emphasis on legal, ethical, political, and social parameters.

1. Introduction

OSINT is a currently more debated research field than ever before. Obtaining intelligence from publicly available data [1] has become undeniably important since the Russian invasion of Ukraine in 2022. The real-time analysis of social media in particular has brought highly relevant findings to light [2, 3]. However, OSINT is not a new technique [4, 5]. However, despite numerous attempts to define OSINT (e.g. [6, 7, 8]), the concept remains controversial to this day [9, 10, 11]. This is not least due to the fact that every definition of OSINT is subject to the advances in computer and data sciences, which continuously produce improvements in

(intelligent) collection and analysis possibilities [9, 11]. In addition, this is accompanied by numerous new open means of communication, which have caused a veritable "information explosion" [1, 6, 8]. Data sources that were originally reserved for the defense and intelligence services are now also publicly accessible [6, 11]. The understanding of intelligence thus changed completely [12].

To date, there has been a lack of decisive, fundamental scientific publications to penetrate the subject area [13] and address its rapid developments [9, 11]. In particular, there is a shortage of studies that identify the technologies behind OSINT and determine their characteristics. The question of whether "third generation" OSINT systems in the sense of robust, autonomous [4, 7] already exist has not yet been clarified [9, 7, 8]. Moreover, the majority of studies focus exclusively on analyzing the OSINT trend area "cyber security" [6, 4, 8]. The literature thus failed to cover the topic in its entirety. Important application scenarios ("use cases") of OSINT have therefore not yet been taken into account in research [14, 12, 9]. In addition, supplementary qualitative field research is absent to contrast theory with the corresponding practical implementation. Although OSINT has a major impact on topics such as security and defense, there is a lack of insight into these sectors [13, 4]. This paper is hence dedicated to answer the research question: *How can the current trends in OSINT in the form of the technologies used and their characteristics, in particular the maturity level and the use case, be presented in a trend radar and validated by experts within the security sector?*

The aim thereby is to identify the technologies used in OSINT applications and to present them systematically in a trend radar, according to their characteristics. Through expert interviews, trends will then be validated. In this way, a well-founded knowledge base will be compiled, and practice-relevant research gaps will be identified for a coordinated examination of the research field. The study thereby

follows the "Design Science Research Model" (DSRM) [15]. First, the relevant literature on OSINT will be analyzed and classified through a systematic literature review [16]. Second, the OSINT technologies and their characteristics identified will be visualized in the form of a trend radar. The radar will then be validated using systematizing interviews [17] conducted with experts in the security sector. Finally, the interviews will be evaluated using a qualitative content analysis [21].

2. Theoretical Background

The domain of OSINT is continuously expanding due to the ongoing development of improved collecting and analysis possibilities [14, 9, 11]. In addition, the new means and methods of communication associated with advances in information and communication technology have turned OSINT into a complex discipline [14, 18, 19, 11]. OSINT and its components are therefore defined in detail hereinafter.

2.1. Open Source Intelligence (OSINT)

One of the earliest and still frequently referenced definitions [1] was published by NATO in 2001. OSINT according to this definition is information that has been deliberately discovered, discriminated, distilled, and disseminated to a select audience, [...], in order to address a specific question. OSINT, [...] thus applies the proven process of intelligence to the broad diversity of open sources [...] and creates intelligence [23]. However, today the discipline is no longer seen as a purely governmental matter. Private research institutions and organizations [20, 21] are also massively driving the development of such systems, e.g. for competitive analyses or marketing activities [14, 12, 9]. The focus is thereby shifting to developing OSINT into a robust, autonomous solution [4,8,9].

2.2. Open Source Data (OSD)

The starting point for all OSINT activities is data. Data forms the basis of the analysis and the conclusions derived from it [22]. In this context, OSD refers to non-processed [1], general raw data that is openly available [23] as well as legally and ethically accessible [5, 24]. In practice, sources whose access requires additional effort [20] or must be acquired commercially [11, 24, 25] are not excluded.

2.3. Open Source Information (OSIF)

OSD are of little use on their own and only become of (intelligence) relevance when they are aggregated

[11]. Before intelligence can be obtained from them, the data must therefore undergo a preparation process that includes filtering, validation and summarization [1, 24]. The result of this data organization [5] is referred to as OSINF. It provides the basis for the resulting intelligence creation [1, 5].

2.4. Intelligence and Intelligence Cycle

The core task of OSINT is to generate intelligence [6, 12] from the condensed information in the terms of a profound basis for decision-making [26, 27]. The generation process of such an intelligence product is also referred to as the intelligence cycle [13, 28]. It represents the central element of every intelligence discipline [29, 12]. The representation of the process as a cycle [30] dates back to the CIA in 1987 [28]. The link between the phases is that the result of the preceding phase serves as input for the subsequent phase [25, 31]. Furthermore, the individual phases are also continuously iterated due to the fulfillment of previous requirements and new demands [22]. However, to improve the representation of external influences or the assignment of responsibilities [32, 33, 34], numerous variations can be found in the literature today [20, 29]. The Intelligence Cycle should therefore be seen less as a guideline and more as an informal coordination element, with a partly very intuitive [26] interpretation [6]. In an updated version from 2013, the JCS segment the cycle into 6 phases [32] (see Fig 1).

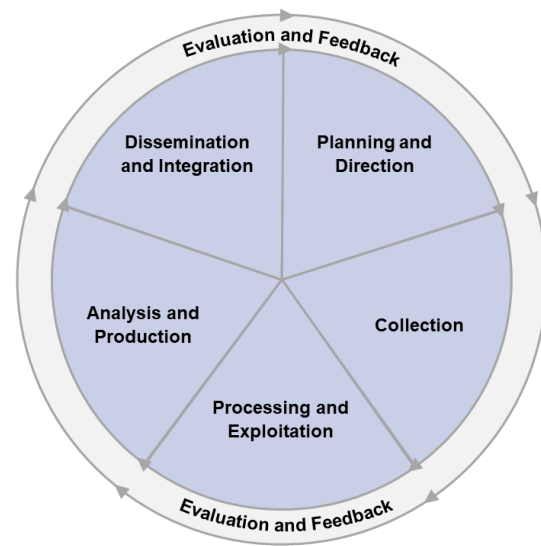


Figure 1. Intelligence Cycle, according to [25]

The planning and direction phase combines the identification, definition and prioritization of the requirements for the cycle. It is also responsible for

developing the activities required to achieve these [35] and monitoring their implementation [25, 35]. The collection phase refers to the gathering of raw data [28]. The core of this phase consists of the iterative repetition of research [24] to make the query more precise with each run [7]. The processing and utilization phase involves condensing these data volumes into valuable and action-relevant information for further processing [30, 25, 7, ?]. Analysis and production refers to the synthesis of the information obtained into a user-oriented, timely and accurate intelligence product [35, 6, 24]. The final phase consists of handing over the finished product to the "customer" in a usable form [36, 35, 11]. Evaluation and feedback are not to be regarded as individual phases within the cycle but take place continuously throughout the entire process. The aim is to achieve progressive optimization [30, 25, 24].

2.5. Previous Studies

In the literature, eight previous, publicly accessible literature reviews can be identified concerning OSINT. In 2017, Dos Passos [1] showed how big data and data science can make the decision-making process more useful and effective. Pastor-Galindo et al. then described the current state of OSINT in 2019 [4] and 2020 [7] focusing on services and techniques to improve cybersecurity. Moreover, they are also responsible for the first and only rudimentary mapping of OSINT trends. They observed that OSINT is used in social opinion and sentiment analysis, cybercrime and organized crime, and cybersecurity and cyberdefence. Two further literature reviews were published in 2020. García Lozano et al. [37] identified methods for computer-assisted veracity assessment of public information. Herrera-Cubides et al. [13] investigated how the production of research and educational materials in the field of OSINT has developed. They concluded that the number of publications is lower compared to other trending topics. Finally, in 2021, Yogish and Krishna [8] explored the state of implementation and use of AI (Artificial Intelligence) technologies in the context of cybersecurity. The result of this study showed that Machine Learning (ML), pattern recognition and Natural Language Processing (NLP) can simplify OSINT in view of increasing data volumes. In the following year, Hwang et al. [6] investigated security threats and cybercriminality in the context of OSINT misuse. In 2023, Ghioni et al. [9] then examined the political, ethical, legal and social implications of OSINT in conjunction with AI. They discovered that there is still no framework for addressing these. They also found that third-generation OSINT

is still in its early stages and that human components cannot yet be replaced.

3. Research Methodology

The structure of the study is based on the iterative DSRM [15]. It is a theory-based research paradigm for developing an explicitly applicable solution, in the form of an innovative artifact, [38], solving a (practical) problem [15, 39]. The model is therefore ideally suited to create the trend radar. It consists of six successive activities (see Fig. 2) [15]. Moreover, the first three evaluation steps according to Sonnenberg and vom Brocke [40] were applied to continuously evaluate the design process.

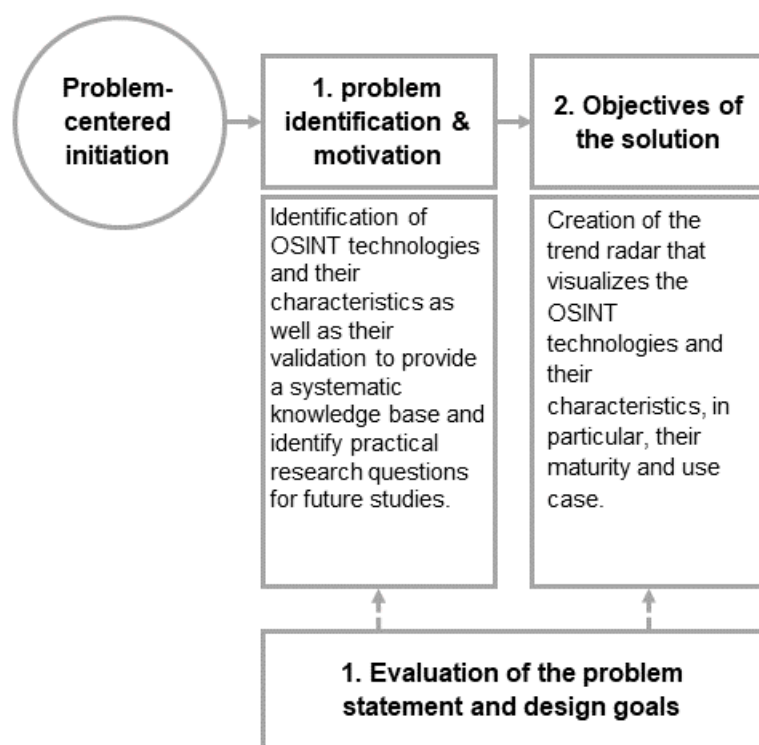


Figure 2. Design Science Research Model (DSRM)

3.1. Problem Identification and Motivation

First, the research problem must be defined and the benefits of the solution explained. This activity can be found in the chapter Introduction.

3.2. Design Objectives of the Solution

The next step entails defining the design objectives of the solution. The objectives can be divided into two content-related objectives (CO) and two formal

objectives (FO).CO1 requires the trend radar to follow a procedural structure that reflects the process of generating knowledge from public information. This will allow a structured mapping of the identified technologies according to their use. In doing so, any research gaps that become apparent can be directly assigned to the respective phase. It will thus be possible to verify if a robust, autonomous OSINT system exists. As CO2, it was defined that the key characteristics, in particular the maturity level of the technologies and their use cases, must be taken into account. The maturity level of the technologies makes it possible to determine the respective state of research. Through the use cases the research directions can be revealed. As FO1, it was determined that the trend radar must follow a simple structure in order to provide an easy-to-understand knowledge base for the immediate identification of research gaps. In addition, the radar should have a high degree of standardization to be transferable to other intelligence (gathering) disciplines in later studies. As FO2, it was set out to design the trend radar in an adaptable and continuously expandable way allowing it to capture the high field dynamics.

3.3. Evaluation of the Problem Statement and Design Objectives

In the theoretical background section, the main definitions and, in particular, the intelligence cycle, which serves as a basis for the development of the trend radar ("suitability"), were presented. In addition, the presentation of previous research work showed that there is still a lack of comprehensive studies examining OSINT. Both the relevance of the research question raised and the suitability of the adopted design objectives are thus underpinned, fulfilling the evaluation criteria according to Sonnenberg and vom Brocke [?].

3.4. Design and Development

The third activity involves the creation of the artifact. For this purpose, the relevant literature on OSINT was analyzed by means of a systematic literature review, following the guidelines of Cleven et al [47]. To clearly determine the scope of the literature review, Cooper's taxonomy [48] was used. Subsequently, classification categories were defined to establish concept matrices, based on the model of Webster and Watson [24], for systematically analyzing the researched literature. To achieve a high degree of standardization, general categories were defined for each phase of the intelligence cycle. Thereby, the categories of the collection phase differ from those of the remaining phases, as the data basis is also considered

here. The evaluation and feedback phase was not treated separately due to its iterative nature.

The following six categories were defined with regard to the survey phase: "Use case", "Data", "Process", "Technology", "Technology complexity" and "Maturity level". First of all, the use cases were used to record the area of application of the technologies. The category data reveals the composition of the data foundation. Therefore, it is further subdivided into the data type to record the formats and the source to show the origin of the data. The third category, the process, is used to determine the degree of automation or autonomization of the technologies. For this purpose, the following four levels were defined: manual, semi-automated, automated, [Duncheon 2002] fully automated/autonomous [Billing 1997, Endsley and Kaber 1999]. To improve categorization, the ten levels of automation [Sheridan and Verplank 1978, Parasurama 2000] were additionally used.

The fourth category serves to capture the technologies, according to Bleck's definition [49], defined as "the totality of material and immaterial means available for the input, output, conversion, transmission and storage of information".

The fifth category evaluates the complexity of the technologies examined. For the collection phase, the three subcategories "Volume"[53], "Variety" and "Velocity" were used [50-52]. Variety is further subdivided according to the data structure (structured [54], semi-structured and unstructured [55,56]) [53]. Velocity, is further subdivided into the levels "Batch" [58], "Near Real-Time" [59,60] and "Real-Time" [59].

The sixth category reflects the maturity level of the technologies used. For this purpose, the three macro-maturity phases of the German Federal Trend Radar are used: the innovation phase, the prototype phase and the market establishment phase [61].

For the remaining phases of the intelligence cycle, the categories use case, process, technology, technology complexity and maturity level were likewise defined. However, the complexity within the remaining phases is measured on the basis of the underlying analysis, in ascending order: descriptive analysis, diagnostic analysis, predictive analysis through to prescriptive analysis [63,64].

After defining the classification criteria, the literature research was carried out (see Fig. 15). For this, a search string based on general terms for the highest possible number of hits was determined. The string was then applied to the four databases "Web of Science", "IEEE Xplore", "ACM Digital Library" and "arXiv". The search results were restricted to publications in German or English. Only studies with full access were

considered. As the number of publications increased significantly between 2020 and 2023, the period was reduced to these years to ensure the most up-to-date coverage possible. The studies were downloaded on 05.06.2023 (see Appendix D.B). The SQR3 method [65] was applied for the systematic analysis and further delimitation. Altogether 60 studies were categorized following this procedure.

For categorization, a dedicated Excel spreadsheet was created for each Intelligence Cycle phase and the OSINT technologies and their characteristics methodically recorded [24,47]. The identified technologies were then grouped according to related technologies in an additional Excel column to enable the clear final illustration in the trend radar (see Repository X). Afterwards, the categorization was reviewed by using a "Python" script (see Appendix A). The script queries the occurrence of predefined keywords in the included papers. If a deviation was found, the study in question was analyzed again (see review matrix X). Finally, the trend radar was created on the basis of the concept matrices.

3.5. Evaluation of the Design Specifications

The intelligence cycles as the basis of the trend radar provides a structured overview ("clarity") (Breakspear, 2013, p. 689 f.). This meets the requirement of an intuitively understandable illustration ("under-standability") for ensuring a simple extraction ("simplicity") of technologies and research gaps. Furthermore, the other intelligence disciplines are also derived from the cycle, which enables a later application to these ("applicability"). In addition, by aligning the architecture of the radar with the federal government's trend radar (cf. Stich et al., 2022, p. 11 f.), a proven, robust, user-friendly design ("userfriendliness") with an appropriate level of detail ("level of detail") is used. Moreover, with the concept matrices, a template is used to constantly update the radar in order to reflect the highly dynamic nature of the research topic. Particular attention was paid to the general validity of the categories in order to achieve the highest possible degree of standardization ("commonality"). The categorization was also reviewed using the Python script ("internal consistency"). The decisive evaluation criteria, according to Sonnenberg and vom Brocke (2012, p. 13), are thus fulfilled.

3.6. Demonstration

Next, the trend radar was demonstrated in the context of guided, systematizing expert interviews. Especially in less structured and sparsely linked subject

areas, the method enables dense data collection [66,25]. Additionally, it is suitable in cases where access to the social field is limited [66,26]. The execution thereby follows the concepts of Bogner et al [25], Meuser and Nagel [67] and Gläser and Laudel [26].

The experts (see Tab. 1) were selected according to the method of theoretical sampling [68,69]. It was specified that only experts in Germany should be interviewed in order to validate the trends in this country. Furthermore, it was determined that at least one expert each from a security authority, the established security industry and a start-up would be selected in order to capture different points of view. Following Meuser and Nagel [70], an expert was considered to be someone with a knowledge advantage in the specific area of interest. A "prestigious" company position is they regarded as a reliable guarantee that the respondents possess research-relevant knowledge [72].

The qualitative data collection that followed was carried out using semi-structured interviews. These are particularly suitable for revealing the underlying relationships of a theory [25]. The questionnaire used is based on the structure of the Intelligence Cycle. At the beginning of the interview, the subjects were introduced to the trend radar. In order to compare this theoretical construct with the practical experience of the participants, higher-level, open questions were asked for each phase. These questions reduce the influence of subjectivity [73]. In order to direct the conversation flow in a targeted manner, the superordinate questions are supplemented by exploratory questions [73]. At the third level, these questions are complemented by specific closed questions for targeted follow-up queries [73]. The interviews were held within a time frame of up to one hour, with a maximum of three main questions for each section [25] (see Repository X). The questionnaire was pilot-tested with a domain expert. The expert interviews were conducted online via platforms with video chat functions.

3.7. Evaluation of the First Instance of the Trend Radar

The demonstration of the trend radar confirmed its intuitive comprehensibility ("ease of use"). It was furthermore perceived by the practitioners as a useful tool for providing an overview of OSINT technologies ("effectiveness"). In addition, they confirmed its completeness ("completeness") and internal consistency ("consistency") (cf. expert E1, 14.08.2023; expert E3, 28.07.2023; expert E4, 02.08.2023). The design of the trend radar thus proved to be a suitable instrument for identifying research gaps in

theory-practice comparisons and for serving as a guideline to practitioners ("fidelity with real world phenomenon"). The essential evaluation criteria, according to Sonnenberg and vom Brocke (2012, pp. 15-18), are thus demonstrably fulfilled in practice.

3.8. Evaluation

The evaluation was carried out using a qualitative data analysis according to Gläser and Laudel [26]. The aim of this is to extract, synthesize and structure the information contained in the interviews on the basis of a predefined search grid. This enables the targeted extraction and summarization of relevant, cross-interview information according to a "top-down approach" [26,25]. The established category system was applied for this purpose.

First, the recorded interviews were transcribed. Second, the software MAXQDA [75] was used for the subsequent categorization and analysis. The categorization system was therefore established within MAXQDA (see repository X). The categories of the first level correspond to the individual phases of the intelligence cycle. The categories of the second level allow a classification of whether the experts express themselves in support of or in contradiction to the theory. The categories of the third level reflect the identified use cases. As many of the experts' statements were of a general nature, the category "General statements" was added at this level. The categories of the fourth level reflect the individual technologies classified under them. Altogether, 257 statements from the experts were assigned to corresponding categories by this method. Examples of the coding procedure can be found in (Table 3).

3.9. Communication

The research results are communicated in the form of this study. A new iteration, proceeding from this chapter, is therefore not carried out.

4. Graphics/Images

All images must be embedded in your document or included with your submission as individual source files. The type of graphics you include will affect the quality and size of your paper on the electronic document disc. In general, the use of vector graphics such as those produced by most presentation and drawing packages can be used without concern and is encouraged.

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5. Main text

Type your main text in 10-point Times, single-spaced. Do not use double-spacing. All paragraphs should be indented 1/4 inch (approximately 0.5 cm). Be sure your text is fully justified—that is, flush left and flush right. Please do not place any additional blank lines between paragraphs.

Figure and table captions should be 9-point boldface Helvetica (or a similar sans-serif font). Callouts should be 9-point non-boldface Helvetica. Initially capitalize only the first word of each figure caption and table title. Figures and tables must be numbered separately. For example: "Figure 1. Database contexts", "Table 1. Input data". Figure captions are to be centered below the figures. Table titles are to be centered above the tables.

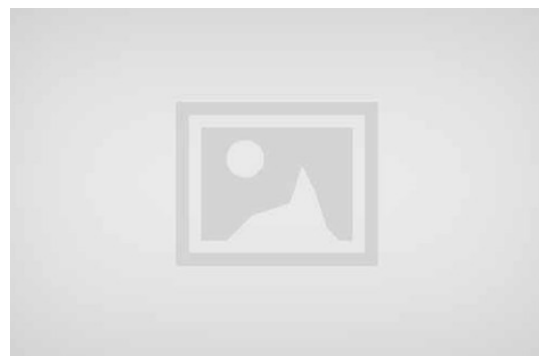


Figure 3. Sample figure with caption.

6. First-order headings

For example, “1. Introduction”, should be Times 12-point boldface, initially capitalized, flush left, with one 12-point blank line before, and one blank line after. Use a period (“.”) after the heading number, not a colon.

6.1. Second-order headings

As in this heading, they should be Times 11-point boldface, initially capitalized, flush left, with one blank line before, and one after.

6.1.1. Third-order headings. Third-order headings, as in this paragraph, are discouraged. However, if you must use them, use 10-point Times, boldface, initially capitalized, flush left, followed by a period and your text on the same line.

7. Footnotes

Use footnotes sparingly and place them at the bottom of the column on the page on which they are referenced. Use Times New Roman 8-point type, single-spaced. To help your readers, try to avoid using footnotes altogether and include necessary peripheral observations in the text (within parentheses, if you prefer, as in this sentence). asdöfj

8. References

References

- [1] D. S. Dos Passos, “Big data, data science and their contributions to the development of the use of open source intelligence,” *Sistemas & Gestão*, vol. 11, no. 4, pp. 392–396, 2017.
- [2] J. M. Hatfield, “There is no such thing as open source intelligence,” *International Journal of Intelligence and CounterIntelligence*, pp. 1–22, 2023.
- [3] V. Smith-Boyle, “How osint has shaped the war in ukraine,” 24.07.2023.
- [4] J. Pastor-Galindo, P. Nespoli, F. Gomez Marmol, and G. Martinez Perez, “Osint is the next internet goldmine: Spain as an unexplored territory.”
- [5] F. Schaurer and J. Störger, “The evolution of open source intelligence.”
- [6] Y.-W. Hwang, I.-Y. Lee, H. Kim, H. Lee, and D. Kim, “Current status and security trend of osint,” *Wireless Communications and Mobile Computing*, vol. 2022, pp. 1–14, 2022.
- [7] J. Pastor-Galindo, P. Nespoli, F. Gomez Marmol, and G. Martinez Perez, “The not yet exploited goldmine of osint: Opportunities, open challenges and future trends,” *IEEE Access*, vol. 8, pp. 10282–10304, 2020.
- [8] P. U. Yogish and P. K. Krishna, “Open source intelligence and its applications in next generation cyber security - a literature review,” 2021.
- [9] R. Ghioni, M. Taddeo, and L. Floridi, “Open source intelligence and ai: a systematic review of the gelsi literature,” *AI & society*, pp. 1–16, 2023.
- [10] D. Ish, J. Ettinger, and C. Ferris, *Evaluating the Effectiveness of Artificial Intelligence Systems in Intelligence Analysis*, vol. RR-A464-1 of *Research reports*. Santa Monica, Calif.: RAND Corporation, 2022.
- [11] H. J. Williams and I. Blum, *Defining Second Generation Open Source Intelligence (OSINT) for the Defense Enterprise*, vol. RR-1964-OSD of *Research reports*. Santa Monica, Calif.: RAND Corporation, 2018.
- [12] T. Dokman and T. Ivanjko, “Open source intelligence (osint): issues and trends,” in *INFuture2019: Knowledge in the Digital Age*, International Conference The Future of Information Sciences INFuture, Faculty of Humanities and Social Sciences, University of Zagreb Department of Information and Communication Sciences, FF press, 2020.
- [13] J. F. Herrera-Cubides, P. A. Gaona-García, and S. Sánchez-Alonso, “Open-source intelligence educational resources: A visual perspective analysis,” *Applied Sciences*, vol. 10, no. 21, p. 7617, 2020.
- [14] H. Alkilani and A. Qusef, “Osint techniques integration with risk assessment iso/iec 27001,” in *International Conference on Data Science, E-learning and Information Systems 2021* (J. A. Lara Torralbo, S. A. Aljawarneh, V. Radhakrishna, and A. N., eds.), (New York, NY, USA), pp. 82–86, ACM, 2021.
- [15] K. Peffers, T. Tuunanen, M. A. Rothenberger, and S. Chatterjee, “A design science research methodology for information systems research,” *Journal of Management Information Systems*, vol. 24, no. 3, pp. 45–77, 2007.
- [16] J. Webster, Watson, and T. Richard, “Analyzing the past to prepare for the future: Writing a literature review,” *MIS Quarterly*, vol. 26, pp. xiii–xxiii, 2002.
- [17] A. Bogner, B. Littig, and W. Menz, *Interviews mit Experten: Eine praxisorientierte Einführung*. Wiesbaden: Springer Fachmedien Wiesbaden, 2014.
- [18] L. Benes, “Osint, new technologies, education: Expanding opportunities and threats. a new paradigm,” *Journal of Strategic Security*, vol. 6, no. 3Suppl, pp. 22–37, 2013.
- [19] Chen, Chiang, and Storey, “Business intelligence and analytics: From big data to big impact,” *MIS Quarterly*, vol. 36, no. 4, p. 1165, 2012.
- [20] I. Böhm and S. Lolagar, “Open source intelligence,” *International Cybersecurity Law Review*, vol. 2, no. 2, pp. 317–337, 2021.
- [21] S. C. Mercado, “Reexamining the distinction between open information and secrets,” *Studies in Intelligence*, vol. 49, no. 2, 2005.
- [22] H. Gibson, “Acquisition and preparation of data for osint investigations,” in *Open Source Intelligence Investigation* (B. Akhgar, P. S. Bayerl, and F. Sampson, eds.), Advanced Sciences and Technologies for Security Applications, pp. 69–93, Cham: Springer International Publishing, 2016.

- [23] C. Burke, "Freeing knowledge, telling secrets: Open source intelligence and development," *Research paper series: Centre for East-West Cultural & Economic Studies*, no. 13, 2007.
- [24] "Nato open source intelligence handbook."
- [25] "Joint intelligence."
- [26] A. Breakspear, "A new definition of intelligence," *Intelligence and National Security*, vol. 28, no. 5, pp. 678–693, 2013.
- [27] "The security intelligence handbook: Third edition: How to disrupt adversaries and reduce risk with security intelligence."
- [28] "Factbook on intelligence."
- [29] A. Reuser, "The rise of open source intelligence cycle," *Journal of Mediterranean and Balkan Intelligence*, vol. 10, no. 2, 2017.
- [30] "U.s. national intelligence: An overview 2011."
- [31] R. Pellissier and T. E. Nenzhelele, "Towards a universal competitive intelligence process model," *SA Journal of Information Management*, vol. 15, no. 2, 2013.
- [32] M. M. Lowenthal, *Intelligence: From secrets to policy*. Thousand Oaks, Calif.: SAGE/CQ Press, eighth edition ed., 2020.
- [33] M. Phythian, ed., *Understanding the intelligence cycle*. Studies in intelligence, London: Routledge, 2013.
- [34] R. Johnston, *Analytic culture in the US intelligence community: An ethnographic study*. Washington, DC: Central Intelligence Agency (CIA), 2005.
- [35] "Open-source intelligence."
- [36] "The intelligence cycle: Briefing."
- [37] M. García Lozano, J. Brynielsson, U. Franke, M. Rosell, E. Tjörnhammar, S. Varga, and V. Vlassov, "Veracity assessment of online data," *Decision Support Systems*, vol. 129, p. 113132, 2020.
- [38] J. vom Brocke, A. Hevner, and A. Maedche, "Introduction to design science research," in *Design Science Research. Cases* (J. vom Brocke, A. Hevner, and A. Maedche, eds.), Progress in IS, pp. 1–13, Cham: Springer International Publishing, 2020.
- [39] Hevner, March, Park, and Ram, "Design science in information systems research," *MIS Quarterly*, vol. 28, no. 1, pp. 75–105, 2004.
- [40] C. Sonnenberg and J. vom Brocke, "Evaluations in the science of the artificial – reconsidering the build-evaluate pattern in design science research," in *Design Science Research in Information*, vol. 7286, pp. 381–397, 2012.