

A Review on Security and Privacy Concerns With Big Data Analytics

Harshith D M¹, Abhiram A Gowda², Induraj A³, Abhishek B K⁴

Dr. Madhusudhan S [Associate Professor]

Department of Computer Science and Engineering.

Visvesvaraya Technological University.

Alvas Institute of Engineering and Technology.

Moodabidri, Dakshina Kannada, Karnataka, India

Abstract— In the ever-expanding realm of Big Data, the rapid proliferation of data collection, processing, and analysis has bestowed upon us unparalleled prospects and trials. The protection of security and privacy in the field of big data analytics is the most important of these worries. The purpose of this introductory segment is to present a broad outline of the contemporary technologies and methodologies employed to tackle these pivotal quandaries. The field of big data analytics involves extracting priceless insights, and the facilitation of decision-making through the processing of colossal datasets. However, Ensuring privacy and security is significantly hampered by the enormous volume, diversity, and speed of data. Traditional approaches to data protection often prove inadequate in this context, necessitating the advent of innovative and groundbreaking solutions. Recent strides in technology have been primarily focused on bolstering privacy and security within the domain of Big Data analytics. An exemplary instance of this is differential privacy, wherein noise is introduced into datasets to protect individual privacy while simultaneously preserving the overall utility of the data. Secure multi-party computation allows multiple parties to collectively compute functions based on their inputs, all without the need for disclosure. In addition, a new era of privacy-preserving strategies has been brought about by advances in machine learning and artificial intelligence. Additionally, blockchain technology has been used to create transparent and safe data exchange platforms, guaranteeing data accountability and integrity.. Despite the remarkable progress achieved, challenges continue to persist in the pursuit of privacy and security within the realm of Big Data analytics. Scalability, interoperability, and regulatory compliance stand as pivotal domains that necessitate further research and development. Moreover, the ever-evolving nature of data privacy regulations, coupled with the escalating sophistication of cyber threats, underscores the vital importance of unceasing innovation within this field.

I. INTRODUCTION

Today, data are integrated into every area and function of the global economy, and much of contemporary economic activity simply could not function without them, much like other necessary components of production like physical assets and human capital. Large data sets that can be combined and examined to identify trends and improve decision-making will form the foundation of individual businesses' growth and competition as data analysis improves productivity and adds substantial value to the global economy by cutting waste and raising the caliber of goods and services. As a matter of fact, no firm can thrive in the absence of data analysis. Think about the following scenarios. A pharmaceutical company is conducting clinical trials on multiple people to evaluate a novel medicine intended to treat a specific ailment. A corporation wants to introduce a fresh version of its current perfume line. It seeks to do the survey analysis and reach a

significant conclusion. The director of sales at a firm is aware that there is a problem with one of its best-selling goods, but he hasn't yet analyzed market research data. What does he conclude and how? These incidents show enough evidence to draw the conclusion that data analysis is essential to any company. Data analysis holds the key to solving all issues, be it making marketing decisions or fine-tuning a new product launch strategy. Social media is being monitored by marketing and advertising companies, and big data analysis is being used by insurance companies to determine which house insurance applications may be completed right away and which ones require an in-person visit for validation. [1] [2].

Retail establishments are interacting with brand evangelists, modifying the viewpoint of brand detractors, and even empowering passionate consumers to promote their goods. Through the use of social media, all of these things are achieved.

Hospitals predict which patients are likely to request readmission within a few months of being discharged by using medical data and patient records. Next, the hospital can save another expensive hospital stay.

Web-based firms are creating information products that integrate client data to provide more enticing recommendations and successful coupon programs.

Sports teams employ data. [3].

The majority of large data sets are intricate and extremely large. Big data's heterogeneity, volume, timeliness, complexity, and privacy issues impede any advancement at all stages in the procedure that allow data to be valuable. Big data is present in many different sources., such as emails, attachments, videos, audio files, and posts on social media. Individuals use Twitter in many different methods, keeping track of 250 million tweets every day. Every day, 4 billion people watch YouTube. Data is produced in gigabytes these days. Financial services, healthcare, retail, web/social, manufacturing, and government are just a few of the industries that stand to benefit from big data. Big data is currently present in every area of the world economy. According to our estimates, almost every area of the US economy had an average by 2005 [4].

II. BIG DATA

Recently, datasets that get so big that using conventional database management systems becomes difficult have been referred to as "big data." These are data collections that are too big for frequently used software programs to handle, and methods for storing, managing, processing, and capturing data in a fair length of time [5].. A single data collection can contain anywhere from a few dozen terabytes (TB) to several petabytes (PB) of data, as big data quantities are always growing. Consequently, gathering, storing, searching, sharing, analysing, and displaying data are some of the challenges associated with big data. Businesses are now examining vast amounts of extremely precise data to find previously unknown information.

254

In its research on big data, the McKinsey Global Institute (MGI) outlines the several commercial prospects that big data presents [12].

Paulo Boldi, among "Big data needs big intelligence, not big machines," the authors claim. "A key basis of competition and growth" is big data. Any decision-making process that is based on data is frequently referred to as analytics, including big data. The Corporate analytics and academic research analytics are the two categories into which term analytics is separated. The team applies their knowledge of statistics and data mining to corporate analytics.

Organize Information

The analysis of these data is straightforward. It is presented numerically, with numbers, transaction data, and so forth

Unorganized Information Complex information like email attachments and comments on images from social media platforms is contained in this data.

It is difficult to analyze these facts. The first person to discuss the five V's of big data management was Doug Lancy.

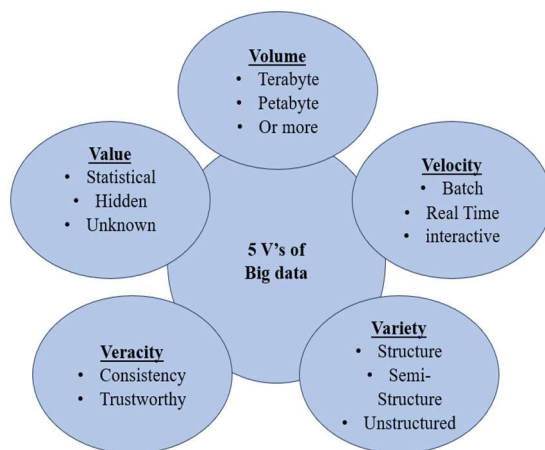


Fig.1: 5V's of Big Data [3]

Volume: This indicates the quantity of data. It alludes to large amounts of data.

Variety: It covers a range of data sources and formats, such as unstructured, semi-structured, and structured data.

Velocity: This describes the way that data moves. Quickly generated, processed, and evaluated data.

Veracity: In essence, veracity is the quality and reliability of your data. Just as it would be unwise to build a house on unstable ground, faulty data can have untrustworthy and perhaps deadly results. It is imperative that you confirm your data is: Precise: Devoid of mistakes, discrepancies, and prejudices.

- Complete: No substantial gaps or missing values prevent analysis from being done.
- Consistent: The dataset's formatting and structure hold true throughout.
- Traceable: You are aware of the source and history of your data. [6].

Value: Contrarily, value denotes the importance and use of your discoveries. Without actionable insights that add value to your firm, even the most accurate data is meaningless. This value can appear in a number of ways, including: Reducing costs involves finding inefficiencies and streamlining procedures. Increased revenue through

campaign targeting or the discovery of new market prospects.

Better ability to make decisions Strategic decisions are informed by data-driven insights [8].

Improved customer experience: Attending to specific needs and tailoring interactions.

III. CLASSIFICATION OF PRIVACY AND SECURITY

A. Hadoop.

Hadoop is a distributed process platform; security was not considered in its design.. Its intended operating environment was one of trust. Since Hadoop has gained popularity [5].

platform, security measures are beginning to be created. Additionally, scholarly interest in it has begun to grow.

Two methods were put forth when creating a Hadoop system that ensures the security and privacy of data on the cloud in order to thwart a hacker's desire to obtain all data on the cloud [7]. A trust mechanism has been established between the user and the name node, which is responsible for managing data nodes and is a component of HDFS. This technique requires the user to authenticate in order to access the name node. The name node generates a hash function after receiving a hash function from the user, and it compares the two functions. If the comparison outcome is accurate; a system for accessing it is offered. One of the hashing algorithms, SHA-256, is utilized for authentication at this stage. Data has also been encrypted using random methods including RSA, Rijndael, AES, and RC6, which prevent hackers from accessing the entire data set. This method uses MapReduce to carry out the encryption and decryption operation [12].

Lastly, a Twitter feed is used to test these two methods in order to provide maintenance tips for security vulnerabilities.

The Hadoop Distributed File System is another component that contributes to the security flaw (HDFS). HDFS security can now be fortified in three different ways.

B. Cloud Security:

The growing use of cloud computing due to its elastic nature, on-demand services, resource pooling, and wide network connectivity has created an ideal environment for big data.

Yet, cloud computing is home to both novel and established risks. [21] These days, cloud data storage is one of the biggest issues. The service provider must therefore proceed with caution. Consequently, a secure approach to the management and distribution of massive volumes of data on cloud platforms has been unveiled. To safely store large amounts of data, it incorporates a variety of security techniques such authentication, encryption, decryption, and compression. The authorized person has used password and email authentication. Compressed and encrypted data has been used to avoid security problems. It also adopts safety measures in the natural phenomenon. [6]

C. Monitoring and auditing:

The goal of security monitoring is to detect and investigate network events in order to identify intrusions. Security auditing is a methodical, quantifiable security policy that employs many techniques. These two in active security, factors are crucial.

It can be challenging to identify and prevent intrusions on the entire network flow. To tackle this problem, a security monitoring architecture has been created that examines IP flow records, HTTP and DNS traffic, as well as honeypot data. The recommended method for processing and storing data from dispersed sources makes use of data correlation algorithms. To assess the possibility that a domain name, packet, or flow is malicious, three likelihood metrics have currently been calculated.

D. Key MANAGEMENT

Another important data security concern is key generation and sharing between servers and consumers. Nevertheless, rapid and dynamic authentication techniques can be used in massive data centers. Recommended.

A layered approach for quantum cryptography with minimal complexity and PairHand protocol for authentication in mobile or fixed data centers have been suggested in [1]. The front end, data reading, quantum key processing, quantum key management, and application layers make up the model's layers, in that order. In addition to increasing efficiency, this paradigm has decreased passive attacks and important search activities.

E. Anonymization:

Gathering data for analytics raises serious privacy issues.

Because the data are transferred too quickly, protecting personally identifiable information (PII) is becoming more and more challenging.

Policies must govern the agreement between the individual and the business in order to remove privacy concerns [25].

It is necessary to anonymize (de-identify) personal data before transferring it via secure channels. Nevertheless, the person's identity may be discovered based on the company's artificial intelligence analysis and algorithms. This analysis's predictions may result in unethical situations.

F. Analytic Processing of Big Data

I'll now go over how the Hadoop framework is used to execute big data analytics on large amounts of data. Hadoop is an open source framework, as was previously said.

framework that consumers can readily access. Its foundation is the Map Reduce programming paradigm. The Map Reduce model is a programming approach that utilizes two phases of computing. The key component of Hadoop, which handles huge data analysis, is MapReduce. The map reduce algorithm is based on the Map and Reduce functions. Partitioning the work into smaller jobs and processing them concurrently is the basic idea. Its foundation is the divide and conquer strategy.

As was mentioned, it has two phases. During the initial stage, a Map function is executed. The map function receives data sets and uses them to carry out a number of tasks, such as filtering [6].

sorting, generate key/value pairs as the output. The data is summarized and applied to the output key/value pairs produced in the first phase during what is known as the Reduce phase that follows. There is only one execution of the Reduce function for every key/value pair. The MapReduce model's fundamental principle is to increase the number of nodes in a cluster rather than boosting a single node's power and to execute the task in parallel on every cluster node.

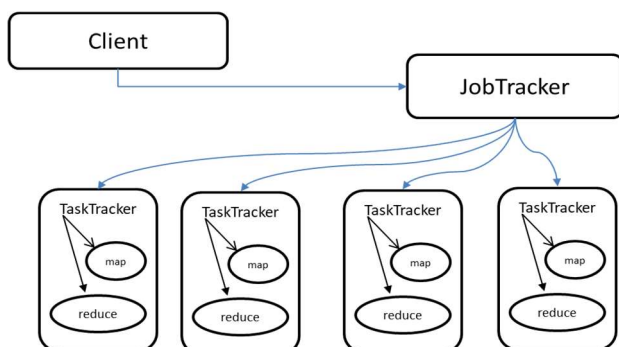


Fig 2:MapReduce and HDFS (Hadoop Distributed File System)

In a MapReduce cluster paradigm, two different kinds of nodes can be found. The Job Tracker node is the first, and Task Tracker is the second node. The primary purpose of

All of the tasks completed by Task Trackers are managed by Job Tracker. The Map and Reduce functions on the Task Tracker nodes are scheduled by the Job Tracker nodes. To do this, they employ a scheduling method.

The Task Tracker's primary purpose is to carry out the task. On the nodes that the Job Tracker has scheduled, this Task Tracker executes the Map and Reduce function. It then returns the outcome to the Job Tracker. [13].

Initially, the client sends the job to Hadoop in order to conduct the Map Reduce function. The Hadoop Job tracker then receives a notification that the job has been terminated. Next, this Job Tracker locates the node that is free to perform the task. A node or task tracker that is prepared to do the task is assigned the job by job tracker. A queue is used to hold jobs if there are no free nodes. A task is assigned when a node becomes available. Additionally, the job that the Job Tracker assigns in completed at the node.

IV. BIGDATA CHALLENGES

Big Data's heterogeneity, scale, timeliness, complexity, and privacy issues obstruct advancement at every stage of the data-value creation process. The issues arise immediately during the acquisition of data, when the deluge of data forces us to make decisions—currently on an as-needed basis—about which data to keep and which to discard, as well as how to consistently store the data we do keep with the appropriate metadata. A significant amount of data available today is not natively structured; for instance, blogs and tweets are disorganized text fragments, whereas images and videos are organized for both display and storage. However, do not search for semantic content.

One of the primary tests is converting such material into an organized format for thereafter analysis. When data can be connected to other data, its value soars. Data integration is therefore a significant value generator. Today, most data is created directly in digital format; this presents both an opportunity and a challenge: we can influence the development of data to make it easier to link later and link it automatically before it is created. [16]

Other core difficulties include recovery, modelling, data analysis, and organizing. Due to the intricacy of the data that needs to be processed as well as the limited scalability of the original algorithms, data analysis is a glaring bottleneck in many applications. In conclusion, the crucial aspect of obtaining actionable knowledge is the non-technical domain experts' explanation and display of the findings.

A. Volume of data

The amount of data, particularly machine-generated data, is expanding at an exponential rate every year as a result of fresh data sources emerging. For example, 800,000 petabytes (PB) of data were stored worldwide in 2000. By 2020, IBM predicts that it will have reached 35 zettabytes (ZB). Social media is important; for example, Twitter produces more than 7 terabytes (TB) of data per day. Facebook, 10 TB. Mobile gadgets are also quite significant.[

B Big data skills are in short supply

Data scientists are already in short supply in the market. One aspect of this is the lack of workers with the skills necessary to handle massive data sets and volumes of data. To help make sense of the data streams that are streaming into their companies, businesses need to combine the right people. This involves a skill set that even the majority of data scientists lack: applying prophetic analytics to large data. [14].

C. Data Security issues

Privacy, unequal internet access, legal and security concerns are major difficulties in public affairs; managers and policymakers in these fields

should endeavour to overcome these obstacles. Nonetheless, public managers and legislators typically operate within the confines of a restricted [10].

financial constraints, a variety of stakeholders, and condensed timelines for extracting knowledge from massive amounts of data (Mergel, Rethemeyer, and Isett, 2016; Grover and Kar, 2017) [11].

Watson (2019) discussed a few big data security concerns and offered some advice on how to lower the dangers related to large data security. Big data's inherent security concerns stem from its diverse origins, some of which might have lax protection, as well as its enormous volumes and range of forms. Therefore, any security lapses could impact several businesses.

can cause monetary losses; as a result, proper steps should be done to lower such big data security risks [15].

Organizations should keep an eye on their data sources and utilize end-to-end encryption to keep anyone from seeing the data while it's in transit. Businesses should also verify with their cloud providers because many of them do not encrypt data due to the amount of data that is transferred constantly as well as the fact that data flow is slowed down by encryption and decryption. [20]

D. Data privacy issues

User data collection may give rise to privacy issues since the procedure may change the context and semantics of the data, resulting in flawed and ineffective rules (Ali et al., 2016).

Data security and privacy are potential issues with big data, as demonstrated by Legal. (2017). This is because sensitive information, such bank account details and medical records, is frequently included in big data applications and is not suitable for standard data transfer protocols. Therefore, before adopting any protocol for sharing information, data security and privacy must be taken into account. It's common knowledge that the inclusion of delicate information and the necessity of access control or certification present difficulties; however, safe certification procedures are still difficult to put into place, and anonymization techniques reduce data [4]

The two parts of big data privacy are as follows: the first is the protection of users' personal information, including their habits, hobbies, and physical characteristics, that they may not be aware of or find easily accessible. The individual's privacy is the second factor.

Even when the user granted permission, data may leak during storage, transport, and use. As an illustration, Facebook is currently regarded as a big data corporation with the greatest amount of SNS (social networking service) data. However, using an information-gathering tool, some researchers were able to obtain data from Facebook users' public pages who had not changed their privacy settings (Chen, M., Mao, S., and Liu, Y., 2014) [1].

V. PRESENT DIFFICULTIES AND FUTURE RESEARCH VIEWS PRIVACY AND SECURITY OF BIG DATA

Big data security and privacy are surrounded by a number of unsolved concerns. and potential directions for further research. A list of some of the most significant ones is provide below. Preserving Privacy through Social Network Mining.

Because popular Web social networks like Facebook and are readily available to the public, social network data are actually the most trustworthy sources of real-life big data.

Twitter. While mining such data is of great interest in this case, the actual impact of these jobs is frequently limited by the need for privacy and security.

Challenges:

- Recognizing sensitive information: Depending on the situation and the individual, it might be difficult to define whether information is "private" or sensitive. In addition to explicit identifiers like names, it also includes social relationships, behavior patterns, and implied characteristics.
-

Striking a balance between utility and privacy: methods that completely anonymize data frequently make it useless for insightful analysis. Finding a balance between privacy and usefulness: and safeguarding privacy.

- Preventing "de-anonymization" attacks: An attacker having access to other information, such as a person's location or interests, can re-identify even anonymised data.[19]

Methods:

- Data minimization: Reducing the attack surface and potential privacy hazards can be achieved by gathering and retaining just the information that is absolutely required for the intended purpose.
- Encryption and anonymization: While anonymization techniques like k-anonymity mask individual identities, techniques like homomorphic encryption enable calculations on encrypted data without the need for decryption.[24]
- Differential privacy: In order to guarantee that results are accurate for the group while preventing individual identity, this method adds controlled noise to the data.[10]

Additional thoughts

- Accountability and transparency: Companies that engage in social network mining ought to be answerable to users for their data privacy policies and open about their methods [29].
- Regulatory frameworks: In the case of social network data, laws and regulations can aid in establishing precise boundaries and protections for individual privacy.

Security Challenges of (Big) Outsourced Databases

- Databases are frequently outsourced in cloud infrastructures using the well-known DaaS (Database as a Service) model . This provides
- lead to extremely dangerous security problems since query processing techniques can quickly access private information and identify security breaches.[30]

Visibility and Access Control:

- Limited Visibility: There is an inherent trust risk when you provide the vendor and their staff access to your data. To keep an eye on activities and ensure accountability, regular audits and access records are crucial.
- Data Partitioning and Segregation: Limiting access to critical information and dividing up sensitive data can help minimize the impact of a breach.
- User Activity Monitoring: To keep tabs on unauthorized access attempts, data alterations, and questionable activity in the outsourced database, put in place extensive user activity monitoring.

Other Difficulties:

- Data encryption: Although encryption is essential, maintaining encryption keys and making sure that they are rotated properly become more difficult in a system that is outsourced.
- Backups and Disaster Recovery: To guarantee data availability and resilience in the event of outages or emergencies, confirm the vendor's backup and disaster recovery procedures.

- Reporting and Auditing: Get reports on system performance, potential vulnerabilities, and access records, as well as conduct routine audits of the vendor's security procedures.[28]

VI. PRIVACY -PRISERVING BIG DTA ANALYTICS

Big data are significant because they are cherished information sources that can be applied to forecasting and decision-making (e.g., Web promotion). Analytics are used to this purpose, however because they handle enormous amounts of (big) data, the privacy of target data sets is compromised, which exposes the underlying knowledge discovery process to difficult research problems.[29]

Maintaining Privacy in Large-Scale Data Analytics:

- Data minimization reduces the amount of sensitive information exposed by gathering and storing only the data required for the particular analytics objective.
- De-identification and Anonymization: While maintaining some data utility, methods such as differential privacy and k-anonymity can mask individual identities.
- Secure Multi-party Computation protects personal information by facilitating cooperative analysis without requiring the release of raw data.
- Homomorphic Encryption: Preserves privacy while permitting analysis and allowing computations on encrypted data.
- Consent and Control of the User: Give people the power to manage their data, decide what they share, and opt-in or out of particular analytics operations.[27]

Extra Things to Think About:

- Accountability and Transparency: Companies that employ big data analytics should answer to users for data privacy and be open and honest about their procedures.
- Regulatory Frameworks: Organizations managing personal data are subject to rights and obligations regarding data privacy, which are established by laws and regulations such as the CCPA and GDPR.
- Technologies that Promote Privacy: Research into new technologies such as blockchain and federated learning is ongoing in order to facilitate safe and private big data analysis.[25].

VII. CRYPTOGRAPHY-BASED DATA-CENTRIC PROTECTION

To limit the visibility of data to people, organizations, and systems, there are two main methods. Limiting access to underlying systems, such as operating systems or hypervisors, is the first step. The latter is summarizing the data itself within a secure envelope thanks to cryptography. Greater attack surface is offered by the first strategy, sometimes known as the system-based strategy. Numerous attacks, such as buffer overflow and privilege escalation, can get around access control systems and access data. Encrypting data from beginning to end results in a considerably smaller, more defined attack surface.[14]

Although it is an impossible task, it can extract secret keys and is susceptible to covert side-channel attacks. Numerous risks connected to access control methods enforced by cryptography. Using encryption are: The matching plaintext data should not be recognizable to the opponent, even if he must select between an incorrect and correct plain text when seeing the encrypted text. An adversary using a cryptographic protocol to enable searching and filtering encrypted data should only be able to deduce the appropriate predicate—whether satisfied or not—and nothing else about the

encrypted data. Additionally, since the claimed source of the data may not be genuine, the cryptographic protocol must prevent adversaries from forging data, which could compromise the integrity of the data.

VIII. MONITORING SECURITY IN REAL TIME

In the big data analysis scenario, real-time security monitoring has proven to be a persistent difficulty, mostly because of the volume of warnings that security equipment create. These warnings, which may or may not be related to one another, produce a lot of false positives and, since humans are unable to process so many of them quickly, they are either clicked away or ignored [9]. For security monitoring to be effective, the Big Data platform or infrastructure must be intrinsically secure. Big Data infrastructure dangers include eavesdropping on the line, (web) application attacks, and rogue administrator access to apps or nodes [22].

Infrastructure, which is really an ecosystem of various components, requires careful consideration of both the security of individual components and the interaction of those components. If a Hadoop cluster is operating in a public cloud. [30]

It is necessary to take into account the security of the public cloud, which is an ecosystem of interconnected network, storage, and computer components. It is necessary to take into account the security of the Hadoop cluster, the security of the nodes, the interconnection between the nodes, and the security of the data kept on a node. It is also necessary to take into account the monitoring application's security, including any relevant correlation rules that adhere to secure coding guidelines. It is also necessary to consider the security of the input source where the data originates.

IX. CONCLUSION

We are living in a data-driven world. These days, all decisions are made in light of the available facts. The idea of data analysis—a vast amount of data can be gathered and examined to identify trends and help with decision-making—will soon serve as the cornerstone of business growth and competition, boosting output and raising the Caliber of goods and services. Both organized and unstructured data are being produced in massive quantities. In order to gain the knowledge necessary to make the best decisions, businesses are now searching for innovative approaches to use data to alter their operations. Consequently, data and its analysis have changed more than only the data industry. but also elevated employee analysts to new heights.

When big data systems are accessible, requirements for distribution that may appear absurd today will soon become standard practice. We discover how to take use of them. Systems comparable to Facebook and Google would have sounded like science fiction not so many years ago. It was a stretch at the time for banking and aviation systems to process 100 transactions per second. Consequently, a survey of the literature was conducted in order to generate an analysis of the big data analytics principles that are now being

Large data necessitates higher security and privacy standards during the collection, storing, processing, and sharing of data. In this work, we reviewed research on privacy and security for large data. In contrast. The literature suggests that network traffic be encrypted using appropriate standards, that device access be inspected, that employees have access rights to systems, that anonymized data be analyzed, that communications be made over secure channels to prevent leaks, and that the network be watched over for threats. The three primary areas that require further in-depth discussion in the next years are big data privacy, safety, and security. To achieve correct results, new methods, tools, and technologies for human-computer interaction must be invented, or current ones must be enhanced.

X. ACKNOWLEDGMENT

Big data analytics is a potent instrument that is revolutionizing our way of working, living, and making decisions. To find hidden patterns, trends, and insights, it involves gathering, storing, analyzing, and displaying vast amounts of data.

Imagine it like sorting through a mound of sand in search of the buried jewels. We can extract useful information from large datasets using big data analytics that would be impossible to process by hand.

The following are some of the salient features of big data analytics:

Volume: Petabytes or even zettabytes of data are frequently involved in big data operations. Numerous sources, including social media, sensors, transactions, and online logs, may provide this data.

Velocity: The high velocity of big data is frequently used to describe it.

Variety: Unstructured, semi-structured, and structured big data are all possible. Spreadsheet data and other ordered, easily comprehensible data types are examples of structured data. Less structured data, such that found in customer emails, is called semi-structured data. The hardest kind of data to handle is unstructured data because it frequently consists of text and lacks a defined format.

Veracity: Big data's precision and dependability are essential. Companies need to be able to depend on the precision and error-free nature of the data they are examining.

Uses of big data analytics are numerous and span many different sectors. Here are few instances:

Business: Companies may enhance marketing efforts, streamline processes, and control risk by utilizing big data analytics.

Healthcare: Analytics using big data can be applied by hospitals and other healthcare providers to detect fraud, enhance patient care, and create novel medicines.

Government: Tracking criminal activity, enhancing public safety, and optimizing resource allocation are all possible with the use of big data analytics.

Science: By analyzing huge databases of scientific data, scientists can leverage big data analytics to make new discoveries and advancements. Big data analytics holds great potential. As time goes on, the ability to assess and understand the data we generate will become increasingly important. Big data analytics is a useful tool that may help us address some of the most crucial problems facing the globe now and in helping us make better decisions for the future.

REFERENCES

- [1] Sun, Zhaohao, Kenneth David Strang, and Francisca Pambel. "Privacy and security in the big data paradigm." *Journal of computer information systems* (2018).
- [2] Bertino E, Ferrari E. Big data security and privacy. In *A comprehensive guide through the Italian database research over the last 25 years* 2017 May 31 (pp. 425-439). Cham: Springer International Publishing.
- [3] Terzi, D. S., Terzi, R., & Sagioglu, S. (2015, December). A survey on security and privacy issues in big data. In *2015 10th International Conference for Internet Technology and Secured Transactions (ICITST)* (pp. 202-207). IEEE.
- [4] Moura, José, and Carlos Serrão. "Security and privacy issues of big data." *Handbook of research on trends and future directions in big data and web intelligence*. IGI Global, 2015. 20-52.

- [5] Bao, Rongxin, Zhikui Chen, and Mohammad S. Obaidat. "Challenges and techniques in Big data security and privacy: A review." *Security and Privacy* 1.4 (2018): e13.
- [6] Singh, Manbir, Malka N. Halgamuge, Gullu Ekici, and Charitha S. Jayasekara. "A review on security and privacy challenges of big data." *Cognitive Computing for Large-scale Data Systems Over IoT: Frameworks, Tools and Applications* (2018): 175-200.
- [7] Khanan, A., Abdullah, S., Mohamed, A. H. H., Mehmood, A., & Ariffin, K. A. Z. (2019). Big data security and privacy concerns: a review. In *Smart Technologies and Innovation for a Sustainable Future: Proceedings of the 1st American University in the Emirates International Research Conference—Dubai, UAE 2017* (pp. 55-61). Springer International Publishing.
- [8] Tiwari, Archit, Nikhil Sharma, Ila Kaushik, and Ratik Tiwari. "Privacy issues & security techniques in big data." In *2019 International Conference on Computing, Communication, and Intelligent Systems (ICCCIS)*, pp. 51-56. IEEE, 2019.
- [9] Xu, Lei, Chunxiao Jiang, Jian Wang, Jian Yuan, and Yong Ren. "Information security in big data: privacy and data mining." *Ieee Access* 2 (2014): 1149-1176.
- [10] Joshi, N. and Kadhiwala, B., 2017, April. Big data security and privacy issues—A survey. In *2017 Innovations in Power and Advanced Computing Technologies (i-PACT)* (pp. 1-5). IEEE.
- [11] Jha, Anupama, Meenu Dave, and Supriya Madan. "Big data security and privacy: A review on issues, challenges and privacy preserving methods." *International Journal of Computer Applications* 975 (2017): 8887.
- [12] Manikandakumar, M., & Ramanujam, E. (2018). Security and Privacy Challenges in Big Data Environment. In *Handbook of Research on Network Forensics and Analysis Techniques* (pp. 315-325). IGI Global.
- [13] Strang, Kenneth David, and Zhaohao Sun. "Big data paradigm: what is the status of privacy and security?." *Annals of Data Science* 4 (2017): 1-17.
- [14] Rad, B.B., Akbarzadeh, N., Ataci, P. and Khakbiz, Y., 2016. Security and privacy challenges in big data era. *International Journal of Control Theory and Applications*, 9(43), pp.437-448.
- [15] Ogbuke, Nnamdi Johnson, Yahaya Y. Yusuf, Kovvuri Dharma, and Burcu A. Mercangoz. "Big data supply chain analytics: ethical, privacy and security challenges posed to business, industries and society." *Production Planning & Control* 33, no. 2-3 (2022): 123-137.
- [16] Memon, Mashooque Ahmed, et al. "Big data analytics and its applications." *arXiv preprint arXiv:1710.04135* (2017).
- [17] LaValle, S., Lesser, E., Shockley, R., Hopkins, M. S., & Kruschwitz, N. (2010). Big data, analytics and the path from insights to value. *MIT sloan management review*.
- [18] Sagioglu, S. and Sinanc, D., 2013, May. Big data: A review. In *2013 international conference on collaboration technologies and systems (CTS)* (pp. 42-47). IEEE.
- [19] Nguyen, T., Li, Z.H.O.U., Spiegler, V., Ieromonachou, P. and Lin, Y., 2018. Big data analytics in supply chain management: A state-of-the-art literature review. *Computers & operations research*, 98, pp.254-264.
- [20] Saggi, M. K., & Jain, S. (2018). A survey towards an integration of big data analytics to big insights for value-creation. *Information Processing & Management*, 54(5), 758-790.
- [21] Ghazal, A., Rabl, T., Hu, M., Raab, F., Poess, M., Crolotte, A., & Jacobsen, H. A. (2013, June). Bigbench: Towards an



- industry standard benchmark for big data analytics. In *Proceedings of the 2013 ACM SIGMOD international conference on Management of data* (pp. 1197-1208).
- [22] Mishra, A. D., & Singh, Y. B. (2016, April). Big data analytics for security and privacy challenges. In *2016 international conference on computing, communication and automation (ICCCA)* (pp. 50-53). IEEE.
- [23] Mehta, Brijesh B., Udai Pratap Rao, Nikhil Kumar, and Sravan Kumar Gadekula. "Towards privacy preserving big data analytics." In *Proceedings of the 2016 Sixth Int. Conf. Advanced Computing and Communication Technologies, Ser. ACCT-2016, Rohtak, India: Research Publishing*, pp. 28-35. 2016.
- [24] Hussein, Abou_el_ela Abdou. "Fifty-six big data V's characteristics and proposed strategies to overcome security and privacy challenges (BD2)." *Journal of Information Security* 11, no. 4 (2020): 304-328.
- [25] Shoji, Nobubele A., and Jabu Mtsweni. "Big data privacy and security: A systematic analysis of current and future challenges." In *Proceedings of the 11th International Conference on Cyber Warfare and Security*, pp. 296-303. 2016.
- [26] Rajamäki, J., & Simola, J. (2019, July). How to apply privacy by design in OSINT and big data analytics. In *ECCWS 2019 18th European Conference on Cyber Warfare and Security* (p. 364). Academic Conferences and publishing limited.
- [27] Gosain, A. and Chugh, N., 2014. Privacy preservation in big data. *International journal of computer applications*, 100(17).
- [28] Kamakshi, P. (2014). Survey on big data and related privacy issues. *International Journal of Research in Engineering and Technology*, 3(12), 68-70.
- [29] Gundu, T. (2019). Big Data, Big Security, and Privacy Risks. *Journal of Information Warfare*, 18(2), 15-30.
- [30] Manogaran, Gunasekaran, et al. "Big data security intelligence for healthcare industry 4.0." *Cybersecurity for Industry 4.0: Analysis for Design and Manufacturing* (2017): 103-126.