



A GENERALIZE OVERVIEW ADAPTION OF DIGITAL FORENSICS PROCESS

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Abstract:

Digital forensics is defined as the process of conserving definitive retrieving and chronicle computer evidence for adoption by a court. It is the science of finding evidence in digital media such as computers, mobile phones, servers or networks. It equips the investigative team with the best techniques and tools to solve complex digital cases. Digital Forensics assists the forensic team in analyzing, decisive definitive and averting digital evidence found on

different types of electronic devices. This chapter covers various aspects of digital forensics, its process, function and variables affecting crime scene analysis. In this chapter, we are also emphasizing the advantages and disadvantages of digital forensics.

Keywords: Digital forensics, crime scene investigation.

1. Introduction

Digital forensics (sometimes known as virtual forensic technology) is a department of forensic technology encompassing the healing and research of material observed in virtual devices, frequently about pc crime. The term virtual forensics became at the beginning used as a synonym for laptop forensics however has increased to cover the investigation of all devices capable of storing digital information. With roots in the private computing revolution of the past due to the 1970s and early 1980s, the field developed haphazardly at some stage in the 1990s, and it was not till the early twenty-first century that national regulations emerged. The technical component of research is divided into numerous sub-branches, regarding the type of digital gadgets involved; computer forensics,

community forensics, forensic information evaluation, and mobile device forensics.

The regular forensic method encompasses the seizure, forensic imaging (acquisition) and analysis of digital media, and the manufacturing of a record into gathered proof.

As nicely as figuring out direct evidence of a crime, virtual forensics can be used to attribute proof to unique suspects, verify alibis or statements, decide the intent, perceive sources or authenticate documents.

Literature Review:

An efficient method for detection and localization of anomalies in videos is depicted in this paper. Using fully convolutional neural networks (FCNs) and temporal data, a pre-trained supervised FCN is transferred into an unsupervised FCN ensuring the detection of (global) anomalies in scenes. Experimental results on two benchmarks suggest that detection and localization of the proposed method outperforms existing methods in terms of accuracy and processing speed. Besides this, it is a solution for overcoming limitations in training samples used for learning a complete CNN. This method enables us to run a deep learning-based method at a speed of about 370 fps. Altogether, the proposed method is both fast and accurate for anomaly detection in video data [1, 4].

| Sr. No. | Author | Year | Method used | Methodology used | Conclusion |
|---------|--|------|--|--|--|
| 01 | Li et. al. [1] | 2014 | Convolutional Neural Networks | Tracking with deep track, CUDA-PTX | It employs a CNN architecture and a structural loss function that handles multiple input cues and class-specific tracking. |
| 02 | Hrishikesh Bhaumik, Siddhartha Bhattacharyya, Mausumi Das Natha, Susanta Chakraborty [24] (ELSEVIER) | 2016 | | Hybrid soft computing techniques | The systems have increased in robustness, efficiency and effectiveness as compared to earlier used traditional approaches. It also helped to reduce user interaction and manual annotation to a great extent. |
| 03 | Shi-Zhe Chen, Chun-Chao Guo, Jian-Huang La [22] | 2016 | Traditional and CNN-based method | Deep cnn | Formulate the person re-identification task as a learning-to-rank problem. Extensive experimental results clearly demonstrate the effectiveness of our proposed approach |
| 04 | Antonio C. Nazare Jr., William Robson Schwartz [30] (ELSEVIER) | 2016 | Smart Surveillance Framework (SSF) | Framework 1 user modules 2 SSF kernel | Smart Surveillance Framework (SSF) allows the simultaneous execution of multiple user modules that can be developed independently since they have communication and synchronization through a shared memory, which contributes to the scalability and flexibility. |
| 05 | Shao et. al. [14] IEEE transaction | 2016 | Big Data | LDA, KNN, fuzzy clustering | By a combination of snapshot images, original surveillance videos and unusual events, valuable clues can be found out much easier, which thus helps the police boost their investigation efficiency. |
| 06 | Revathi and Kumar [5] SPRINGER | 2016 | Deep learning-based anomaly detection (DLAD) | Background Estimation (BE) Module, an Object Segmentation (OS) Module, a Feature | Better error rate of 0.75% and precision of 85% |

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|----|--|------|--|---|---|
| | | | | Extraction (FE)Module, and an Activity Recognition (AR) Module | |
| 07 | Sabokroua et. al. [4] | 2017 | Deep learning - Fully Convolutional neural network (FCN) | AlexNet | Proposed method is both fast and accurate for anomaly detection in video. |
| 08 | Chang and Tay [2] | 2017 | Spatiotemporal architecture | ConvLSTM | Detects Abnormal events but it may produce more false alarms as compared to other evnts |
| 09 | Thanh Vu et. al. [3] | 2017 | Restricted Boltzmann machine (RBM) | ConvAE | RBM's are trained to capture different image statistics localized at different regions. This framework is readily generalized to a more powerful deep unsupervised abnormality detection framework. |
| 10 | Sinha et. al. [11] IEEE, International Conference on Intelligent Computing and Control (I2C2) | 2017 | Deep Learning | Convolutional Neural Networks(CNN), Restricted Boltzmann,Machines (RBM) and Autoencoders, Recurrent Neural Networks and Extreme Learning. | This paper summarizes and gives an idea to the new researchers to explore more in the vast yet young area of Deep learning. |
| 11 | Muhammad et. al. [12] IEEE Access | 2018 | CNN | GoogleNet | This paper improved the flame detection accuracy, but the number of false alarms is still high and more research is required. |
| 12 | Bajestani et. al.[7] | 2018 | Faster R-CNN | Object Detection | This method improves the true positive with the tradeoff of trivial false positive. |
| 13 | Manisha Kaushal a, Baljit S. Khehra b, Akashdeep Sharma [28] (ELSEVIER) | 2018 | Soft computing | Neural network Fuzzy Logic Neuro-fuzzy Hybrid | Various soft computing based approaches for moving object detection and tracking in videos. Article provides various techniques along which scope, pros, cons and the limitation associated with each of them |
| 14 | Huang et. al.[6] SCOPUS- | 2018 | Restricted Boltzmann machine (RBM) | SCL | It improves the average accuracy of multimodal deep representation by |

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|----|---|------|----------------------------------|--|---|
| | HINDAWI | | | | 2.65% |
| 15 | Nasir et. al. [15] ELSEVIER | 2018 | Fog Computing | | |
| 16 | Muhammad et. al.[21] IEEE | 2018 | Video summarization | a fast probabilistic and lightweight algorithm | Experimental results verify the efficiency, security, and robustness of proposed algorithm compared to other image encryption methods. |
| 17 | Munir et. al. [9] IEEE Access | 2018 | Convolutional Neural Networks | DeepAnT | Evaluation of DeepAnT on 10 different data sets comprising of 433 time series in total and provided a detailed comparison with 15 state-of-the-art anomaly detection methods. |
| 18 | Raahat Devender Singh, Naveen Aggarwal [23] (SPRINGER) | 2018 | | Passive-blind technique Inter-frame forgery detection | Presents a repository of information regarding the kinds of tamper attacks a video can suffer from and a comprehensive source of references for the passive-blind techniques proposed for detecting attacks |
| 19 | Sultani et. al. [8] IEEE Xplore | 2018 | Multiple instance learning (MIL) | Deep MIL Ranking Model, binary SVM classifier | A new large-scale anomaly dataset consisting of a variety of real world anomalies is introduced. |
| 20 | Alberto Castillo, Siham Tabik, Francisco P´erez, Roberto Olmos, and Francisco Herrera [27] (ELSEVIER) | 2018 | Deep learning | DaCoLT- a brightness guided preprocessing approach | This paper proposed a brightness guided preprocessing approach. DaCoLT model shows a high potential even in low quality videos and provides satisfactory results as an automatic alarm system. |
| 21 | Bouindour et. al. [10] Applied sciences MDPI | 2019 | Convolutional Neural Networks | Matlab | This method is robust, takes into account rare normal events present in the training phase. Besides, it can be incorporated in online CCTV. |
| 22 | James A.D. Camerona, Patrick Savoiea , Mary E. Kayea , Erik J. | 2019 | CNN based algorithms | GoogLeNet , AlexNet VGG16, ResNet50, SSD | This work has highlighted some of the practical challenges of designing a processing system for a |

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| | Scheme [26] (ELSEVIER) | | | | CNN-based automated surveillance system using off-the-shelf hardware and open-source algorithms. |
| 23 | Jianyu Xiao , Shancang Li , Qingliang Xu (IEEE) | 2019 | Deep learning | Object detection and tracking | Proposed a framework for video based digital forensics investigation, useful for anti-crime or fast response when crime activities or behaviors are detected |
| 24 | Summra Saleema, Aniqa Dilawari , Ghani Khana, Razi Iqbal c, Shaohua Wand, Tariq Umer [29] (ELSEVIER) | 2019 | Deep Convolution Neural Networks (CNN). | Feats-rich model encodes the visual contents to visual and facial features using CNN architecture. | It is a framework for generating multi-line textual descriptions for video captioning. Feats-rich model extends feature matrix to visual (2-D and 3-D) and facial features. Spatio-temporal characteristics are encompassed by employing deep neural networks. |
| 25 | Sreenu and Durai [13] SPRINGER OPEN ACCESS | 2019 | Big Data | ImageNet2012, PASCAL VOC, Frames Labeled In Cinema (FLIC), Leeds Sports Pose (LSP) | Methods analyzing crowd behavior were discussed. |

Goals of PC crime scene investigation

Here are the fundamental destinations of utilizing Computer crime scene investigation:

- It assists with recuperating, dissecting, and protecting PC and related materials in such a way, that it causes the examination office to introduce them as proof in an official courtroom.
- It assists with proposing the rationale behind the wrongdoing and personality of the fundamental guilty party.
- Structuring systems at a speculated wrongdoing scene which causes you to guarantee that the advanced proof acquired isn't debased.
- Information procurement and duplication: Recovering erased records and erased parcels from advanced media to extricate the proof and approve them.
- Causes you to recognize the proof rapidly and permits you to evaluate the potential effect of the malevolent

movement on the person in question delivering a PC measurable report which offers a total report on the examination procedure.

- Protecting the proof by following the chain of authority.

2. Process of Digital Forensics

Digital forensics entails the following steps that are explained below:

- Identification: This is the first step in forensic medicine. The identification process largely consists of such things as what evidence exists, where it is stored and ultimately how far away it is stored. Electronic data carriers can be personal computers, mobile phones, PDAs, etc.
- Preservation: It consists of stopping humans from using virtual devices so that digital evidence isn't always tampered with.
- Analysis: In this step, investigation marketers reconstruct fragments of facts and draw conclusions based totally on evidence found. However, it'd take numerous

iterations of the exam to support a particular crime theory.

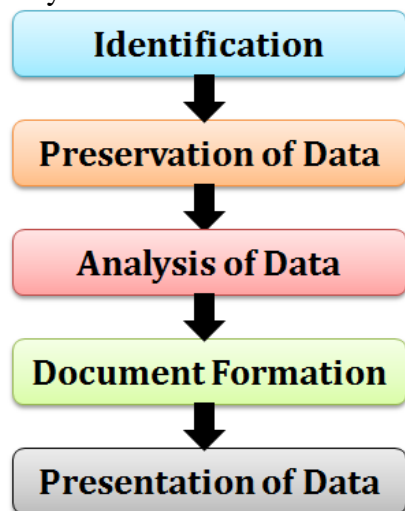


Figure 1: Process of digital forensics

➤ **Documentation:** In this technique, a report of all of the visible statistics sought to be created. It can be used to recreate and visualize a crime scene. This includes proper documentation of the crime scene, as well as photographing, sketching and mapping the crime scene.

➤ **Presentation:** In this final step, the technique of summarization and clarification of conclusions is done.

3. Difficulties looked through Digital Forensics

Here, are full-size problems looked at with the aid of the Digital Forensic:

- The enlargement of PCs and extensive utilization of internet access
- Simple accessibility of hacking instruments
- Absence of physical proof makes indictment troublesome.
- The large degree of more room in Terabytes makes these investigations troublesome.
- Any innovative changes require an update or adjustments to arrangements.

4. Model Uses of Digital Forensics

In late time, enterprise associations have applied automatic prison sciences in the following a sort of cases:

- Licensed innovation robbery
- Mechanical reconnaissance
- Business questions
- Misrepresentation examinations

- Improper usage of the Internet and email inside the working environment
- Fabrications related issues

5. Advantages of Digital criminal sciences

Here, are some advantages of Digital criminal sciences

- To guarantee the honesty of the PC framework.
- To deliver evidence in the court, this can spark off the field of the guilty party.
- It encourages the agencies to catch massive information if their PC frameworks or systems are undermined.
- Proficiently finds cybercriminals from anywhere on the planet.
- Assists with ensuring the association's cash and significant time.
- Permits to concentrate, process, and decipher the verifiable proof, so it demonstrates the cybercriminal activity inside the court.

6. Detriments of Digital Forensics

Here, are sizable cons/downsides of using Digital Forensic

- Advanced proof stated in court. In any case, its miles have to be proven that there's no altering.
- Creating electronic records and placing away them is an amazingly luxurious undertaking.
- Lawful specialists ought to have huge PC data.
- Need to create authentic and persuading evidence.
- In the occasion that the instrument applied for automatic criminology isn't always as per indicated gauges, at that point in the legit courtroom, the evidence may be opposed via equity.
- Absence of specialized information by way of the examining reputable probably might not offer the proper outcome.

7. Conclusion

The preservation, recognition, extraction, and recording of evidence that can be used in a court of law are known as digital forensics. Identification, preservation, analysis, documentation, and presentation make up the digital forensics process. Disk forensics, network forensics, wireless forensics, database

forensics, malware forensics, email forensics, memory forensics, and other kinds of digital forensics exist. Thus, we draw the conclusion that 1) intellectual property theft and 2) industrial espionage are instances where digital forensic science can be used. 3) Conflicts at work; 4) fraud inquiries.

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