A Novel Uniminer frameslog for data mining

Ms.Ashwini A.Kale¹, Prof. S.K.Korde²

¹ PG Student, Computer Department, PREC, Maharashtra, India. ² Prof. S. K. Korde, Computer Department, PREC, Maharashtra, India.

ABSTRACT

We analysis the background and state-of-the-art of large data. We first introduce the general background of big data and appraisal related machineries. Now days devices and Smartphones produce huge data streams in universal and ubiquitous environments. Usually, big data systems gather all the data at a central data processing system (DPS). These data storage tower are additional analyzed to create approximated patterns for different claim areas. This attitude has one-sided value (i.e. at big data treating end) but two main side-possessions that main towards user's displeasure and added computational costs. These effects are since all the data is being collected at central DPS, user privacy is compromised and the gathering of vast raw data streams, most of which could be unrelated, at dominant systems required more **computational** and packing resources hence rises the overall operative cost. Possession in view these limitations, we are proposing a unified structure that balances between value and cost of big data system with improved user satisfaction. That's why we applying mining algorithm on local devices,We studied different data mining organisms and planned a new framework, named as UniMiner, to impact data mining systems with smartphones, and cloud computing technologies. The idea of UniMiner is the scalability of data mining method differentiates UniMiner from existing systems by enabling maximum data processing near data resources.

Keywords — Data mining, cloud computing, smartphone, Devices, Internet of thing etc

1. INTRODUCTION

Now a days we are using great quantity strategies, smartphones, Internet-of-Things, and computing technologies are key-enablers for big data analysis. That all devices have the different uses if we are using for data mining purpose then its basic uses not that much perfect or easily possible for us. That's why we are proposing a unified frameslog that can enable the users to mine their patterns on their local devices and divest to other DPSs in case of 1) resource shortage or/and for data and knowledge distribution to big data systems. we are proposing a unified framework that can facilitate the users to mine their patterns on their local devices and offload to other DPSs in case of 1) resource scarcity or/and for data and knowledge sharing to big data systems. In addition, the proposed framework will enable to handle local raw data streams with additional features of filtering and communicating only relevant data streams to big data systems.

Another perspective of this study is to provide an effective 4P (Private, Personal, Portable, and Powerful) strategy for KD in big data environments. The motivation is to propose a unifide frameslog that 1) preserve the privacy of the user in big data environments, 2) make sure the user can benefit from personal data patterns 3) ensure data stream mining and knowledge fusion everywhere/anywhere and 4)

provides sufficient resources to provide knowledge discovery to users at one end and big data environment on the other side.

This motivation implementation purpose we are using the IOT, Cloudcomputing , data mining techquine.

Cloud computing is internet-based computing, where shared servers provide computing power, storage, development platforms or software to computers and other devices on demand. Second is **IOT**:Extending the current Internet and providing connection, communication, and inter-networking between devices and physical objects, or "Things," is a growing trend that is often referred to as the *Internet of Things*."The technologies and solutions that enable integration of real world data and services into the current information networking technologies are often described under the umbrella term of the Internet of Things (IoT)". Last one is **Data Mining:**Which is used for fetching the exact data form huge amount of data source.

2. LITERATURE SERVEY

Chee sun Liew [1] which paper has implemented the framework which has reduce the cost. MIT technology review [2] informed that 95% of the data generated these days rests unanalyzed. The operative knowledge discovery (KD) and managing techniques are vital to expose actionable hidden patterns from this huge amount of data. Moreover, big data systems generate near result.M.ChenS.Mao [3] patterns useful for individuals, are still needed. Conversely, analyzing data near the source in a modified way is another option to bind big data with personal patterns. J.Gama and M.M.Gaber [4] (energy, computations, imagining, and storage) and absence of wholly scalable KD platforms are big hurdles in this esteem. OMM [5] enable LA by acting all KD process inside mobile phones.

StreamAR [6] procedure data streams locally privileged mobile phones and expose action patterns from on-board accelerometers. PDM [7], a multi-agent based data mining systems, works in collaborative environments. Different intelligent mediators are used for resource-findings, data mining, and conclusion making in the PDM workflow to competently utilize the resources in collaborative situation. It performs LA as well as utilizes Fog cloud services [8] to gain more computational powers. CARDAP proposed an effective data uploading strategy called Local Analytics + Smart Data Reduction + On-demand Sensing (LA-DR-OS). The purpose of LA-DR-OS is to perform LA and upload data to Fog when a significant change is detected. It should be noted that Fog computing is still at its initial stage providing cloud computing resources near data sources. Hence, there is an opportunity for creating privacy-preserving data mining systems in ubiquitous environments that support both big data mining and LA.

MobSafe [9] provides cloud based data mining services for forensic analysis in mobile applications. Cloud computing technologies provide massively parallel computing infrastructures to analyze big data but the user needs to pay for each computational cycle. Hence, LAs based system that support LA-DR-OS could become the key factors to reduce financial burden of using cloud based analytics.

We are proposing a unified framework called UniMiner to address the limitations of existing systems and provide a seamless data mining approach to meet analytic requirements in big data environments.

3. PROPOSED SYSTEM OVERVIEW

The proposed frameslog has included three layer which is Local analytics second collaborative analytics third one cloud enable analytics etc.

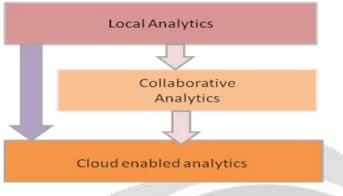


Fig. 1: Architecture of Frameslog

A. Local Analytics(LA)

The LA layer in UniMiner is supposed to be the main contributor for privacy and data reduction in big data environments. This layer is based on Personal Ecosystem (PE)i.e our particular device. A PE is based on four modules for 1) knowledge discovery, 2) knowledge management, 3) system management and 4) visualization.

B. Collaborative Analytics(CA)

This is the Second layer of the uniminer frameslog. Which layer is used for peer-2-peer Adhoc Network. When the local device not having that much memory then we can select the another network for sending data on cloud enabled Analytics layer via the peer-2-peer Adhoc.

Data processing flow for CA is bellow:

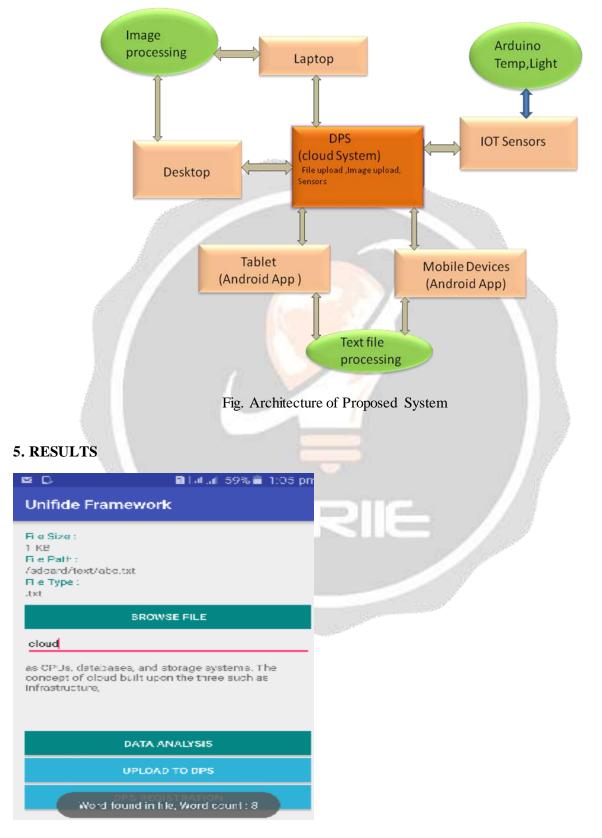
Start Collaboration
Local Device Discovery
Device Found if not then END
Ad-Network Formation
Agent-oriented KDP
DATA streams
Fogcloud uploadding
END

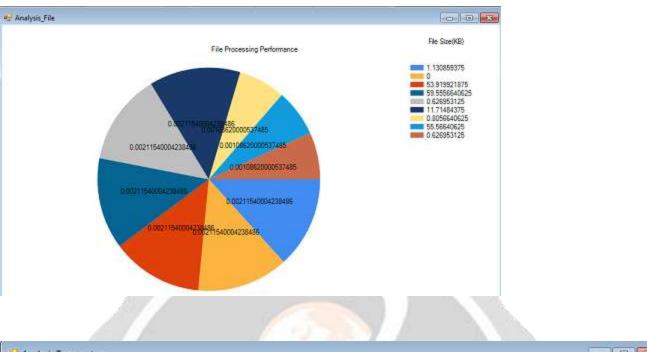
C. Cloud-enabled Analytics

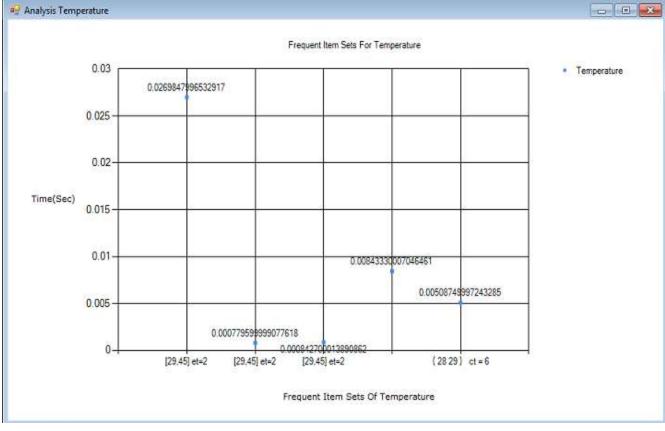
Fog cloud technology, recently introduced by Cisco, is the enablement of cloud services near data sources.

The basic purpose is to enable Internet of Things (IoTs) based infrastructures to actively participate in big data environments. Fog enables to connect with all types of devices, systems and 'things' that generate digital data. Moreover, the availability of computational power near data sources enable to process the data initially which lowers the computations and energy consumption at the central big data system. UniMiner utilizes Fog services to meet the extra-ordinary computational requirements of sensors' data

stream mining algorithms for processing continuous data streams. Additionally, the CA layer in UniMiner further reduces raw data streams hence reducing overall uploaded data in big data ecosystem.







Both above figure shows the final result on the Smartphone, desk system and Internet of things respectively.

We can see the first figure shows the analysis on the smart phone we choose the file if word find then we can forward on Cloud second figure shows different files taking how many space and time for the execution. And last one shows the temperature analysis with frequent item account.

6. CONCLUSION

The big data systems collect raw data stream from users in crowd sensing applications. A large number of data streams are not useful that puts a huge computational and financial burden on big data systems. In addition, the massive collection of data created privacy concerns that are needed to be addressed by big data systems. To this end, we proposed a three layer framework, called UniMiner, for data reduction in big data environments and provision of privacy preserving services to users. The framework is designed in such a way that user can execute whole KD process in local device. The local device has most of reduce the raw data on local base only that means the LA and CA layer has the done raw data reduction on this layer. In case of resourcescarcity, data mining tasks could be offloaded to other devices in proximity or utilize Fog cloud services. We are evaluating the different pattern mining algorithms.

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BIOGRAPHIES:

