



**IJIRCCCE**

e-ISSN: 2320-9801 | p-ISSN: 2320-9798



# INTERNATIONAL JOURNAL OF INNOVATIVE RESEARCH

IN COMPUTER & COMMUNICATION ENGINEERING

**Volume 9, Issue 9, September 2021**

**ISSN** INTERNATIONAL  
STANDARD  
SERIAL  
NUMBER  
INDIA

**Impact Factor: 7.542**



9940 572 462



6381 907 438



ijircce@gmail.com



www.ijircce.com

# Human Activity Patterns from Smart Home for Healthcare

Akshay Wannare, Rishikesh Sawant, Vrushal Sapkal, Geeta Navale

Department of Computer Science and Engineering, Sinhgad Institute of Technology and Science, Pune, India

**ABSTRACT:** Frequently, there is an ever-increasing migration of humans to urban areas. Healthcare offerings are one of the maximum hard factors this is significantly stricken by the sizeable influx of people to towns. Consequently, towns round the world are investing closely in virtual transformation with the intention to provide healthier ecosystem for people. In such transformation, millions of homes are being prepared with smart devices (e.g. Smart meters, sensors and so on.) which generate huge volumes of excellent-grained and indexical data that may be analysed to guide clever town offerings. Authors endorse a version that makes use of clever domestic huge records as a approach of studying and discovering human interest styles for healthcare programs. Authors also propose use of common pattern mining, cluster analysis and prediction to analyse strength utilization. Since humans' habits are usually identified by everyday exercises, coming across those workouts allows us to anomalous activities that suggest human being's difficulties in taking care of themselves. Authors propose the use of frequent pattern mining, cluster analysis and prediction to measure and analyze. The work also shows the results of identifying human activity patterns from appliance usage are presented in detail along with the accuracy of short and long term predictions.

## I. INTRODUCTION

Studies show that by means of year 2050, 66% of the population may be residing in city areas. The demand for fitness care assets could be substantially suffering from this large influx of human beings to metropolis centers. In such transformation, tens of millions of houses are being equipped with smart devices which generate massive volumes of pleasant-grained and indexical information that may be analyzed to aid fitness care offerings. Advancement of vast record mining technology, which offer approach of processing massive quantity of statistics for actionable insights, can useful resources in understanding how human beings move approximately their existence. For example, tracking the adjustments of appliance usage interior a clever home can be used to not directly determine the character's well being based on past information. Since humans habits are mostly identified by means of normal routines, coming across these exercises lets in us to understand anomalous sports that may indicate human being's difficulties in taking care for themselves. The underlying correlation among equipment usage inside the smart home can be used by health care programs. The statistics from clever meters are recursively mined inside the quantum/statistics slice of 24 hours, and the effects are maintained across successive mining sports. It uses the Bayesian community, a probabilistic graphical version, to expect the usage of more than one home equipment and family power consumption.

## II. LITERATURE REVIEW

In [1], ABDULSALAM YASSINE specified, Presentation of mechanism for sharing power consumption data in deregulated smart grids. Used the concept of differential privacy as an anonymity mean to minimize the leakage of information to specify the privacy level and the associated payment. The importance of privacy as a crucial element in designing balanced markets for fair data sharing. And provide a principled way to choose reasonable values for privacy levels that are more relevant to real-world scenarios

In [2], William Hurst Carl Chalmers presented a design Data Processing and Neural approach future work will corporate additional data sets including home plugged readings showing how much activity. Concerns over privacy issues may limit the access to valuable information.

Paper [3] "Data Mining Techniques for Detecting Household characteristic Based on Smart Meter", proposed the that main goal of this research is to discover the structure of the home appliances usage patterns thereby providing more intelligence in smart metering system.

In [4], Yi-ChengChen ,Hsiu-Chieh Hung introduced a Dynamic Correlation Miner , is developed to incrementally maintain the correlation patterns among appliances in a smart home environment.

In [5], ShervinShirmohammadi further proposed the incremental mining of frequent power consumption patterns from smart meters big data. Our model exploits the benefits of pattern growth strategy and mine in quantum of 24hour period.

In [6], Md Zia UddinaDeok-Hwan KimaJeong proposed LBP-based features of depth silhouettes for HAR and obtained superior recognition results over the traditional silhouette based approaches. Automatically recognizes various indoor activities of a resident.

In [7], Proposed the incremental mining of frequent power consumption patterns from smart meters big data. Our model exploits the benefits of pattern growth strategy and mine in quantum of 24hour period. highlighted.

In [8], author claims that an unsupervised progressive incremental data mining mechanism applied to smart meters energy consumption data through frequent ern mining to overcome these challenges.

In [9], Gautam surveyed the Health monitoring approach consist of some processing steps, which are signal enhancement, watermarking, feature extraction, ECG analysis, and signal reconstruction

In [10], Authors proposed a data analytic approach that helps detect energy usage anomalies corresponding to the behavioral abnormality of the residents.

### III. SYSTEM ARCHITECTURE

It first cleans up and prepares the data, and then performs frequent pattern analysis to determine the relationship between devices, that is, determine which devices can work together. Then use cluster analysis to determine equipment allocation over time. The system can extract device usage patterns and use them as input to the Bayesian network to predict short-term and long-term behaviour. The system output is used by a specific healthcare application, depending on its intended use. For example, health care providers may be interested in learning activities related to cognitive impairment only when tracking the order of daily activities is critical. If abnormal behaviour is detected, alert the patient. The next section will explain these processes and briefly introduce the theoretical background. The brief representation of above explained system architecture is given figure.

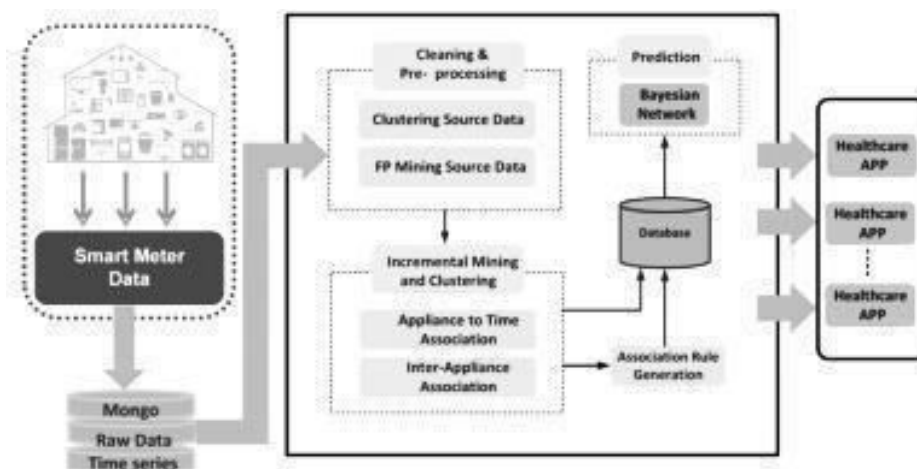


Fig.1. system architecture

#### Algorithm 1 Incremental Frequent Pattern Mining:

Transaction database (DB), Frequent pattern discovered database (FP DB) . Incremental discovery of frequent patterns, stored in frequent patterns discovered database (FP DB)

1. For all Transaction data slice db24 in quanta of 24 hours in database DB do data is processed in slices of 24 hour period.
2. Determine database size Database Sizedb24 for data slice/quantum db24
3. Mine Frequent patterns in FP DBdb24 using extended FP-growth approach
4. for all Frequent Pattern FP in FP DBdb24 do
5. Search a frequent pattern FP in FP DB

6. if Frequent Pattern found then
7. Update frequent pattern in FP DB
8. Else
9. Add a new Frequent Pattern to FP DB
10. end if
11. end for
12. For all Frequent Patterns in Database FP DB increment Database Size by Database Sizedb24
13. end for

**Algorithm 2: Clustering Analysis: incremental k-means :**

1. Select the number K to decide the number of clusters.
2. Select the number K to decide the number of clusters.
3. Assign each data point to their closest centroid, which will form the predefined clusters.
4. Repeat the third steps, which means reassign each datapoint to the new closest centroid of each cluster.
5. If any reassignment occurs, then go to step-4 else go to FINISH.
6. The model is ready.
7. end for

#### IV CONCLUSION

In this Paper .we presented a model for recognizing human activities patterns from low resolution smart meters data. Occupants' habits behaviour and follow a pattern that would be utilized in health applications to trace the wellbeing of individuals living alone or those with self-limiting conditions. Most of those activities are often learned from appliance-to-appliance and appliance-to-time associations. We presented incremental frequent mining and prediction model supported Bayesian network. In our current work, through experiments.we found that 24-hour period was optimal for data processing, but we built the model to work on any quantum of your time . From the experiment results we've demonstrated the applicability of the proposed mode l to properly detect multiple appliance usage and make short and future prediction a thigh accuracy.

#### REFERENCES

- [1] ABDULSALAM YASSINE , “Smart Meters Big Data : Data Sharing in Smart Grids.”,
- [2] William Hurst Carl Chalmers , “Smart Me- ters Profiling For Health Application”,
- [3] Krzysztof Gajowniczek , “Data Mining Techniques for Detecting Household characteristic Based on Smart Meter Data .”,
- [4] Yi-ChengChen ,Hsiu-Chieh Hung, “Incrementally Mining Usage Correlations among Appliances in Smart Homes.”,
- [5] Shervin Shi Mohammadi, “Incremental Mining of Frequent Power Consumption Patterns from Smart Meters Big Data”
- [6] Md Zia UddinaDeok-Hwan KimaJeong, “An Indoor Hu- man Activity Recognition System for Smart Home Using Local Binary Pattern”, [7] Springer- Verlag, Joseph Ferreira, “Clustering daily patterns of human activities in the city”,
- [8] SHAILENDRA SINGH, “Mining Energy Consummation Behaviour Patterns for Households in Smart Grid”,



- [9] Ghulam Muhammad, “Cloud-assisted Industrial Internet of Things enabled framework for health Monitoring.”,  
[10] S. Singh, A. Yassine, and S. ShrimohanMadi, “Smart- Energy Group Anomaly Based Behavioural Abnormality Detection.”,  
[11] J. Liao, L. Stankovic, and V. Stankovic, “Smart-energy group anomaly based behavioural abnormality detection”,  
[12] A. B. G. Hernando, and I. P. de la Cruz, “The Elderly’s independent living in smart homes: A characterization of activities and sensing infrastructure survey to facilitate services development.



**INNO**  **SPACE**  
SJIF Scientific Journal Impact Factor  
**Impact Factor: 7.542**



**ISSN** INTERNATIONAL  
STANDARD  
SERIAL  
NUMBER  
**INDIA**



# INTERNATIONAL JOURNAL OF INNOVATIVE RESEARCH

IN COMPUTER & COMMUNICATION ENGINEERING

 **9940 572 462**  **6381 907 438**  **ijircce@gmail.com**



[www.ijircce.com](http://www.ijircce.com)

Scan to save the contact details