# Human Activity Patterns Prediction (HAPP) System for Smart Healthcare Applications

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Abstract: Human services administration is a standout amongst the most testing angles that is significantly influenced by the immense flood of individuals to downtown areas. Thusly, urban communities around the globe are putting intensely in advanced change with an end goal to give more advantageous biological community to individuals. In such change, a large number of homes are being furnished with brilliant gadgets (for example shrewd meters, sensors and so on.) which produce gigantic volumes of fine-grained and indexical information that can be investigated to help savvy city administrations. In this paper, we propose a HAPP model that uses keen home enormous information as a methods for learning and finding human movement designs for medicinal services applications. We propose the utilization of successive example mining, group examination and expectation to quantify and dissect vitality use changes started by tenants' conduct. Since individuals' propensities are for the most part distinguished by regular schedules, finding these schedules enables us to perceive irregular exercises that may demonstrate individuals' challenges in taking consideration for themselves, for example, not planning nourishment or not utilizing shower/shower.

Keywords: Social media, Machine learning, Clustering, Smart meter, prediction HAPP model

#### 1. Introduction

Studies demonstrate that by year 2050, 66% of the total populace will live in urban territories [1]. The interest for medicinal services assets will be enormously influenced by this tremendous convergence of individuals to downtown areas. This exceptional statistic change places tremendous weight on urban areas to reconsider the customary methodologies of giving well being administrations to occupants. In reacting to the new needs and difficulties, urban communities are at present grasping gigantic advanced change with an end goal to help maintainable urban networks, furthermore, give more advantageous condition [2] [3]. In such change, a large number of homes are being furnished with keen gadgets (for example keen meters, sensors and so on.) which create monstrous volumes of fine-grained and indexical information that can be investigated to help human services administrations. Progression of enormous information mining advancements, which give methods for preparing colossal measure of information for significant experiences, can help us in seeing how individuals approach their life. For instance, observing the progressions of machine utilization inside a keen home can be utilized to in a roundabout way decide the individual's prosperity dependent on verifiable information. Since individuals' propensities are for the most part distinguished by regular schedules, finding these schedules enables us to perceive bizarre exercises that may show individuals' troubles in taking care for themselves, for example, not planning sustenance or not utilizing shower/shower [4] [5].

The hidden relationship between's apparatus utilization inside the keen home and routine exercises can be utilized by medicinal services applications to identify potential medical issues. This isn't just going to reduce the load on medicinal services frameworks, yet in addition giving 24 hour checking administration that consequently distinguish ordinary and unusual practices for freely living patients or those with self-restricting conditions (for example old and patients with psychological impedances). This paper proposes the utilization of vitality information from brilliant meters introduced at homes to uncover vital exercises of occupants. Our investigation expect that there are instruments set up to shield individuals' security from being shared or estimated for unlawful uses as talked about in [6] and [7]. The proposed model watches and dissects readings from keen meters to perceive exercises and changes in conduct. Disaggregated control utilization readings are straightforwardly identified with the exercises performed at home."Stove" is ON, the activity of this apparatus is the best bet related with action "Getting ready Food".

The time (for example morning or night) of this activity may likewise demonstrate the sort of the supper, for example, breakfast or supper. Moreover, individuals frequently perform more than one action at the equivalent time, for example, "Getting ready Food" and "Tuning in to Music" or on the other hand "Sitting in front of the TV", which implies different machines are worked together. In this unique situation, we break down shoppers' worldly vitality utilization designs at the machine level to distinguish numerous machine utilizations and foresee theirtasks over short and long haul time periods. This is especially conceivable without extra equipment since the keen meter information have time-arrangement thought ordinarily comprising of use and utilization estimations designs of segment apparatuses over a period interim [8]. Such attempt, in any case, is trying since it is difficult to distinguish use conditions among different apparatuses at the point when their activity cover or happen in the meantime. Moreover, inferring exact forecast of human movement designs is impacted by the probabilistic connections of machine use occasions that have dynamic time interims.

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To handle the previously mentioned issues, this paper proposes visit mining and expectation model to gauge and investigate vitality utilization changes started by inhabitants' conduct. The information from brilliant meters are recursively mined in the quantum/information cut of 24 hours, and the outcomes are kept up crosswise over progressive mining works out. We additionally use the Bayesian system, a probabilistic graphical model, to anticipate the utilization of different machines and family unit vitality utilization. The proposed model is able to do present moment forecasts running from one hour from now as long as 24 hours and longterm expectation for a considerable length of time, weeks, months, or seasons. For the assessment of the proposed instrument, this examination employments the UK Domestic Appliance Level Electricity dataset (UKDale) [9] - time arrangement information of intensity utilization gathered from 2012 to 2015 with time goals of six seconds for five houses with 109 machines from Southern England. It must be noticed that practically speaking burden disaggregation is conveyed by Non-Intrusive Appliance Load Monitoring (NALM) method. NALM is a system used to disaggregate a home's control use into individual apparatuses and name them for further mining and investigation.

Electrical apparatus utilization examples and after that utilizes machine learning-based calculation to disengage the real exercises inside the home. The issue is that the investigation needs to perform two stages on the information to totally segregate the fundamental exercises. Abusing apparatus use designs and recognize them for sudden social change is displayed in [12]. The point of the investigation is to give nonstop checking framework to help individuals' experiencing Alzheimer or Parkinson illness at least interruption level. The investigation utilizes order strategies to recognize unusual conduct of individual vitality use designs in the home. Different investigations for example, [13] [14] [15], and [16] despite the fact that don't use savvy meters information, they use Internet of Things (IoT) foundations in savvy urban communities for creating applications that screen and give wellbeing administrations to patients.

Data analytics is a technology-enabled strategy for gaining richer, deeper, and more accurate insights into the details of the customers, partners and the businesses and hence ultimately gaining competitive advantage. By processing a steady stream of real-time data, organizations can make time-sensitive decisions faster than ever before, monitor emerging trends, course-correct rapidly and jump on new business opportunities. It is observed that nowadays, various organizations are consuming large amount of essential information which can be useful in other related fields such as monitoring of an object activity, sensor deployment, tracking of data etc. This is not actively explored because of the challenges in the present infrastructure of data storage management. This kind of challenging situation arises even when an organization wants to explore its own data from its personal websites for analyzing the customer's feedbacks, customized services towards a product, etc. The concept of big data helps us to understand the abstraction and in-depth understanding behind various hidden values.

As a result, the decision makers would convey their conclusions grounded on the analysis of extracted data or those data which carry some value or weight age. Data analytics is also applied on unrelated attributes of datasets. In most of the scientific fields, grids had been developed with storage, processing, and availability of data. A grid is a collection of distributed computing resources available over a local or wide-area network that appears to an end user or application as one large virtual computing system. It can be seen that some specific technologies and implementations are required for the cloud to provide infrastructure, platform and software sources.

## 2. Literature Survey

The social network is a branch of data mining which involves finding some structure or pattern amongst the set of individuals, groups and organizations. A social network involves representation of these societies in the form of a graph with the individuals as the vertices and the relationship among the individuals being represented by the edges. Community structure in any given social network gives us an indication of some important pattern which may be hidden on normal analysis, and thus can help us to understand a lot of processes and phenomenon of social networks and communities better. This also helps when someone makes an application using the social network and its communities [11].

The social network is naturally characterized by multiple community memberships. For example, a person usually has connections to several social groups like family, friends, and colleagues; a researcher may be active in several areas. Further, in online social networks, the number of communities an individual can belong to is essentially unlimited because a person can simultaneously associate with as many groups as he wishes. This also happens in other complex networks such as biological networks, where a node might have multiple functions **[16]**.

The society, is possible to find groups, such as families, coworkers' circle, friendship circles, villages, and town that naturally form. Similar to this, in an online social network, we can find virtual groups, which live on the web. For example, in World Wide Web it will help to optimize the Internet infrastructure in a purchase network it can boost the sell by recommending appropriate products and in computer network it will help to optimize the routing table creation. Again, identifying special actors in the network is also a motivating force behind community detection **[24]**.

The Modularity-based community detection methods aim to find a hard partition of a given network, where a vertex can belong to only one community. However, a person usually has different involvements in several communities, e.g., splitting time between a circle of friends, a club, and her family. Thus, it is common to see that communities of a realworld social network tend to be overlapping. Since social network players can have partial belongingness to multiple communities in real world networks, fuzzy partitions are appropriate **[13]**.

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The Social media networks provide people with the ability to ensure complete connectivity, bringing people with common interests together, creating a platform to share one's life experiences with the rest of the world. A few examples for types of social media are websites and applications concerned with discussion forums, blogging, social networking, social bookmarking, and audio and video conferencing where it is used in both web and mobile applications, thus enhancing knowledge sharing among people [17].

The ordinal optimization using rough models and fast simulation is introduced to obtain suboptimal solutions in a much shorter timeframe. While the scheduling solution for each period may not be the best, ordinal optimization can be processed fast in an iterative and evolutionary way to capture the details of big-data workload dynamism. Experimental results show that this evolutionary approach compared with existing methods, such as Monte Carlo and Blind Pick, can achieve higher overall average scheduling performance, such as throughput, in real-world applications with dynamic workloads. The authors investigate the localrecoding problem for big data anonymization against proximity privacy breaches and attempt to identify a scalable solution to this problem. Specifically, they present a proximity privacy model with allowing semantic proximity of sensitive values and multiple sensitive attributes, and model the problem of local recoding as a proximity-aware clustering problem. A scalable two-phase clustering approach consisting of a t-ancestors clustering (similar to kmeans) algorithm and a proximity-aware agglomerative clustering algorithm is proposed to address the above problem. Designed the algorithms with MapReduce to gain high scalability by performing data-parallel computation in cloud.

# 3. Methodology

## 3.1Data Mining Process

Data mining involves six common classes of tasks, namely

- *Anomaly detection* (outlier/change/deviation detection) The identification of unusual data records, that might be interesting or data errors that require further investigation.
- Association rule learning (dependency modelling) Searches for relationships between variables. For example, a supermarket might gather data on customer purchasing habits. Using association rule learning, the supermarket can determine which products are frequently bought together and use this information for marketing purposes. This is sometimes referred to as market basket analysis.

*Clustering* – is the task of discovering groups and structures in the data that are in some way or another "similar", without using known structures in the data.

*Classification* – is the task of generalizing known structure to apply to new data. For example, an e-mail program might attempt to classify an e-mail as "legitimate" or as "spam".

*Regression* – attempts to find a function which models the data with the least error that is, for estimating the relationships among data or datasets.

*Summarization* – providing a more compact representation of the data set, including visualization and report generation.

#### **3.2. Proposed HAPP Model**

This HAPP incorporates visit mining and expectation model to quantify and break down vitality utilization changes started by tenants' conduct. The information from savvy meters are recursively mined in the quantum/information cut of 24 hours, and the outcomes are kept up crosswise over progressive mining works out. Expectation incorporates Bayesian system, a probabilistic graphical model, to anticipate the utilization of different machines and family unit vitality utilization. The HAPP demonstrate is prepared to do transient expectations going from one hour from now as long as 24 hours and long haul forecast for a considerable length of time, weeks, months, or seasons. The main objectives of this model includes.

**Phase 1:** To develop a model for mining human activity patterns from smart home Big data for health care application.

**Phase 2:** To use the frequency pattern mining to predict the anomalous behavior of the human based on their routine activities.

**Phase 3:** To apply k means clustering effectively to group the usage of appliances.

Phase 4: To apply Bayesian networks for predictions.

#### Smart Meter

A smart meter is an electronic gadget that records utilization of electric vitality and conveys the data to the power provider for observing and charging. Shrewd meters ordinarily record vitality hourly or all the more as often as possible, and report in any event every day. Savvy meters empower two-path correspondence between the meter and the focal framework. Such a progressed metering foundation (AMI) contrasts from programmed meter perusing (AMR) in that it empowers two-path correspondence between the meter and the provider. Correspondences from the meter to the system might be remote, or by means of settled wired associations, for example, control line bearer (PLC). Remote correspondence alternatives in like manner use incorporate cell interchanges (which can be costly), Wi-Fi (promptly accessible), remote impromptu systems over Wi-Fi, remote work systems, low power long range remote (LORA), ZigBee (low power, low information rate remote), and Wi-SUN (Smart Utility Networks).

The proposed HAPP framework utilizes vitality information from shrewd meters introduced at homes to disclose vital exercises of occupants. HAPP expect that there are systems set up to shield individuals' security from being shared or estimated for unlawful utilizations .The proposed model watches and breaks down readings from keen meters to perceive exercises and changes in conduct. Disaggregated control utilization readings are specifically identified with the exercises performed at home. For example, if the "Broiler" is ON, the task of this machine is doubtlessly connected with action "Planning Food". The time (for example morning or night) of this task may likewise demonstrate the sort of the supper, for example, breakfast or supper.



Figure 1: Work Flow

#### Algorithm

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 DBdb*24 using ex- tended 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

## 4. Result and Discussions

The following efficiency report shows the network structures are much more mind boggling, in actuality: the system is colossal, the quantity of vertices in different systems are specific and there is wonderful differentiation between middle points' degree. The systems benchmark controls the lucidity of the system structure. With the expansion of network in the structure of system becomes vague, and the detection of diseases becomes more difficult.



Graph 1: Clustering Analysis



# 5. Conclusion

HAPP display for perceiving human exercises designs from low goals savvy meters information has been exhibited effectively. Inhabitants' propensities and conduct pursue an example that could be utilized in wellbeing applications to follow the prosperity of people living alone or those with self-constraining conditions. The vast majority of these exercises can be gained from apparatus to-machine and apparatus to-time affiliations. In stage 1 execution of HAPP demonstrate we have given an image of gradual successive mining and forecast model dependent on Bayesian system. In current work, through examinations, it is discovered that 24-hour time span was ideal for information mining. From the test results it has been exhibited the appropriateness of the proposed model to effectively distinguish different machine utilization and make short and long haul expectation at high precision.

## 6. Future Enhancement

The future work of phase 2 implementation of HAPP model, includes a plan to refine the model and introduce distributed learning of big data mining from multiple houses. This will help health applications to promptly take actions such as sending alert to patients or care providers. This means HAPP model can be efficiently trained and can increase the accuracy of detecting human activities.

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