

# Prophecy of Electricity Price and Units for Home Appliances Using Random Forest

Ramyadevi K, Muthumari M

Department of Computer Science, S.A.Engineering College, Chennai, India

**ABSTRACT:** The well-aimed prediction of electricity consumption is a vital foundation for smart energy management. Decision based electricity is contributed for different scenarios. Electricity consumption of home appliances is gathered in the form of datasets and given in hadoop Environment. This datasets is given to the random forest classifier technique for further process. The forecasted electricity consumption is displayed through the GUI. Based on the predicted value the electricity is transferred for following days. This approach outfits all benchmarks in terms of accuracy, stability and generalization.

**KEYWORDS:** electricity prediction, random forest ,hadoop, dataset, Label encoder.

## I. INTRODUCTION

Electricity is well connected to the human beings now days so the demand of electricity is also increasing rapidly. The consumers of electricity should know how much electricity is consumed by them every month to effectively use the available energy. Prophecy of electricity plays a vital role.

A huge amount of the data that may contain number, text, picture etc. is called has a Bigdata. A hadoop is a open source tool where you can handle the bigdata. A map reduce model in the hadoop is used for a very storage and retrieval. The real time data about electricity is given to the hadoop environment for storing it in distributed computers and partition the data's according to the appliances. Machine learning is the technique that proposed to increase the computers thinking ability and to reduce the human effort. Here the data from the big data part is given to the label encoder that covers the data into machine understandable part then a random forest classifier algorithm is applied . That predict the amount of electricity consumed and amount per month as a result.

## II. RANDOM FOREST

Random forest or Random decision forest is an algorithm that integrates few machine learning techniques into a single prognostic model. It steers by assembling a heaps of decision trees at drilling time and outputs the class in the form classification and regression in discrete trees. Random Forest is a scheme of supervised machine learning algorithm based n ensemble learning. Ensemble learning employs manifold learning algorithm to pick up a better predicted execution. Random forest is the most powerful which integrates the outcome of various learning algorithm resulting as trees of a forest. Hence Known as RANDOM FOREST.

## III. LITERATURE SURVEY

### Electricity prices forecasting using artificial neural network

Alma Y. Alanis, et al.

IEEE Latin America Transactions (volume:16, Issue:1, Jan 2018)

It makes use of the neural network based on Extended Kalman Filter. This algorithm is used in both prediction i.e 1step ahead and n step ahead. It also includes the Lyapunov method. The applicability of proposal system is shown by both predictions using data from European power system.

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**Self adaptive radial basis function neural network for short term electricity price forecasting**

K.Meng, et al.

IET Generation, Transmission and Distribution (Volume:3; Issue:4, Apr 2009)

It uses RBF neural network model and initiated by fuzzy c-means and differential evolution. It also configures the structure of network and acquires parameters. A moving window wavelet de-noising technique is used to get better network performance. This method is effective by using RBF neural network in predicting the Mackey Glass Chaos time series in Queensland.

**Electricity consumption behaviour analysis based on adaptive weighted feature k-means ap clustering**

Chunyan Li, et al.

IET Generation, Transmission and Distribution (Volume:13; Issue:12, Jun 2019)

An Adaptive weighted feature k-means affinity propagation clustering algorithm is presented to analyse electricity utilisation. Feature set is constructed and applied to divide and conquer method to analyse consume electricity usage. AP clustering results are produced by combining time and frequency domain. The performance analyses datasets and verifies the approach.

**A new recursive dynamic factor analysis for point and interval forecast of electricity price**

h.c.wu, et al.

IEEE transactions on power system (volume:28; Issue:3, August 2013)

The electricity price is predicted using the Recursive Dynamic Factor Algorithm (RDFA) algorithm where PC's are recursively tracked using subspace tracking algorithm, and generate the PC scores and it given to Kalman filter.

**Forecasting the mean and variance of Electricity price in Deregulated markets**

Claudio m. rubal, et al

IEEE transactions on power system (volume:23; Issue:1, February 2010)

A fundamental bid based stochastic model is used to predict the electricity price in a given period. It captures the economical and physical aspects of pricing. The sensitivity analysis is performed on the number of firms, peak demand and electricity to predict the value.

**On the use of functional additive models for electricity demand and price prediction.**

Paula rana, et al.

IEEE on February 13, 2018.

The application of functional additive models is used for electricity demand and price prediction. Data obtained from pointwise prediction is displayed on web or mobile using bootstrap procedure.

**Electricity price prediction for Geographically distributed data centres in multiregion electricity market**

MohMohThan, et al.

2018 3rd International Conference on Computer and Communication Systems

The electricity price prediction is designed for GDCs in multi region electricity markets. Experiment is conducted on real life electricity price data set with machine learning algorithm. The most accurate one is selected by comparatively assessing the prediction accuracy of the models,

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**The electricity bill charge risk analysis in power supply company based on a novel prediction method**

Gang Zhou,et al;

2016 IEEE Innovative Smart Grid Technologies - Asia (ISGT-Asia)

The SMOTE(synthetic minority over sampling technique) algorithm is used to under sampling the majority class and sampling the minority class and then it is combined with some state of the art classification method to predict the electricity charge risk based on an imbalance data sets obtained from a power supply enterprise. The results of empirical analysis show that combination of SMOTE algorithm with random forest method will achieve better classifier performance under several criterion.

**Load scheduling based on an advanced real-time price forecasting model**

Xing Luo,et al.

2015 IEEE International Conference on Computer and Information Technology

An advanced RTP forecasting model on the basis of the least-square(LS) fitting function and the grey prediction technique(GPT) is proposed by considering the factors such as predicted real-time price ,type of appliances ,user's preferences behaviour etc., To schedule the operating time of home appliances intelligently and estimate the resulting electricity bill load scheduling approach is introduced.

**Support vector machine regression for forecasting electricity demand for large commercial buildings by using kernel parameters and storage effect.**

N. G. I. S. Samarawickrama,et al.

2016 Moratuwa Engineering Research Conference (MERCon)

This paper provides accurate and efficient energy forecasting tool based on support vector machine regression(SVMR).

**Forecasting brazil's electricity consumption with pegel's exponential smoothing**

P.M.Macaria,etal.

IEEE Latin America Transactions ( Volume: 14 , Issue: 3 , March 2016 )

It aims and frame work the Brazilian energy utilisation aggregated and disaggregated by the class of consumption with pegel's exponential smoothening model.Theoptimisation procedure is carried out to alter forecasted values.The output shows that it is achievable to predict assusaging the electricity consumption upto to 2050 for all energy sector.

**Hybrid machine learning system to forecast electricty consumption of small grid based air-conditioners**

Jui-sheng chou,et al.

IEEE Systems Journal ( Volume: 13 , Issue: 3 , Sept. 2019 )

It develops a hybrid prediction system to forecast 1-day ahead electricity consumption in office space.It estimates the electricity consumption and smart grid based monitoring devices are installed.Data are assembled to train and test the system.Sensitivity analysis is performed and the result provides good pact between the predicted value and the actual consumption.

**Clustering of electricity consumption behaviour dynamics towards big data application**

Yi Wang,et al.

IEEE Transactions on Smart Grid ( Volume: 7 , Issue: 5 , Sept. 2016 )

It focuses on novel transistion and relation between consumption behaviour.Symbolic aggregate approximation is performed to reduce scale of data set,time based Markov model is applied.A clustering technique by fast search and find of Density Peaks is carried out.Finally verifies the effectiveness of models and approaches.

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### **Time-Coordinated Multi-Energy Management of Smart Building under Uncertainties**

Sumedha Sharma, et al.

IEEE Transactions on industrial informatics( August 2018 )

A multi timescale coordinated building energy management system (BEMS) for multi energy buildings integrated with renewable energy sources (RES). It aims to dispatch active building components in two different timescales to counteract uncertain variations in RES generation and load. In the longer timescale (hourly), fuel-cell-based micro combined heat and power and energy storage system (ESS) are dispatched before uncertainty is realized.

### **Virtualizing Power Distribution in Data Centers**

Di Wang, ChuangangRen,et al.

IEEE Transactions on Smart Grid ( Volume: 7 , Issue: 5 , Sept. 2016 )

Batteries and other energy storage devices have been proposed to provide a complementary alternative to these knobs, which when decentralized (or hierarchically placed), can temporarily take the load to suppress power peaks propagating up the hierarchy.vPower allows applications to specify their power needs, performs admission control and placement, dynamically monitors power usage,and enforces allocations for fairness and system efficiency.

### **Iterative bidding in electricity markets:rationality and robustness**

Aslishcherukuri,etal

IEEE transactions on Network Science and Engineering(2019)

The goal is to solve the optimal power flow problem.Each generator submits to the ISO a bid,consisting of the price per unit of electricity at which it provide power.It established the algorithm robustness to collusion.There is no incentive for any group of generator to share information with the intent of tricking the system to obtain a higher payoff.

### **Benefits and Limitations of Tapping into Stored Energy for Datacenters**

Sriram Govindan,etal

38<sup>th</sup> Annual Internation symposium on Computer Architecture

Datacenter power consumption has a significant impact on both its recurring electricity bill (Op-ex)and one-time construction costs.The detailed analysis of battery operation to figure out feasible operating regions given such battery lifetime and datacenter availability concerns. Using insights learned from this analysis, we develop peak reduction algorithms that combine the UPS battery knob with existing throttling based techniques for minimizing data center power costs.

### **Integrated optimization of Smart Home Appliances with Cost-effective Energy Management System**

TesfahunMolla,et al.

CSEE Journal of Power and Energy Systems ( Volume: 5 , Issue: 2 , June 2019 )

A general architecture of home energy management system (HEMS) is developed in smart grid scenario with novel restricted and multi-restricted scheduling method for the residential customers.To optimize the formulated problem, a powerful meta-heuristic algorithm called grey wolf optimizer (GWO) is utilized, which is compared with particle swarm optimization (PSO) algorithm to show its effectiveness.various techniques and algorithms have been adopted to minimize the cost of electricity billing based on TOUP tariffs and incentives. Consumers can generate renewable energy to supply their appliances and if there is any excess production.

### **Intelligent Modeling and Optimization for Smart Energy Hub**

TianhaoLiu,et al.

IEEE Transactions on Industrial Electronics ( Volume: 66 , Issue: 12 , Dec. 2019 )

It focuses on the multi-energy planning, and gradually form a consensus on the development of multiple energy system (MES). In the MES analysis, there are three main aspects: planning scheme, optimal strategy for the energy management system (EMS), and energy trading mechanisms.

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**Prediction of day-ahead electricity price based on information fusion**

Huixiantian, et al.

2010 International Conference on Computer and Information Application

A novel information fusion method is used. A new neural network method ELM is selected for improve on performance as the core algorithm of information fusion a new neural network method ELM is selected. A new modeling approach is proposed to establish the prediction model using the information fusion ideas. The day-ahead electricity price prediction model is tested by the real-time data. The new prediction model established by ELM information fusion method has surpass performance.

**IV. CONCLUSION**

The study on Power Distribution to Datacenters data is help to give awareness about how much power consumption in different datacenters. To run a analysis on the Power Distribution to Datacenters data in hadoop ecosystem. Hadoop ecosystem uses hive, pig, map reduce to analysis faster.

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