

Aalto University
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Using Big Data Analytics for Measuring Energy Consumption Patterns:

Applying big data for energy efficiency

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Thank you, and keep up the good work!

Espoo, June, 2014

Hussnain Ahmed

Abbreviations and Acronyms

2k/4k/8k mode	COFDM operation modes
3GPP	3rd Generation Partnership Project
ESP	Encapsulating Security Payload; An IPsec security protocol
FLUTE	The File Delivery over Unidirectional Transport protocol
e.g.	for example (do not list here this kind of common acronyms or abbreviations, but only those that are essential for understanding the content of your thesis.
note	Note also, that this list is not compulsory, and should be omitted if you have only few abbreviations

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Chapter 1

Introduction

In modern era we have seen phenomenal increase in human dependency on information and communication technology (ICT) enabled products and services. It has transformed the way of life on the planet. From fulfilling very basic physiological needs like health and food to the human needs of communicating with others and being part of wider social groups, we need and depend on ICT. There are many research areas and opportunities that are emerging as bi-products of this continuous transformation. One of them is the availability of digital traces of human activities. With every instance of use of these services we produce a digital trace that can be recorded and analysed. Big Data is a term that is being widely used to refer to these digital traces of human activity. Ubiquity of computing resources, fast and highly mobile connectivity and advent of social media usage has caused a great surge in volumes of data. Realizing the true potentials of data, businesses are not only utilizing it as source of decision making but new revenue lines and opportunities are emerging that are reshaping the business models of many companies around the globe.

To support this transfiguration, we have seen a rapid development in distributed parallel computing, data communication software and machine learning. Industry giants like Google and Yahoo has opened technologies and tools like MapReduce and Hadoop to facilitate these advancement and open source software communities like Apache Software foundation has further developed the tools to provide a complete ecosystem for handling big data and generate insights. The new specialized big data companies like Cloudera and Hortonworks has emerged that has acted as catalyst for this data revolution. In this research we try to formulate a model for end to end big data analytics platform based on these technologies that can ingest data from heterogeneous sources, process it in an efficient way, mine the data to generate the insights based on business logic and then present the information using interactive

visualizations. This thesis includes the development as well as implementation of the mentioned big data platform to perform analysis on real life use case and generate useful insights. The model that we present in this thesis is based on open source software components available free of charge. There are other closed source software alternatives that can fit into the presented model but they are not discussed in this scope of this thesis.

This thesis is also part of European Union CIVIS- Cities as drivers of social change project under 7th framework. CIVIS project focuses on adaption of ICT tools and techniques for low carbon smart energy grid, distributed energy and information flow. The use of pervasive ubiquitous computing is driving the smart energy solutions. Combined with internet of things (IoT) for home/building automation, smart commuting, and remote monitoring is becoming the basis for energy conservation and energy efficiency. All the smart energy devices as part of this ecosystems generates high volumes of data, that needs to be instantaneously transferred, stored, analysed and visualized for knowledge discovery and improvements of services for the goal of achieving high energy efficiency. The platform that was developed as part of this thesis has the capability to automate the whole process.

Energy usage pattern detection, classification of buildings on basis of energy efficiency and a prediction model for energy consumption per household will be the use cases for validating the developed big data analytics platform. These use cases also provide the basis for designing, planning and implementing schemes for improving energy related services for sake of achieving higher efficiency in both production and usage that contributes to cause of green environment in terms of less CO₂ emissions. The insight generated from these use cases can also help in educating the consumer about benefits of energy conservation and spread the awareness about behavioural changes that can benefit society as well as individuals themselves.

This master thesis is also supported by VTT, Technical Research Centre of Finland as part of their Green Campus initiative that focuses on use of ICT based solutions for innovative energy management and control systems capable to optimize the consumption without compromising the indoor environment. VTT is also a supporter and partner of CIVIS project. VTT has installed specialized smart devices in selected test sites that are the buildings owned by Aalto University. VTT has contributed to this thesis by providing the data generated by these smart devices. VTT has also helped in scoping for the use cases for energy efficiency by the experience and the knowledge they have from the related projects and research.

In a nutshell, this thesis focuses on providing a solution for collecting, storing, analysing and visualizing data generated by smart energy device for generating insights about energy consumption patterns and discovering the

performance of different building units in terms of energy efficiency. This thesis also provides the models for knowledge discovery that can be used to improve energy efficiency at both producers and consumers ends. The big data analytics platform developed as part of this thesis is not limited to be used only for energy efficiency. It has the capability of handling other big data uses cases as well but we shall discuss its use for energy pattern detection and usage efficiency only in scope of this thesis report.

1.1 Problem statement

Energy conservation is required to reduce CO₂ emissions from energy production and usage. To achieve this goal we need to understand and improve the energy efficiency on both producer and consumer end. ICT enabled smart energy grids and devices are being rolled out globally to measure energy consumption and improve on energy efficiency. These smart devices produce high volumes of data that may or may not be predicted and planned at time of setting up the infrastructure. The data generated by different devices comes in different formats. For knowledge discovery from this data it is required to collect, store analyse the data and then visualize the generated insights so the information can be understood efficiently. The challenge gets even tougher when data needs to be collected and analysed in real time. Then with the time, volume of data and scope of analysis is expected to increase. So to cater for all this a highly scalable and flexible data analysis platform is required that can automate the whole process. This platform needs to be very cost effective for global adaptation.

In scope of this research we provide a model for big data analytics platform that can provide the solution for these requirements. We also implement the proposed model and test it with real life energy smart devices data and use cases. The proposed solution is based on open source components that can be deployed on general purpose commercially available, hence it is very cost effective. The proposed platform can be scaled according to data volumes and additional functional components can be integrated as per the scope of analysis.

1.2 Helpful hints

Read the information from the university master's thesis pages before starting the thesis. You should also go through the thesis grading instructions together with your instructor and/or supervisor in the beginning of your work.

1.3 Structure of the Thesis

You should use transition in your text, meaning that you should help the reader follow the thesis outline. Here, you tell what will be in each chapter of your thesis.

Chapter 2

Background

The problem must have some background, otherwise it is not interesting. You can explain the background here. Probably you should change the title to something that describes more the content of this chapter. Background consists of information that help other masters of the same degree program to understand the rest of the thesis.

Transitions mentioned in Section 1.3 are used also in the chapters and sections. For example, next in this chapter we tell how to use English language, how to find and refer to sources, and enlight different ways to include graphics in the thesis.

2.1 Smart grids

Energy industry across the globe is facing numerous challenges. There is a huge pressure from regulatory authorities and environmental organizations to reduce carbon foot print, expand their renewable energy portfolios, and take energy conservation measures. The demand response (DR)¹ and its impacts on consumer behaviour requires rapid adaptations in energy service providers business models. According to United States Federal Energy Regulatory Commission (FERC) , “Demand response can provide competitive pressure to reduce wholesale power prices; increases awareness of energy usage; provides for more efficient operation of markets; mitigates market power; enhances reliability; and in combination with certain new technologies, can support the use of renewable energy resources, distributed generation, and

¹Demand Response(DR); Changes in electric usage by end-use customers from their normal consumption patterns in response to changes in the price of electricity over time, or to incentive payments designed to induce lower electricity use at times of high wholesale market prices or when system reliability is jeopardized.

advanced metering. Thus, enabling demand-side resources, as well as supply-side resources, improves the economic operation of electric power markets by aligning prices more closely with the value customers place on electric power” [7]. Traditionally, power system participants have been strictly producers or consumers of electricity. The demand response and reliability issue with conventional electric power distribution models on consumer side are causing a major trend in motivating consumers to produce electricity at domestic level mostly using the renewable energy production methods. “Prosumer” is an emerging term used for an economically motivated entity that: [13]

- Consumes, produces, and stores power,
- Operates or owns a power grid small or large, and hence transports electricity, and
- Optimizes the economic decisions regarding its

The current energy grids support unidirectional distribution models and are centralized in nature. They are very limited to handle the prosumer needs. Line losses and hierarchical topology makes them less reliable. They usually become bottle neck when rapid adaptations are required for demand response. Farhangi, 2010 define smart grids as “The next-generation electricity grid, expected to address the major shortcomings of the existing grid. In essence, the smart grid needs to provide the utility companies with full visibility and pervasive control over their assets and services. The smart grid is required to be self-healing and resilient to system anomalies. And last but not least, the smart grid needs to empower its stakeholders to define and realize new ways of engaging with each other and performing energy transactions across the system” [10].

In our research, we used data collected from smart metering devices as part of a pilot smart grid project. The data was used to generate analysis that recommends improvement for both demand and supply side to achieve energy efficiency as well as provide understanding to enable correct decision to adapt for demand response.

2.2 CIVIS project

CIVIS is the abbreviated name for “Cities as drivers of social change” project under European Union 7th framework. It is a part of the programme for optimising energy systems in smart cities. CIVIS project is a collaborative

effort of 10 European universities)². It aims to embed the social aspect into the advancements of energy technology. To unleash the full potential of this vision, smart grids need to be coupled with broader social and cultural considerations and understood as complex socio-techno-economic systems with multiple decision making layers that are in effect at the physical, cyber, social, and policy [9].

ICT acts as one of the main enabler of smart grids, distributed and bidirectional information flow models. On the other hand ICT also provides a lot of new mediums for social aggregation e.g. internet based social media. CIVIS projects tends to connect these two different dimensions with innovative ICT solutions. An integrated approach to energy efficiency is the basic manifesto of CIVIS project. [9]

Understanding energy usage patterns and benchmarking energy efficiency performance of small units within cities are some preliminary items in list of CIVIS objectives. Within scope of our research we analyze energy data to understand the consumption patterns and try to evaluate various factors that can effect directly or indirectly on the usage patterns. We also try to classify the building on basis of energy efficiency and try to test the sensitivity of energy efficiency with respect to factors that can cause shift in usage patterns. For the CIVIS project aim of social aspect integration, we also present an ICT application framework that can be used to collect and analyse social media data. However the analysis of that data is not within the scope of this research.

2.3 Green campus initiative

Green campus initiative is a project by VTT “Technical Research Centre of Finland” . It is part of EcoCampus 2030 program. EcoCampus is an attempt to contribute to increased energy efficiency in districts and buildings by innovative management and control systems capable to optimize the local consumption without compromising the indoor environment, occupant comfort and building performance, and by introducing new ICT enabled business models [20]. The vision of the program is to realize a net zero energy model for a world class research, development and educational facility. Program focuses on co-designing this model with user by educating them and then

²1. Associazione Trento RISE, Italy 2. Aalto university, Finland 3. Imperial College London, UK 4. ENEL Foundation, Italy 5. Instituto Superior TÁ©cnico, Portugal 6.Karlsruhe Institute of Technology, Germany 7.Kungliga Tekniska Hogskolan, Sweden 8.SANTER REPLY SpA Italy 9.Nederlandse Organisatie voor toegepast Natuurwetenschappelijkonderzoek, Netherlands 10. Delft University of Technology,Netherlands

collecting feedbacks for improvement. The main aim is to gain energy efficiency by building infrastructure in the building units that can make them self sustain for future requirements. The aim is build to build a performance based ecosystem that can help both consumers and producers to adapt with demand response.

Green campus initiative is a pilot project for EcoCampus program in which VTT has installed smart devices inside Aalto University, Finland campus building in cities of Espoo and Helsinki. These specialized devices contained smart metering for energy consumption and indoor environment monitoring sensors. The data used for analysis in our research was collected from 100 buildings as test sites. The data includes hourly consumption of electricity and electricity used for heating. For one of the test sites VTT provided us the data with the details up to use of respective electric devices used in that site. This was achieved using smart NIALM ³[16] meters that can distinguish between different electric devices used on basis of their signal thumb print.

Apart from providing the data, VTT green campus researchers have also helped us in formulating the use cases for this thesis research.

2.4 Big data analytics

Big data analytics is application of advance data analytics techniques on large volumes of data. Advance analytics is a generalized term used for data analysis techniques like statistical analysis, data mining, machine learning, natural language processing, text mining and data visualization etc [31]. Although volume of the data is a widely used factor for qualification of a data set as big data but when it come to big data analytics there few other important attributes i.e. variety, velocity, valuation and veracity. The concept of 3Vs (volume, variety and velocity) of data was first given by an analyst, Doug Laney from Gartner in a 2001 MetaGroup research publication, “3D data management: Controlling data volume, variety and velocity” [23]. Gartner used this concept to formulate a data magnitude index that can support decision making for selection of the solutions for tackling big data challenge on use case base. This concept is shown in figure 2.1 below

Number 0 to 3 represents the scale of data that you perceive on each dimension. Adding them together for a big data case can provide the data magnitude index. This method provides some basis for quantifying the data

³ NIALM stands for non-intrusive appliance load monitoring, is a process for analysing changes in the voltage and current going into a house and deducing what appliances are used in the house as well as their individual energy consumption

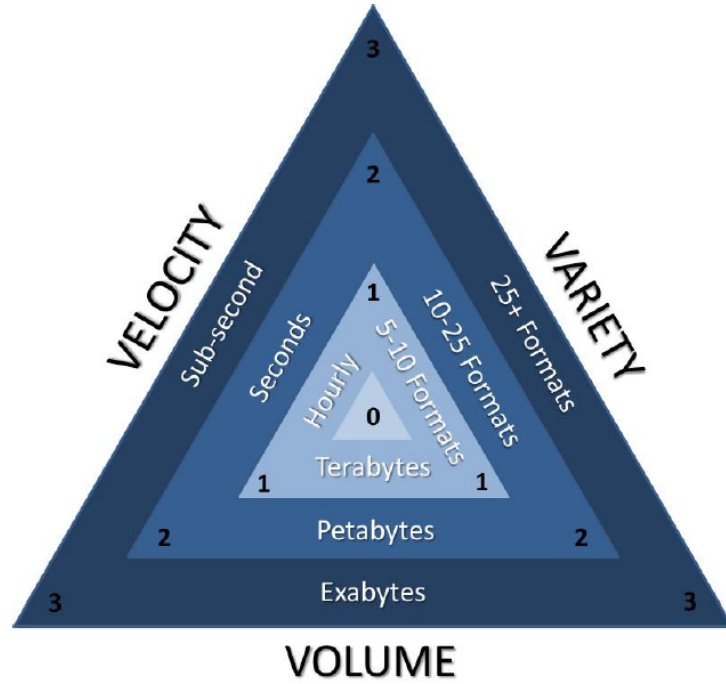


Figure 2.1: Gartner 3Vs of data and data magnitude index [23].

for big data qualification, However it is not providing a definitive model as it allows presumptions to scale the data. Valuation and veracity are two other factors that are being used widely along with Gartner's 3V. Valuation supports the decision making by considering the value of outcomes against the efforts required to collect, manage, process and analyse huge amounts of data. While veracity refers to ambiguity in the data that can cause complexity. There is no standard definition of big data but most of the attempts to define big data can be associated with these five factors that we have discussed.

As a matter of fact, we are not attempting to provide a definition of big data as part of this thesis or stating any criteria for qualification of a data set as big data. Instead we shall be proposing an advance analytics model that should be capable enough to handle big data as well other smaller data sets on need basis. The modular architecture of the model platform can be tweaked to handle volume, variety, velocity, and veracity on need basis while trying to maximize the valuation for the use case. In following subsections we shall discuss the some of the relevant technological advancements that enables to handle the mentioned challenges of big data analytics. These concepts, tools and techniques are also used in developing the data analytics

platform and performing the analysis for our thesis research.

2.4.1 Parallel batch processing with MapReduce and Hadoop

It is hard to predict the size of data and computing power required to process it when dealing with big data. Scaling up ⁴is an option that is always bound by some maximum capacity limits. Also specialized hardware to scale up for higher capacity usually gets very expensive. So the viable option is to scale out ⁵ using required number of smaller machines with relatively low computing resources in parallel. We need a system that can handle large scale parallelization. From programming point of view managing parallel running processes on different machines while ensuring low failure rate is a tough job. So the system should provide programmers an abstraction from lower level system details to enable rapid and fault tolerant development for big data processing. MapReduce is a parallel batch processing framework developed at Google for the purpose of web indexing. The concept of MapReduce was published by Jeffrey Dean and Sanjay Ghemawat in 2008 within their research paper “MapReduce: simplified data processing on large clusters” [8]. This paper describes MapReduce as a “programming model provides a map function that processes a key/value pair to generate a set of intermediate key/value pairs, and a reduce function that merges all intermediate values associated with the same intermediate key. Programs written in this functional style are automatically parallelized and executed on a large cluster of commodity machines. The run-time system takes care of the details of partitioning the input data, scheduling the program’s execution across a set of machines, handling machine failures, and managing the required inter-machine communication”.

Hadoop is the open source implementation of MapReduce developed by Doug Cutting and Mike Cafarella. It was initially created in 2005 to support an open source search engine but then adapted to the published MapReduce framework [8]. It was released by Apache foundation. Apache foundation has also built many supporting tool around Hadoop framework to support end to end big data analytics ecosystem e.g. Apache flume for data collection, Hadoop File system (HDFS) for storing, Apache Pig and Hive for processing,

⁴When the need for computing power increases, a single powerful computer is added with more CPU cores, more memory, and more hard disks and used in parallel.

⁵When the need for computing power increases, the tasks are divided between a large number of less powerful machines with (relatively) slow CPUs, moderate memory amounts, moderate hard disk counts.

Apache Mahout for machine learning etc. We have used some of these tools within scope of thesis research.

MapReduce and Hadoop are batch processing frameworks that empower processing of large volumes of data using commercial grade low cost computing infrastructure. So it supports volume and valuation directly. Variety can also be supported with support of all format files into associated files system e.g. HDFS. Veracity is subjected to supported tools like data collection or data mining tools. Support for such tools is available in Apache hadoop e.g. Flume, Mahout etc. Velocity however is the only feature that a batch processing framework like MapReduce and Hadoop cannot handle. The next subsection answers the question of velocity.

2.4.2 Real time big data processing

Real time data processing is generally associated with live streams of data. Real time data can be processed and analyzed on arrival or it can be buffered for small intervals to provide near to real time analysis. However in many modern data applications instantaneous data need to be analysed in context to large volumes of historic data. To apply advance analytics models like machine learning active feedback loops are also necessary. Even for stored (non live data) big data, applications require data processing system to answer queries very fast. To fulfil these industry driven requirements technology is in rapid advance mode. In last twelve to eighteen months we have seen softwares like YARN (hadoop 2.0), storm, spark, shark , cloudera impala etc with near to real time processing capabilities. On top of it tools like Mlbase and cloudera oryx have started to enable real time advance analytics. Most of these system, frameworks and tools are being developed as the evolution path for MapReduce and Hadoop. All of them have their own purpose, strengths , and limitations. They are mostly used in combinations based on use cases. We shall not be discussing or comparing these systems and tool. Instead, in this article we shall be briefly discussing the two prevailing architectural constructs that can enable real or near to real time big data processing.

2.4.2.1 Lambda architecture

Lambda architecture presents a hybrid model by using fast stream processing together with relatively slow parallel batch processing. It was developed by Nathan Marz on the basis of knowledge and experience he gained from his work with large data sets at Twitter Inc. His approach decompose data processing system into three layers i.e. a batch layer, a serving layer and a speed layer. The stream of data is dispatched to both the batch and speed

layers. Batch layer manages the data set and pre-compute the batch views. Serving layer indexes the batch views so the queries can be served with low latency as compared to traversing through complete data set. Speed layer deals with the recent data thus compensates for the change of data sets during updates of serving layer. An answer to the query is the merged view batch view and the real time view.[28][3]

Figure 2.2 below show the Lambda architecture. Lambda architecture can be implemented using combination of systems and tools e.g. Apache Hadoop along with Apache Storm.

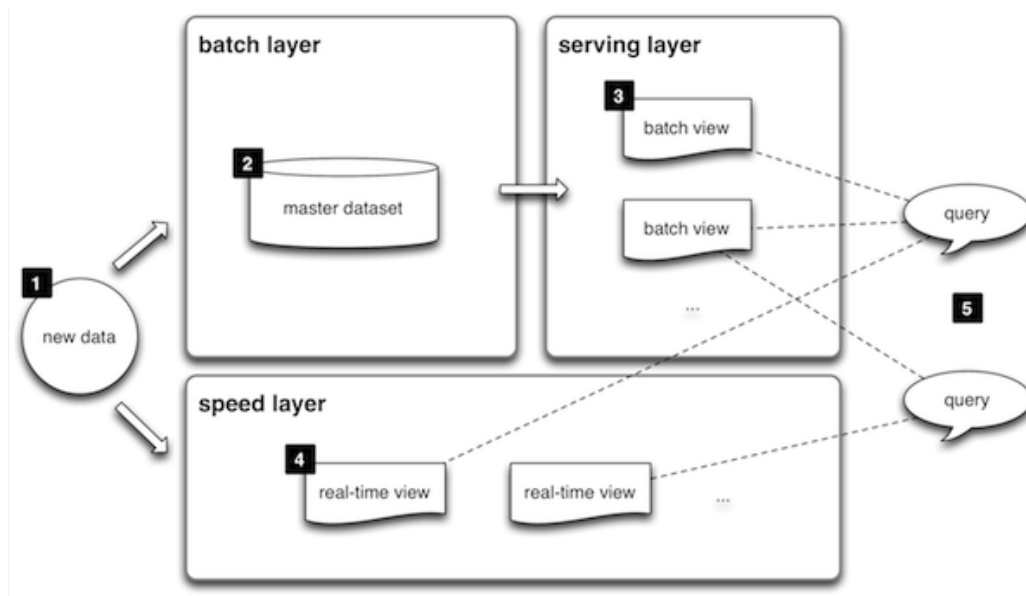


Figure 2.2: Lambda Architecture [3].

2.4.2.2 Massively parallel processing - MPP databases and query engines

MPP based architectures use multiple independent computing resources like servers, processors and storages to execute processing jobs in parallel. Most of the MPP based database approaches implements shared nothing (SN) architecture i.e. a distributed computing architecture in which each node is independent and self sufficient and there is no point of contention across the system [37]. The SN concept for databases was first presented by Michael Stonebraker at University of California Berkeley in 1986 [32].. The SN databases have been very popular in commercial application primarily because of the

high scalability offered by this architecture. Teradata warehousing solutions has been using SN database architectures extensively. Greenplum is an example for open source SN database.

Despite high scalability and other positive aspects, SN databases needs a lot of manual work in terms of partitioning the data, tuning the data and load balancing etc. So usually maintenance such database systems is expensive. MapReduce and Apache Hadoop ecosystem provides high level of automation along with scalability, flexibility and fault tolerance. However parallel batch processing is not as fast SN based MPP databases. Merging both the models solves can solve all these issues. Cloudera Imapala is one of the example of a MPP based online query engine that runs natively on top of Hadoop [1]. It can provide MPP like query response time performance with processing power and flexibility of Hadoop. For our research we have used Cloudera Impala for handling near to real time velocity for big data processing.

2.5 Energy efficiency and eco-efficiency

In previous sections of this chapter, we have highlighted the importance of energy conservation. We discussed the advancements in pervasive smart energy device and grids and their role in improving energy efficiency. We have also discussed the need for collecting and processing large volumes of data from smart energy devices and the available solutions. In this section we shall explain the main motivation and the theoretical concept behind data analysis part of our research.

Unprecedented challenges arising from increasing dependency on conventional energy are part of a global phenomenon. Improving energy efficiency is an important mean to tackle these challenges. Like other economies, European Union is also putting a lot of focus on energy efficiency to ensure energy supply security by reducing primary energy consumption and decreasing energy imports. It helps to reduce greenhouse gas emissions in a cost- effective way and thereby to mitigate climate change [2]. Member states agreed to reduce 20% of the EU's primary energy consumption by 2020 in European Union of council March, 2007. EU's Energy Efficiency Directive 2012 [2] defines energy efficiency as the ratio of output of performance, service, goods or energy, to input of energy. This definition was first discussed in 2006 in European commission action plan for energy efficiency. This generic definition covers all major aspects of the energy efficiency i.e. production, distribution, consumption and the value created in comparison to the resources consumed during the whole process. However, To develop a methodology for measuring energy efficiency and to evaluate the saving, project "Measuring and poten-

tials of energy efficiency (EPO)” was started in 2008[4]. As part of this project VTT published a report “Measuring energy efficiency Indicators and potentials in buildings, communities and energy systems”[12]. This report presents the model for calculating energy efficiency and its correlation with environmental factors. VTT’s research presented in this report considers energy efficiency as a subset of larger eco-efficiency. The ecological factors that can affect energy efficiency are e.g. Temperature, CO₂, NO_x,SO₂ etc. The ecological efficiency itself is a way of measuring sustainable development. VTT summarizes the whole ecosystem in Figure 2.3 below

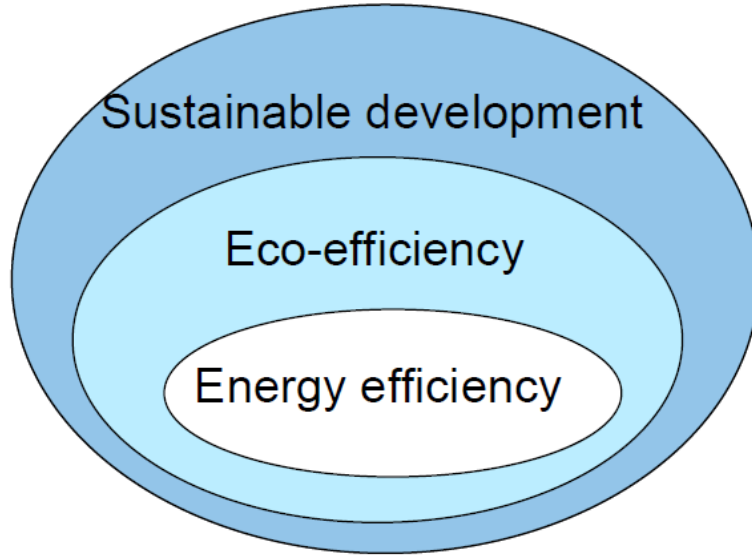


Figure 2.3: Energy efficiency, eco-efficiency and sustainability[12].

The concept of eco-efficiency provides the basis for data analysis in our research. We have applied basic and advanced analytics techniques on data sets collected from building units that are part of VTT’s green campus initiative pilot project with consideration of eco-efficiency model presented in VTT’s report. We calculated energy efficiency of the buildings on basis of formula deduced in Chapter 5 (equation 5.1 and 5.2) of the VTT’s report [12].

$$\text{Energy efficiency of a building} = \frac{\text{Energy consumed}}{\text{Built area}} \quad (2.1)$$

In case of a specific energy consumption (SEC) [12] equation 2.1 can be written as

$$SEC = \frac{Q}{A} \quad (2.2)$$

Where Q denotes the consumption for a single energy type for example electricity and A is the built area in meter square.. In subsequent sections we shall be referring to these equations when we try to identify the usage patterns on building level, discuss the relevance of energy efficiency with these patterns and then discuss a model for classifying buildings on energy efficiency .

2.6 Daily consumption patterns, base load and user load

Daily consumption pattern of a building unit corresponds to the respective usage of the building. Understanding daily usage patterns can help in identifying the optimization point for improving the energy efficiency of that building unit. Base load of a building is one important metric that can be detected through observing the daily consumption. Base load is the consumption that takes place regardless of the actual use of the building and of the user's energy consumption[12]. It is the permanent minimum load that a power supply system is required to deliver. The base load is usually caused by the continuous consumption for building maintenance like air conditioning, ventilation, or night time lighting. Sometimes base load also include some energy consumption by functional components inside building like computer servers, lab equipments, and refrigerators etc. However VTT differentiate this load from user energy load that is characterized by the direct involvement of the users of a building. For example an office building that has peak load during day time because user are using various additional appliances like personal computers, coffee makers, lights etc compared to base load that is generated during night time when office building is not in use. Figure 2.4 illustrates the concept of base load and user load.

Energy efficiency of base consumption and energy efficiency of use shown in figure 2.4 can be calculated using equation 2.1 or 2.2. This provides a weighted metric that can be benchmarked and compared. It can help to narrow down scope of research by referring to problematic buildings and their issues.

2.7 Energy consumption seasonal patterns

Energy consumption has high dependency on seasonal factors like weather. The energy consumption trends vary with outside temperature. Among other things electricity or other energy types required for the air conditioning in

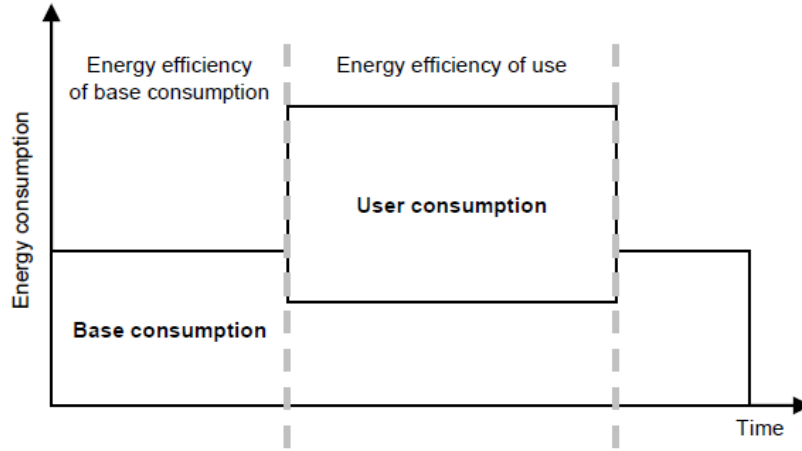


Figure 2.4: Base load , user load and energy efficiency [12].

the buildings is major variable factor dictating the trends. Due to regional weather differences the seasonal energy consumption patterns are also different for different regions e.g. in cold regions of the world energy consumption surges in winters while in warmer regions energy consumption increase is expected in summers because of the air conditioning requirements. Energy service providers usually conduct demand planning with consideration of seasonal trends. Considering seasonal trends is also very important while optimising for gaining energy efficiency.

In scope of our research we have also analysed the seasonal trends. It was not hard for us to perceive the trends while knowing the weather trend for localities of our test building. However, the interesting use case in our research was to check the sensitivity of other consumption patterns and analysis results against the seasonal trend. This will be more explained in the later part of document where we shall discuss the results of our analysis.

Previously, there have been many studies for both daily and seasonal trends in energy consumption. Due to regional differences in trends, many of these studies focused on consumption patterns within a country. Geoffrey K.F. Tso et. al, 2003[36] and Yigzaw G. Yohanis et. al 2007[38] study the energy consumption pattern in Hong Kong and United Kingdom respectively. Buildings units e.g. residential houses apartments and commercial offices etc were considered as basic unit of analysis. Yigzaw G. Yohanis et. al methodology resembles most to our approach as they considered ecological factor along with energy efficiency calculated in similar way as equation 2.1 and 2.2. As discussed before, the main purpose of VTT's green campus

initiative under EcoCampus 2030 plan is to develop a highly efficient model ecosystem for energy production, distribution and consumption that can be expanded further to any scale. Aligned to this goal, we have attempted to provide a data analysis model that is not specific to certain geographic locations. However detailed study is required for adapting such generic models to region specific requirements. In our research we have also attempted to classify the buildings on basis energy efficiency that is explained in next section.

2.8 Classification of buildings based on energy efficiency

Earlier we mentioned that quantifiable energy efficiency through equation 2.1 and 2.2 can be used as a metric for benchmarking and comparison. For energy service providers, governmental energy regulatory agencies or research institute like VTT, it is very important to identify the problematic consumption units in larger number of highly optimized or average performing consumption units. classification of these units into similarly performing groups can help them to narrow down the focus to only problematic units. Sometimes it can also help in understanding the good practices applied by certain consumptions unit that has improved their energy efficiency performance.

Classification for fault detection analysis of a building energy consumption has been used previously as well. Xiaoli Li et. all, 2010 used classification along with outlier detection mechanism to identify the energy inefficient building [24]. They provide a step wise approach to extract the features (types of energy, trends etc) from the data collected as a time series. Then detect identify the daily usage patterns using auto regression technique and pass the results to benchmark against any outlying data point that can refer to faulty behaviour. Imran Khan et all. 2013, proposes different clustering techniques to group building with similar level of energy efficiency together[21]. In our research we used a hybrid method using feature extraction and trend detection techniques like [24] and then applied a clustering technique proposed in [21]. The clustering technique that used is called K-means clustering. It is explained in the next subsection of this article.

2.8.1 K-means clustering

K-means is an algorithm for cluster analysis. In context to machine learning cluster analysis or clustering is an unsupervised task of grouping a set of

objects in a way that objects in same group are similar to each other more than the objects in other group. K-means algorithm clusters the set of objects i.e. energy efficiency values in our case into predefined number of classes. We shall term these values as data points. K represents the number of cluster and groups that we can set in start of the process. K-means means algorithm was first proposed by Stuart Lloyd in 1957[25] but the k-means term was first used by James Mcqueen in 1967[26]. There have many adaptations and optimizations in Lloyd's basic algorithm. K-means algorithm today has many variants like Fuzzy C-means clustering, k-medoids and spherical means etc. Even for original Lloyd's algorithm there has been some modification in methodology. Two very commonly used methods are Forgry method [11] and Hartigan-Wong method[17]. In our approach we are using Hartigan-Wong method. We shall also use some references from Forgry method when explaining the K-means algorithm.

K-means groups the data points in cluster with a logical centre point. The aim of the K-means algorithm is to divide data points in certain dimensions into K clusters so that the within-cluster sum of squares is minimized [17]. Lets assume if we want have K cluster for data points $D = \{x_1, x_2, \dots, x_n\}$ in d dimensions then

$$x_i \in R^d$$

K-means algorithm uses following steps to cluster data into groups[30].

1. Initialize the centroids randomly for each K i.e. for each group.
2. Data points are assigned to closest centroid.
3. Move the centroids to the mean of the data points assigned to that centroid in step 2.
4. Repeat 2 and 3 till convergence Convergence means that values stop changing in further iterations.

Mathematically randomly initialized centroid are

$$\mu_1, \mu_2, \dots, \mu_k \in R^n$$

If c^i is the distance of centroid to assigned data point then Step 2 and 3 with recursive distance minimization and mean adjustment can be explained as

For every i, set

$$c^i := \arg \min_j ||x^i - \mu_j||^2 \quad (2.3)$$

The equation above used Euclidean distance formula for calculating distance between centroid and data point.

For every j , set

$$\mu_j := \frac{\sum_{i=1}^n 1\{c^i = j\}x^i}{\sum_{i=1}^n 1\{c^i = j\}} \quad (2.4)$$

The input to k-means is a set of feature vectors along with the number of clusters required. In our case we shall have two feature hence two dimensional matrix of energy efficiency values for electricity and electricity used for heating. Before inserting data to k-means it is required to set the similar scale for features as well set the standard variance to avoid errors in the results. We required to classify pilot site buildings into four groups with High efficiency, moderate efficiency, low efficiency and poor efficiency classes. So we have set K value as 4.

2.9 Forecasting the energy consumption

Estimating equipment specific energy consumption has been a key focus area for energy service providers. It can help in demand planning, load forecasting, and understanding end user behaviour. Energy service providers can design better service offerings for their consumers. Unit energy consumption (UEC) is a term generally used for estimating equipment specific energy consumption. It is the average annual amount of energy consumed by a user device. Conditional demand analysis (CDA) model has been one the most commonly used method for UEC estimations. K. H. Tiedemann, 2007 explain CDA as a multivariate regression technique which combines utility billing data with weather information and customer survey data to produce robust end-use energy consumption estimates[35]. As part of the green campus project VTT has installed used state of the art nonintrusive load monitoring (NIALM)[16] devices that can distinguish between the usage of different electric devices on basis of changes in voltage and electricity.

We are using the data collected by a NIALM device installed in one of a residential apartment included in VTT's pilot test sites. In our analysis of this data we are not using CDA. However we shall be using auto regression along with concept of moving averages in form of a model known as ARMA model to estimate the future consumption of a device depending on the previous usage. This is an example of quantitative forecasting. Before we go to discuss about ARMA models it is important that we briefly discuss the

basic conditions for quantitative forecasting and the time series analysis as foundation for prediction model based on ARMA.

2.9.1 Main conditions and Steps for Quantitative Forecasting

Rob Hyndman et al, 2014 discuss two main conditions for application of quantitative forecasting[19] in their book “Forecasting: Principles and Practice”.

1. numerical information about the past is available.
2. it is reasonable to assume that some aspects of the past patterns will continue into the future.

In case the conditions can't be met then qualitative forecasting is the only option. However, the qualitative forecasting is not in scope of our research for this thesis. In the same book authors mention following five step approach for solving forecasting problems.

1. Problem definition
2. Information gathering that includes statistical data collection.
3. Exploratory analysis of the data to evaluate the structure of the data and observing relationship between different variables.
4. Choosing and fitting the forecasting model. The model depends upon the relationships of variables. Every model has its own construct. So data needs to be fitted to that construct before applying that model. We shall discuss it more in data analysis part of this document.
5. Using and evaluating forecasting model. It generally includes comparison of results after applying different models.

2.9.2 Time Series Analysis

Time series is the sequence of a random variable collected over time. Among other examples of time series data, energy consumption data from metering devices can also be collected periodically hence constituting a time series. Comparison of a single time series at different point in times is termed as time series analysis [5]. A time series usually consists of a deterministic component and a random component[29]. So if X_t is a time series data then we can have

$$X_t = d_t + \epsilon_t \quad (2.5)$$

where d_t is the deterministic component and ϵ_t is the random component. The deterministic component itself can be in form of trends, periods, and jumps etc. Figure 2.5 illustrates the examples of different time series. In each illustration there is atleast one stochastic random component with and without deterministic components.

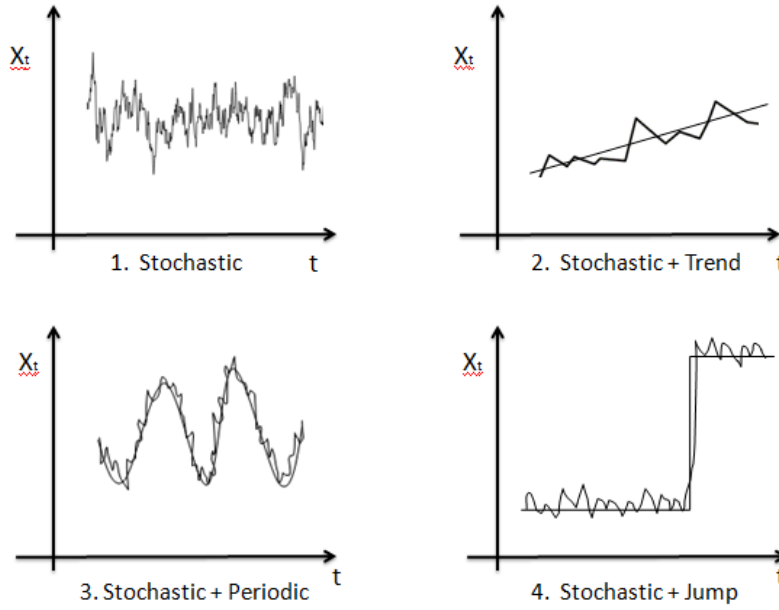


Figure 2.5: Time Series types[29].

In figure 2.5 illustrations 2,3 and 4 contain a deterministic component with a random component. In terms of prediction even the random component can be estimated using the deterministic component. However for stochastic random time series data without any deterministic component it is very hard to predict anything accurately. The time series with no predictable pattern is generally termed as stationary time series. We shall be discussing this in details during the analysis part of this document.

2.9.3 Autoregression, Moving Averages and ARIMA Models

Rob Hyndman's book "Forecasting: Principles and Practice"[19] is the main reference for this section.

2.9.3.1 Regression

The concept behind basic regression techniques for forecasting is that we try to forecast a variable ‘y’ on the basis of another variable ‘x’. For example a liner regression model forecast y assuming it has a linear relationship with variable x e.g. as in equation below.

$$y = \beta_0 + \beta_1 x + \epsilon$$

Parameter β_0 and β_1 represents the intercept and slope respectively for the line representing the linear relationship. β_0 represents the predicted value when x is 0. Linear regression for time series analysis can be written as

$$Y_t = \beta_0 + \beta_1 x_{t-1} + \epsilon$$

Here Y_t is the estimate with past value of x_t i.e. $\{x_1, x_2, \dots, x_{t-1}\}$. using differencing⁶ error e_t in estimation can be calculated as

$$e_t = X_t - Y_t = x_t - \beta_0 - \beta_1 x_{t-1} - \epsilon \quad (2.6)$$

2.9.3.2 Autoregression

Autoregressive model is based on the concept of a variable regressing on itself. For autoregression we can drive equation as

$$x_t = \beta_0 + \beta_1 x_{t-1} + e_t + \epsilon \quad (2.7)$$

The aim for good estimation is to select value of β_0 and β_1 that can minimize the sum of square of errors. Above equation can be used to estimate the value based on first previous value. But in case we want to estimate based on multiple previous values e.g. ‘p’ values then we can write it as

$$x_t = c + \beta_1 x_{t-1} + \beta_2 x_{t-2} + \dots + \beta_p x_{t-p} + e_t + \epsilon \quad (2.8)$$

We just replaced β_0 with a constant c as it is a constant value. Adding the summation to the historic values we can write

$$x_t = c + e_t + \sum_{i=1}^p \beta_i x_{t-i}$$

we have also taken out the random component ϵ that does not meet the basic conditions for forecasting as described in subsection 2.9.1. The model presented in equation 2.8 is referd to as AR(p) model.

⁶The differences between consecutive observations

2.9.3.3 Moving Averages

Moving averages model use past forecast errors in regression like manner to forecast future time series values instead of using past time series values as in autoregression. Mathematically model can be explained as

$$y_t = c + e_t + \theta_1 e_{t-1} + \theta_2 e_{t-2} + \dots + \theta_q e_{t-q} \quad (2.9)$$

OR

$$y_t = c + e_t + \sum_{i=1}^q \theta_i e_{t-i}$$

The model presented in equation 2.9 is termed as MA(q) model. In this model each value of y_t can be thought of as a weighted moving average of the past few forecast errors.

2.9.3.4 ARIMA Model

ARIMA stands for Auto-Regressive Integrated Moving Average. As the name suggests it is the combination of autoregression and moving average models. ARIMA is one of the most commonly used forecasting technique. ARIMA model can handle time series data with and without seasonality. We shall be discussing non-seasonal ARIMA because of the nature of the data we shall be processing in our analysis. The nature of the data will be discussed in data analysis part of this document. So combining the autoregression and moving averages using equation 2.8 and 2.9 we can have

$$y'_t = c + e_t + \phi_1 y'_{t-1} + \dots + \phi_p y'_{t-p} + \theta_1 e_{t-1} + \dots + \theta_q e_{t-q} \quad (2.10)$$

In this equation y'_t is the difference series. This constitutes ARIMA(p, d, q) model where

- p is the order of autoregression.
- d is the number of non seasonal differences.
- q is the order of moving averages.

Now to simplify the complex time series equation back shift notations are usually used e.g. y_{t-1} can be denoted by By_t i.e.

$$By_t = y_{t-1}$$

AND

$$B(By_t) = B^2 y_t = y_{t-2}$$

AND

$$y_t - y_{t-2} = (1 - B)y_t$$

In general a d th order difference is written as

$$(1 - B)^d y_t$$

Rearranging equation 2.10 and using backshift notations we can have following equation with labeled p, d and q for ARIMA model.

$$\underbrace{1 - \phi_1 B - \dots - \phi_p B^p}_{\text{AR}(p)} \underbrace{(1 - B)^d y_t}_{d \text{ differences}} = c + \underbrace{(1 + \theta_1 B + \dots + \theta_q B^q) e_t}_{\text{MA}(q)}$$

Explanation and the equations used in section 2.9.3 were cherry picked from Rob Hyndman's book "Forecasting: Principles and Practice" [19] as reference to theory related to our research. For further details please refer to chapter 5 and chapter 8 of this book.

Fitting the ARIMA model and estimating the future time series values need intensive computation. We shall be using software e.g. R to solve these equations for our use cases.

Chapter 3

Methodology

In previous chapters we introduced our research problem, listed and explained the solution options with the theoretical background. In this chapter we shall try to explain our practical approach for carrying out the research along with the software development required to support the experimentation and data analysis for our research. On a practical level following were some major tasks that were required to fulfil scope of our research.

- Understanding energy efficiency, smart grids and available data.
- Requirement engineering and use case preparation.
- Understanding Data Analytics ecosystem, evaluating the big data tools and solutions.
- Exploratory data analysis and selection of algorithms and data analysis tools with respect to use cases.
- Developing an end to end big data analytics platform.
- Data collection, storage and preprocessing.
- Use case specific data analysis and evaluation of results.
- Visualization of results
- Documentation of the research, process, software development and results.

Some of these tasks were required to be performed in a sequential way e.g. requirement engineering and evaluation of big data tools were required before developing the big data analytics platform or selecting the algorithms. Similarly we need to have results before visualizations could be created. On the

other hand some of the tasks could have been executed in parallel. For example the documentation was an ongoing process along with all other tasks. Similarly literature review for understanding each component of our research was also an ongoing process through the time line for this thesis. The regardless of sequential and parallel tasks we need to iterate for continuous improvement.

To tackle these challenges we needed a methodology that can support sequential and parallel task execution with support for iterations to improve. Like most of scientific researches, fail fast and small to succeed was the key for us. In the list of tasks mentioned above. Most of them requires conceptualization and tested quickly using rapid prototyping. Taking it as a software development task initially, we had some candidate models such as water fall model, agile development model, spiral model and incremental model etc. Here we shall briefly discuss the advantages and disadvantages in context to our research project.

- **Waterfall model** offered the simplest approach of requirement engineering, design, implement, test and operate our research. However it is inherently sequential and had weak support for iterations.
- **Agile development model** Agile methodology[27] is rapid, iterative and supports quick prototyping but it requires additional communication and management overhead like scrum meetings. Managing it along with stakeholders like VTT and CIVIS projects was very hard.
- **Spiral Model** is a risk driven process model. It supports prototyping, provides good way of avoiding major failure risks, it is iterative. However it needs a lot of resources during planning phase specially when the spiral keeps growing in size. It is usually very successful for large projects but it has overheads for small projects like our thesis research. We shall be discussing more about using parts of the spiral model later in this chapter.
- **Incremental model** relies on small incremental steps with each step consist of independent design, implement and test phase. In the beginning, Incremental model was the best fit among other candidate models for our thesis research. We were able to prototype small functional units of the big data analytics platform very quickly while independently working on the use cases. However during platform development and data analysis part. It has started creating integration overheads. For example integrating two different data processing tools together for a single use case becomes difficult when they were configured in two different incremental steps.

Learning from the problems that we face from incremental model we altered our approach to adapted version of another very flexible software research and development methodology known as “Kumiega-Van Vliet Trading System Development Methodology” [22].

3.1 Kumiega-Van Vliet Model

Kumiega-Van Vliet Trading System Development Methodology ($K|V$) was developed in 2008 for software development required specifically for trading systems. It is the combination of three general purpose software and new product development models i.e. waterfall model, spiral model and stage gate model. We have already explained the waterfall and spiral models. Stage gate model consists of stages e.g. scoping, development, implementation, testing etc. Each stage or combination of stages can be controlled with an approval gate. Process can not move from a stage to other stage if the gate in between them is not approved. This model provides a good control over the development model to ensure quality. However it may cause delays because of the organizational hierarchies dictating the gates.

($K|V$) model tries to overcome the short comings of these three models by combining them to a single paradigm for trading system development [22]. In spiral model in start smaller time is allocated to four basic steps i.e. research planning implementation and test. These four steps can be performed again and again in cycles. To avoid spiral to grow too much after each cycle a stage gate controls if process can be passed to next stage or it needs to be sent back to perform another cycle in same stage. Just like waterfall there can be number of stages. But for continuous improvements process there is an iteration channel available unlike traditional waterfall model.

3.2 Adaptation of Kumiega- Van Vliet Model

($K|V$) model is designed for software research and development in domain of financial services. With the built in stage gate controls it requires some scale of hierarchical organizational structure to support the model. For our highly academic research case we have made certain adjustments. The most notable adaptation was to use deliverables and team reviews of respective deliverables as the main control for moving from one stage to other instead of stage gate approvals in ($K|V$) model. The waterfall model like stages helped in keeping our focus on the solution for our problem statement. The spiral model cycles enabled us to iterate within a stage and improve the

deliverables quality. Typically the decision of additional cycles was based on the feedback during the team review sessions. The inter stage iteration channels helped us in improving our overall quality. The lessons learnt or the new directions identified during one iteration was include in the scope of research for next iteration. It also allowed us to include supplementary topics in our scope without losing focus on mandatory issues.

In our approach, we have divided complete scope of research in four basic stages. Within each stage we had four steps. These intra stage steps were different for each stage. These steps were corresponding to the main tasks that we discussed in start of this chapter. A typical intra stage cycle ended with a set of deliverables. The deliverables were reviewed in a team review session. If required the other stockholders like VTT was also involved in some of the review meetings. We shall be discussing it in details when we describe our stage wise proceedings. At end of each review session a decision was made to either move to next stage or try to imrove via additional cycle. Using all four intra stage steps for additional cycles was not a must. This was another minor adaptation to the $(K|V)$ model. Similarly iterations were were mostly initiated after stage three. There were three major iterations. During the iterations change of deliverables were not mandatory. However in practice it was observed that iteration had caused some major or minor changes in stage deliverables as well. Small informal team structure reduced management and communication efforts. This also helped in rapid processing during iterations. Figure 3.1 illustrates our approach with the adapted version of $(K|V)$ model. Stage by stage description of our research methodology is explained in next section.

3.3 Stages, steps and cycles

We have already mentioned that there were four stages with each having four respective steps. Each stage was controlled via deliverables review sessions in a stage gate manner. While inside a stage, steps were executed in spiral cycles. First cycle of the spiral had to pass through the four steps. The additional cycles are initiated if the further improvements are decided for deliverables in review session. All Four steps were not mandatory for additional cycles. In this section we shall be listing and describing the stages along with respective steps. We shall be highlighting some major cycles and deliverables. However the iterations will be discussed in next section. Figure 3.1 will be our main reference through out this section. In this section we shall mention some functional components of our project e.g. logical architecture, data processing tools and algorithms etc. Details for these functional

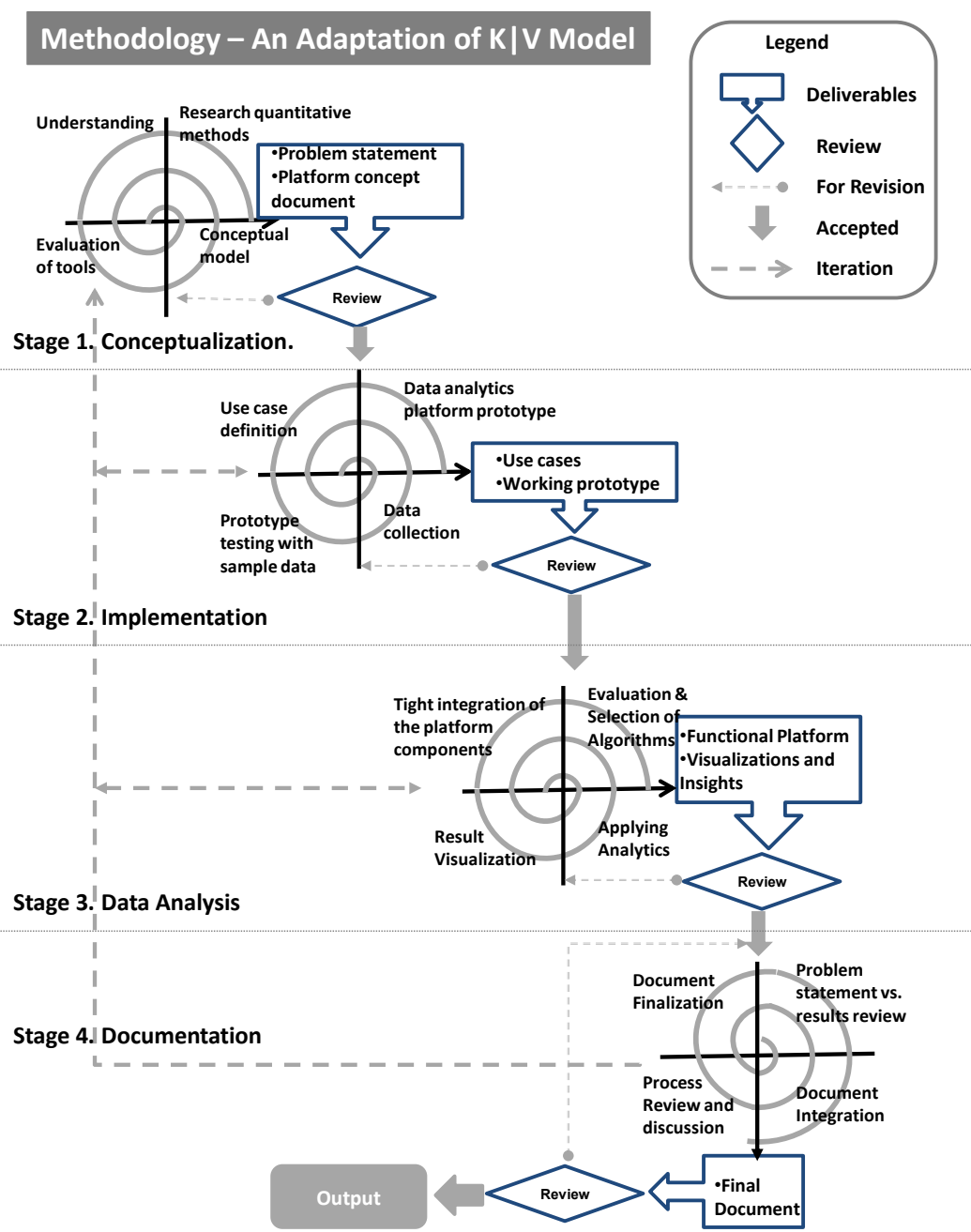


Figure 3.1: Methodology, An Adaptation of $K|V$ Model

components will be given in later part of this document.

3.3.1 Stage 1. Conceptualization

In start our research problem was mainly concerned about processing large volumes of data coming from smart metering devices and understand the consumption patterns. So the primary focus of the conceptualization stage was to describe our problem in detail, understand important factors related to it, find and evaluate methods and tools to solve the problem. The stage had following four steps.

3.3.1.1 Step 1. Understanding

From the start, our research had two focus areas i.e. energy consumption and big data . The main purpose of this step was to understand important concepts related to these topics. Follwoing are some main activities performed during this step.

- Intensive literature review.
- Participation in CIVIS project Helsinki- Use case workshop 26-27 January 2014. It gave good insights about ecological and social factors effecting energy production, distribution and consumption.
- Participation in VTT Green Campus Initiative Introduction session.
- Discussions and informal interviews with VTT's project lead for Green Campus Initiative.
- Aalto University courses.
 1. Scalable Cloud Computing, as a good introduction to parallel batch processing and its uses for big data processing.
 2. Information Visualization, as an introduction to effective communication through data visualization.

Literature review had been a constant step through out this stage, cycles, and iterations.

3.3.1.2 Step 2. Research quantitative methods

This step involved finding and evaluating the various quantitative methods used for measuring energy consumption and benchmarking energy efficiency. Data aggregation methods like daily, monthly consumption, and average consumption etc were evaluated. Identification and theoretical evaluation of advance analytical methods was also performed during this step.

3.3.1.3 Step 3. Conceptual Model

This step was dedicated for finding available open source solutions to make a conceptual model for an end to end big data analytics. This step was mandatory for the big data platform concept paper deliverable. This step was also repeated during various iterations, whenever change was required in data platform.

3.3.1.4 Step 4. Evaluation of Tools

This step was in pair with 3.3.1.3. All the tools listed in conceptual model were tested during this step. A checklist of evaluated and selected tools was maintained. This list is available as annex???

3.3.1.5 Deliverables of stage 1

There were two deliverables of this stage

1. Problem Statement. first two steps of this stage were the main contributors for this deliverable.
2. Platform concept document. A document as result of step 3 and 4 of this stage.

3.3.1.6 Stage 1 cycles

In this stage we observed two cycles i.e. cycle for producing the required deliverables and one additional cycle for modification of platform concept document. The main modification in additional cycle was the replacement of application frame work with an architecture diagram to clarify.

3.3.2 Stage 2. Implementation

This stage mainly includes requirement engineering and intensive software development to prototype and test the big data platform described in concept paper as a deliverable from stage 1. This stage had following four steps.

3.3.2.1 Use case definition

In this step, Based on the knowledge gained from stage 1. we decomposed our problem statement into lower level requirements that can be practically implemented using big data platform concept. Use cases went through several iterations. Details of iterations will be discussed later in this chapter. However here we shall list the final list of use cases.

1. Understanding the seasonal energy usage patterns and its sensitivity with outside temperature.
2. Understanding characteristics of building using daily energy consumption pattern.
3. Calculate the base load of the building to identify non user consumption of buildings
4. Classify building on basis of energy efficiency and analyse seasonal shifts in this classification.
5. Predict daily energy consumption of various house hold devices on basis of previous consumption pattern.

Chapter 4

Methods

You have now stated your problem, and you are ready to do something about it! *How* are you going to do that? What methods do you use? You also need to review existing literature to justify your choices, meaning that why you have chosen the method to be applied in your work.

If you have not yet done any (real) methodological courses (but chosen introduction courses of different areas that are listed in the methodological courses list), now is the time to do so or at least check through material of suitable methodological courses. Good methodological courses that concentrates especially to methods are presented in Table 4.1. Remember to explain the content of the tables (as with figures). In the table, the last column gives the research area where the methods are often used. Here we used table to give an example of tables. Abbreviations and Acronyms is also a long table. The difference is that longtables can continue to next page.

Code	Name	Methods	Area
T-110.6130	Systems Engineering for Data Communications Software	Computer simulations, mathematical modeling, experimental research, data analysis, and network service business research methods, (agile method)	T-110
Mat-2.3170	Simulation (here is an example of multicolumn for tables)	Details of how to build simulations	T-110
S-38.3184	Network Traffic Measurements and Analysis	How to measure and analyse network traffic	T-110

Table 4.1: Research methodology courses

Chapter 5

Implementation

You have now explained how you are going to tackle your problem. Go do that now! Come back when the problem is solved!

Now, how did you solve the problem? Explain how you implemented your solution, be it a software component, a custom-made FPGA, a fried jelly bean, or whatever. Describe the problems you encountered with your implementation work.

Chapter 6

Evaluation

You have done your work, but that's¹ not enough.

You also need to evaluate how well your implementation works. The nature of the evaluation depends on your problem, your method, and your implementation that are all described in the thesis before this chapter. If you have created a program for exact-text matching, then you measure how long it takes for your implementation to search for different patterns, and compare it against the implementation that was used before. If you have designed a process for managing software projects, you perhaps interview people working with a waterfall-style management process, have them adapt your management process, and interview them again after they have worked with your process for some time. See what's changed.

The important thing is that you can evaluate your success somehow. Remember that you do not have to succeed in making something spectacular; a total implementation failure may still give grounds for a very good master's thesis—if you can analyze what went wrong and what should have been done.

¹By the way, do *not* use shorthands like this in your text! It is not professional! Always write out all the words: “that is”.

Chapter 7

Discussion

At this point, you will have some insightful thoughts on your implementation and you may have ideas on what could be done in the future. This chapter is a good place to discuss your thesis as a whole and to show your professor that you have really understood some non-trivial aspects of the methods you used...

Chapter 8

Conclusions

Time to wrap it up! Write down the most important findings from your work. Like the introduction, this chapter is not very long. Two to four pages might be a good limit.

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Appendix A

First appendix

This is the first appendix. You could put some test images or verbose data in an appendix, if there is too much data to fit in the actual text nicely.

For now, the Aalto logo variants are shown in Figure A.1.



(a) In English



(b) Suomeksi



(c) På svenska

Figure A.1: Aalto logo variants