

Final Year Project Report

Automated Software to Understand Functional Relationship Between Dynamic Energy and Performance Events

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Project Specification

General Information:

A energy model representing a relationship between dynamic energy consumption and performance events (PMCs) is constructed experimentally and the experimental dataset has the following format typically (k events, n records):

$$\begin{matrix} E_1, & x_{11}, & x_{12}, & x_{13} \dots x_{1k} \\ E_1, & x_{21}, & x_{22}, & x_{23} \dots x_{2k} \\ \dots & & & \\ E_n, & x_{n1}, & x_{n2}, & x_{n3} \dots x_{nk} \end{matrix}$$

where E_i is the experimentally obtained dynamic energy consumption of i-th record and x_{ij} are the experimentally obtained performance events (PMCs).

Given such an experimental dataset as an input, the goal is to determine/understand the functional relationship between the dynamic energy consumption and performance events (PMCs).

Two real-life datasets will be provided to the student.

Core:

The goal is to write automated software that will detect the following:

1. Existence of records where the dynamic energy consumption is the same (within an input tolerance) but all PMCs (with the exception of one) have same values. Then the relationship between energy and the one PMC is visualized to see the nature of the functional relationship.
2. Having accomplished step (1), understand the monotonicity of the relationship between dynamic energy consumption and performance events (PMCs).
3. Existence of records where the dynamic energy consumptions are different (within an input tolerance) but all PMCs have same values (within an input tolerance) suggesting the non-existence of a functional relationship.

The software must be written using any one mainstream language but preferably one of the following: C, C++, Python

The software must be well documented and tested.

Advanced:

Given an experimental energy model dataset as an input, the goal is to write software that performs intelligent but computationally feasible simulations where combinations of inputs are generated to study the existence/non-existence of a functional relationship between dynamic energy consumption and PMCs.

The software must be written using any one mainstream language but preferably one of the following: C, C++, Python.

The software must be well documented and tested.

Abstract

With the advent of technology, the demand of energy has also increased dramatically. Increasing number of electric driven equipments such as personal devices, hybrid vehicles and embedded systems has made energy management crucial. To help manage the dynamic energy consumption by the systems. This project attempts to create a software that tries to analyse the dependence of some low-level events with the energy consumption by finding the existence of functional relationship between them. Our project detects and analysis the functional dependence rather than recognising the exact form of dependence. This acts as a form of test to determine the nonexistence of a solution to prevent the unneeded search. It also tries to understand the behaviour and contribution of various parameters towards the output. To understand the behaviour linearity between input parameters and output is assumed.

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Chapter 1: Introduction

1.1 Motivation

Modern day technology has developed under incredible speed in recent decade and the computing power growth rate is truly phenomenal and lasting impact can be felt and benefit us in many ways. It is important to realise the worldwide effect on environment by the increase in consumption of power by these technology advancements.

According to [4], the power consumption growth rates of PCs are about 7.5% per year. Data Centres and network play much larger role as they both have power consumption rate of 12% each. This considerable growth is due to increasing data to be accessed, stored and processed. This constant expansion of energy consumption leads to increase in carbon emissions. CO_2 emissions from ICT (Information and communications technology) are increasing at a rate of 6% per year, at such rate by 2020 it will account to 12% of worldwide emissions [5].

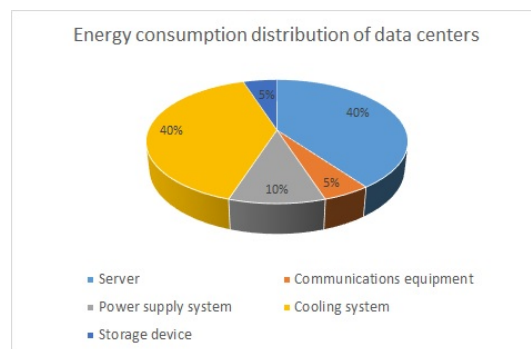


Figure 1.1: Energy consumption distribution of data centers.

The major roles of power consumption by a data ware house is played by the servers and the cooling system which is used to cool down the server physical parts. From the figure 1.1, you can see they both account to 80% of the total consumption with both accounting for 40% each [5]. Hence by finding a relation between different events and energy consumption, we could minimize the power consumption by the systems. Minimising energy consumed will also decrease the heat generated by the system which will in turn lower down the usage of cooling system. Leading us to minimise cost and save environment.

Thus, if we could discover the functional relational between the demand of the power and the performance events (PMCs), that could enable us the ability to adjust, and predict the energy consumption for certain computations.

1.2 Aims

The projects aim is to create a piece of software that is extensible, easy to use and performant. A tool that can perform analyses like regression and clustering and give us an overview of the

data by showing the relationship of various parameters with the output. In this case, our parameters are the performance events and output is the energy consumption. Following are the objectives that we want to achieve.

1.2.1 Existence of functional relation

Datasets are mostly pair of multiple inputs and a single output. And the possibility of the existence is what we are interested in. Is it possible to define the data in a form of function? Question like this is one of the aims. Defining the data in a form of function is explaining the dataset in a form of a formula which can combine various parts of the input parameters and give the output. It is quite impossible to tell that whether a function definitely exists as the data we have is always a subset and there are number of records which are either not in the dataset or we don't know. But, regardless we can definitely explain the non-existence of a formula by finding a pair of input parameters and outputs that violate the definition of a function.

The pair which violates the definition of a function must be looked at as there is high probability of that data set record being corrupt or if not it can actually help and tell the non-existence. If this outlier is in fact corrupt, this process can be used for data cleanups as well.

1.2.2 Analysis of functional relation

Once, we are no more able to prove functional non-existence in the dataset. We can try to analyse the dataset and try to examine the various relations of the parameters with the output. Analysis of the parameters with output is done to know how much change in one parameter contributes to the change in the output. This analysis helps us to understand the correlation between two quantities. Does high page faults correlate to high energy consumption? Questions like this is what the analysis of parameters with their corresponding output answers. This can also be thought of as an explanation of the output parameters.

In other words, understanding the correlation is the other aim. Once we are able to know that there is high correlation. We can then dig deep into the data and try to find out the reason for the same.

1.3 Approach

For each of the aims, two approaches have been employed. Both the approaches try to achieve the same conclusions with different methodology.

The experimental data sets provided has the following format:

$$\begin{matrix} E_1, & x_{11}, & x_{12}, & x_{13} \dots x_{1k} \\ E_2, & x_{21}, & x_{22}, & x_{23} \dots x_{2k} \\ \dots & & & \\ E_n, & x_{n1}, & x_{n2}, & x_{n3} \dots x_{nk} \end{matrix}$$

where k is the total number of parameters (performance events), n is the number of records in the dataset. E_i is the experimentally obtained dynamic energy consumption and x_{ij} are the

experimentally obtained performance events (PMCs).

1.3.1 Existence of functional relation

Our main goal here is to find nonexistence of functional relationship. In other words, it means proving that the dataset cannot be explained in terms of a function/formula.

Our first approach here is to find two performance events tuples $(x_{i1}, x_{i2}, x_{i3} \dots x_{ik})$ and $(x_{j1}, x_{j2}, x_{j3} \dots x_{jk})$ are equal within some tolerance, but their corresponding E_i and E_j dynamic energy consumption is different. Existence of such a tuple in database will lead us to prove the non-existence of a functional relationship.

It is infact easier to find two equal records by ordering the dataset by their parameters and going through the entire dataset once to find equal records and comparing their dynamic energy consumption. The order of complexity is $O(N \log N)$ which is the complexity of sorting as that is the only heavy duty task involved. But it gets really complicated when tolerance comes into play. In other words, we have to find similar records than equal records. The project employs 2 methods to measure the similarity between the two data records.

First approach:

The data records are imagined as data points in $k - space$ as we have k number of parameters. Then we divide the whole space into small $k - dimensioncubes$ whose dimensions $(t_1, t_2, \dots t_k)$ where t_i is the tolerance of each parameter. The data points are then put in their respective cubes. And each cube is then analysed to see that whether the data points output are similar to each other as well.

Second approach:

The data points in $k - space$ are similar if the euclidean distance between them is less than the t provided, where t is the total/max tolerance. Data points which are infact found to be close to each other, their output is compared to see if they are similar to each other.

1.3.2 Analysis of functional relation

Many different type of relations can be there between two variables. But, we are interested in more general trend of the dynamic energy consumption output with the performance events. Many researches have used linear models, to understand the relation between energy consumption and performance events. [3] And this is due to the trickle down effect of the relation. [1] Hence, we also used linear regression to analyse the relationship.

Linear regression is done with 2 variables but we have k variables relation to energy consumption making it hard to understand the nature. Hence to analyse one of the variable we have to isolate it in the space where the other $k - 1$ variables donot change. Data points are clustered using $k - 1$ variables. Each cluster is then visited and linear regression is performed. Two approaches are employed to cluster which are similar to finding similar records in finding the existence of the relation.

First approach:

Data points are clustered by putting the data points in the respective $k - 1dimensioncube$. Forming a cluster with almost similar $k - 1variables$.

Second approach:

Euclidean distance is used to isolate similar $k - 1$ variables. A $k - 1$ dimensionsphere can be imagined for simplicity with its radius being the total/max tolerance.

In both approaches, linear regression is performed on the cluster with the variable which was not used in clustering and the output which is the dynamic energy consumption.

Above is just only one field that possible usage of the functional relationship in data. In real world there are much more application that required such technique. Such as hybrid vehicle power management, power supply of auxiliary power units etc. There are many past work in this field but only few were focus on the relationship between dynamic energy consumption and PMCs. Hence, in this report we will try to observe the relationship between energy consumption and PMCs. We will explore the monotonicity of the relationship between dynamic energy consumption and performance events (PMCs) and suggesting the non-existence of a functional relationship as well.

1.4 Structure of report

In this report, you must have already seen the motivation behind the project and brief description of the approaches that will be taken to reach the objective.

This Introductory chapter is followed by Background research, design aspects of the software, implementation of the software, testing and evaluation of the tool followed by the conclusion and the future works.

Background research explains about performance events and how do they influence energy consumption by the system. It also explains the approaches in detail mentioned above and analyses their complexity and how they can be optimised.

Following Background research, design and implementation of the software is deep dig into as we want an extensible, performant tool. Testing and evaluation of the software is explained. How the software is tested and its evaluation on its performance. The report ends with conclusions and future works that shows how the project can be extended and applied to various other domains.

Chapter 2: Background Research

Many designers are increasingly utilizing dynamic hardware adaptations to improve performance while limiting the power consumption. Some are using software to decrease power usage for e.g. putting the system in sleep mode when it's in idle state. The main goal remains the same, which is to extract maximum performance while minimizing the temperature and power. Whereas, we want to study and examine the relationship affecting the consumption and then analyse the result to minimize or predict the consumption of energy.

2.1 Energy consumption and Performance Events

First let's look at energy consumption. Energy consumption is the power (Usually in watts) consumed by a system. This system could be the processor/CPU, memory, disk, I/O (Input/Output) system, chipset or the whole computer system itself. So, one can take any of the peripherals and read the power consumption for various performance events. Then analyse if there exists a functional dependence to begin with. If the system is not able to disprove the dataset, one could then try to find a function which could help understand relation between each event and predict for any given system. Reading of these performance events can be during a idle state as well as running certain computations.

Now let's look at what are performance events, performance events can be any event which can affect the consumption of energy in some way. Selection of performance events is quite challenging. A simple example would be the effect of cache misses in the processor. For a typical processor, the highest level of cache would be L3 or L2 depending on the type of processor. Now for some transaction which could not be found in the highest level of cache (cache miss) would cause a cache block size access to the main memory. Thus, number of main memory access would be directly proportional to the cache misses. Since these memory access is off-chip, power is consumed in the memory controller and DRAM. Even though, the relation is not simple as it seems but we can see a strong casual relationship between the cache miss and the main memory power consumption [1].

We can use number of other performance events like Instructions executed. As we know on each instruction being executed, more units of the system are on. Hence, power is consumed as opposed to when the processor is in its idle state [2].

Cache miss, TLB misses are also a good performance events as they seem to have a strong relationship between the power consumption as processor needs to handle memory page walks. Same can be said for Page faults where a program is not able to find mapped address in physical memory as it has not been loaded yet. This causes a trap which can result into number of situations, one of them which is to get the data from disk. In simple terms it is longer walk from cache miss. This walk to the disk and raising of exceptions would consume more energy by the disk as well as the CPU.

2.2 Proofs

The Proofs below are for different approaches that have been discussed to find the non-existence of a functional relations between energy consumption and number of performance events.

But first let's look at our dataset that will be provided. We know that the data will be in the following format:

Let k be the number of parameters for the energy and n be the number of records in the dataset

$E_1, x_{11}, x_{12}, \dots x_{1k}$

$E_2, x_{21}, x_{22}, \dots x_{2k}$

\dots

$E_n, x_{n1}, x_{n2}, \dots x_{nk}$

where E_n is the dynamic energy for the n th tuple and x_{nk} corresponds to the k th performance event for n th record.

We will use mathematical definition of functional relationship to prove the approaches:

Definition: Given a dataset of pairs (x_i, y_i) where $i \in [1, n]$ of two variables x and y , and the range X of x , y is a function of x iff for each $x_0 \in X$, there is exactly one value of y , say y_0 , such that (x_0, y_0) is in dataset. [6]

Prove: We need to prove that finding atleast 2 equal performance events with different dynamic energies ensures that there exists no functional relationship in the dataset.

Proof:

Let us assume that there exists a functional relation such that:

$$f(x_{n1}, x_{n2}, \dots x_{nk}) = E_n$$

where f is the functional relation for the dataset.

Our task is to find $f(x_{i1}, x_{i2}, \dots x_{ik}) = E_i$ and $f(x_{j1}, x_{j2}, \dots x_{jk}) = E_j$ where $i \neq j$ and $E_i \neq E_j$ and $(x_{i1}, x_{i2}, \dots x_{ik}) = (x_{j1}, x_{j2}, \dots x_{jk})$.

If such i and j exists. Then, we can conclude that the f is not a function by using the definition of a function as this assumed function has two images.

Which contradicts from our hypothesis stated above. Hence by proof of contradiction we could say that f is not a function on the dataset.

Restating the above we can say dataset does not contain a functional relation.

Prove: Assuming that the dataset given has linear relationship and if we are able to find the constant for any one of the events. And if it does not apply to the other tuples of data that means linear relationship doesnot exist between the dataset.

Proof:

Let us assume that the energy consumption is a linear combination of the performance events.

$$f(x_{i1}, x_{i2}, \dots x_{ik}) = E_i$$

$$f(x_{i1}, x_{i2}, \dots x_{ik}) = (\alpha_1 \times x_{i1}) + (\alpha_2 \times x_{i2}) + \dots + (\alpha_k \times x_{ik}) + \alpha_{k+1}$$

where α_i are the constants in the linear combination of performance events and $i \in [1 \dots k + 1]$

Lets find α_i where $i \in [1 \dots k + 1]$

To find this we will have to find atleast 3 records which have their parameter events equal

x_1, x_2, \dots, x_k except x_i where this x values are value belonging to a row in the dataset.

If we found three records such as:

$$E_a, x_{a1}, x_{a2}, \dots, x_{ak}$$

$$E_b, x_{b1}, x_{b2}, \dots, x_{bk}$$

$$E_c, x_{c1}, x_{c2}, \dots, x_{ck}$$

where the tuples

$(x_{a1} \dots x_{a(i-1)}, x_{a(i+1)} \dots x_{ak}), (x_{b1} \dots x_{b(i-1)}, x_{b(i+1)} \dots x_{bk})$ and $(x_{c1} \dots x_{c(i-1)}, x_{c(i+1)} \dots x_{ck})$ are equal to each other, except x_i for some $i \in [1 \dots k + 1]$ where $a, b, c \in [1, n]$ and a, b, c are not equal to each other.

Then

$$E_a - E_b$$

$$= f(x_{a1}, x_{a2}, \dots, x_{ak}) - f(x_{b1}, x_{b2}, \dots, x_{bk})$$

{ By our assumption that E is a linear combination of its parameters }

$$= ((\alpha_1 \times x_{a1}) + \dots + (\alpha_k \times x_{ak}) + \alpha_{k+1}) - ((\alpha_1 \times x_{b1}) + \dots + (\alpha_k \times x_{bk}) + \alpha_{k+1})$$

{ Gathering terms }

$$= \alpha_1 \times (x_{a1} - x_{b1}) + \dots + \alpha_k \times (x_{ak} - x_{bk}) + (\alpha_{k+1} - \alpha_{k+1})$$

{ Since we know except x_{mi} and x_{ni} all are equal }

$$= \alpha_i \times (x_{ai} - x_{bi})$$

From the above we get:

$$\alpha_i = (E_a - E_b) / (x_{ai} - x_{bi})$$

where $(x_{ai} - x_{bi}) \neq 0$ as $x_{ai} \neq x_{bi}$ by above during our finding phase.

Then we know that using the α_i and applying to result to this equation $E_a - E_c = \alpha_i \times (x_{ai} - x_{ci})$ must be true as well.

If this is false then α_i is not a constant which contradicts our assumption that our E is linear combination of its parameter is false.

Hence using proof by contradiction we can say that the dataset is not linear combination of its parameters.

Prove: Keeping our assumption that the dataset is additive. Creating Pseudo records by simple addition with other records if results into a record whose energy consumption lies in the dataset but with different performance events shows that the dataset is not linear.

Proof:

Let us assume that there exists a linear function f such that:

$$f(x_{i1} + x_{j1}, x_{i2} + x_{j2}, \dots, x_{ik} + x_{jk}) = E_i + E_j$$

$$\text{and } f(x_{i1}, x_{i2}, \dots, x_{ik}) + f(x_{j1}, x_{j2}, \dots, x_{jk}) = E_i + E_j$$

where α_i are the constants in the linear combination of performance events and $i \in [1 \dots k + 1]$

We know that if f is additive, hence new records can be generated via addition of records in the dataset.

Let V be the set of all dataset rows and dataset rows possible by combining various records in the dataset and pseudo records.

Now if we are able to find data record with $(x_{i1}, x_{i2}, \dots, x_{ik}) = (x_{j1}, x_{j2}, \dots, x_{jk})$ where $i \neq j$ and $E_i \neq E_j$, where both records belong to the set V .

Then, by the first proof, f is not an linear function. Which contradicts our assumption.

Hence using proof by contradiction we can say that the dataset does not contain a linear function.

2.3 Applications

As you can see from the proofs above, if any of the conditions above is satisfied then we are able to show that there does not exist any functional relationship between the events and the power consumption. If none of the conditions are satisfied then that shows that there might be an existence of a functional relation. However, it does not guarantee the existence of any form. Non-existence of a function and linear functions are validated. The reason for making a software like this verifies and gives the user the confidence. If data do not fit any functional hypothesis in a space, much time could be saved by preventing the unneeded search of the form of hypothesis as the software will only test the basic conditions that are not supposed to be there for a functional hypothesis.

The software is not restricted to the use of only on dataset which consists of performance events and power consumption. It is a general-purpose software which will for work for any kind of dataset in which user wants to know the existence of functional relation. The refuting of the claim of functional relation on dataset is the objective of the software.

We also know that the dataset provided is usually experimentally measured values which are not accurate. Every measuring device has some margin of error. The software will be flexible in the sense that the equality comparison of values in the approaches will always be done keeping in the margin of error provided.

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