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TDT4900 Computer Science, Master's Thesis

Optimization of Seed Selection for Information Diffusion with High Level Synthesis

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0.1 Assignment

Information diffusion is a field of network research where a message, starting at a set of seed nodes, is propagated through the edges in a graph according to a simple model. Simulations are used to measure the coverage and speed of the diffusion and are useful in modelling a variety of phenomena such as the spread of disease, memes on the Internet, viral marketing and emergency messages in disaster scenarios.

The effectiveness of a given spreading model is dependent on the initially infected nodes, or seeds. Seed selection for an optimal spread is an NP hard problem and is normally approximated by selecting high-degree nodes or using heuristic methods such as discount-degree or choosing nodes at different levels of the k-core.

High-level synthesis (HLS) is becoming an important tool in the optimization/acceleration of algorithms in hardware. Starting with an algorithm written in a high-level language such as C or C++, HLS aids with hardware design by providing a methodology and tools that guide the developer through the design process.

This project should employ HLS as a design methodology for hardware accelerated seed selection in large graphs. The student will study seed selection for a given diffusion model, write a high-level model, and use HLS to implement a hardware design that exploits parallelism in the seed selection algorithm in order to improve performance over a GPCPU implementation. –

Abstract

Information Diffusion are often used for different simulations in network research because it simulates how information propagetes thorough a network, from memes on the Internett, spreading of disease in populations, to viral marketing. Measuring spread and speed, we can find influential targets in the network, such targets are optimal targets to pass message during disaster scenario, vaccinate to prevent spreading of a disease, or even targets for viral marketing.

High Level Synthesis have in recent years matured greatly. With HLS, designing custom architectures is no longer a

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Chapter 1

Introduction

1.1 Motivation

Information diffusion is a field of network research where a message, or data, is propagated through a *network* or a *graph*. The message originates from a chosen set of nodes, known as *seed nodes*. These seed nodes passes the message to it's neighbour through the edges and thus propagate the message over theentire network. There are different models used in Information diffusion, *Independent Cascade Model*, and *Linear Threshold Model*. Information Diffusion can be used to model different phenomena such as the spread of disease, viral marketing, or even spread of viral videos and *"memes"*[2] [3]. The effectiveness of the simulation is measured in the spread and the speed of propagation. The effectiveness of the simulation is dependent on the chosen seed nodes. By finding the most optimal set of seed node, we can potentially stop an epidemic by vaccinating influential nodes, we can find important target for viral marketing by giving free sample, and use this information to quickly spread message during disaster scenarios.

There are multiple studies done regarding information diffusion, [4], [2], [5], [6]. There are few that focus on optimizing the seed selection, especially in hardware. Finding the most optimal set of seed node is useful in multiple fields. We prevent the spread of a disease by vaccinating influential nodes in the network, we can pass critical message through a population in disastrous scenario, or even find optimal target for viral marketing. The current seed selection algorithm is an greedy solution[7][DOUBLE CHECK THIS SOURCE], where every set of node is tested and the set with best coverage and time is chosen. This is a time consuming process and highly parallelizable. This makes it a good candidate for *Field-programmable gate arrays*(FPGAs).

High Level Syntesis(HLS) synthesis high level behaviour and constrains to lower level design.[8]. It allows users to implement an algorithm in high level language, C or C++, and generate an optimal design in *verilog* or *VHDL*. Verilog and VHDL are hardware descriptive language designed to describe digital

systems [9]. In recent years, High Level Synthesis have gotten more attention and more support, the xilinx forums are anwsered quickly by the developers and highly populated with seasoned hardware designers and novices.

Unlike traditional hardware design, HLS allows programmer with limited knowledge to desgn an optimal custom *Intellectual property core*(IP-core). In HLS, programmers can test out multiple different optimization schemes in short period of time. Thus allowing the programmer to quickly test out different optimization schemes.

For our implementation, we focused mainly on the ICM. The ICM is a special case of the common graph traversal algorithm *Breadth First Search*(bfs). For our implementation, we chose to implement the ICM as a custom *sparse matrix vector multiplication*(spmv). By performing ICM as spmv, we can utilize the parallelism options that spmv uncovers.

1.2 Assignment Interpretation

From the assignment text, these task were chosen as the main focus of this thesis:

Task 1 *(mandatory)* Implement Information Diffusion as Sparse matrix vector multiplication, with high level language C.

Task 2 *(mandatory)* Tailor the implementation of Information Diffusion for synthesise with Vivado HLS.

Task 3 *(optional)* Implement said design on a Zynq FPGA board.

Task 4 *(optional)* Extend the system to be able to handle graph in the size of toy graphs(containing 2^{26}) nodes)

1.3 Report Structure

We have here the basic outline for this report and a short overview of the remainder of this report:

Chapter 2: Background contains the information regarding network, Information diffusion, matrix vector multiplication and High level synthesis. Most of the background information regarding this report can be found in this chapter.

Chapter 3: Related Work shows what the related works and state of the art regarding information diffusion.

Chapter 4: Architecture

Chapter 5:.

Chapter 6: Future Work

Chapter 7: Conclusion Find something

Chapter 2

Related Work

Here, we will give you a short overview of the current state regarding Information Diffusion, High Level Synthesis and different optimization options.

- Yamans paper, where there are some works that shows the solution i use

- parallalization of the algorithm

- maybe some examples of HLS to show that HLS is used.

- showe that there are not many HLS implementation, recently matured.

- show that there are not many hardware implementation for information diffusion.

- need to look through Yamans paper and get some refrences from there.

- might be good to look at how this type of sparce matrix multiplication can be used

- show other implementation of SPmv

- Show some examples where image processing is done through vivado HLS.

- [21] A good paper showing the state of the art for HLS.

This chapter, we will look at the state of research regarding High Level Synthesis, network research regarding Information Diffusion, and Optimization of Independent cascade model and Breadth first search.

2.1 Information Diffusion

There are multiple studies done regarding Information Diffusion. One studies shows how information diffusion can be applied during an disease outbreak[2], viral marketing[20], coordinat during crisis situation[22].

Models of influence have been done on blogs[23][24], and twitter[25]. We can see that in an age of social media, the studies of information diffusion is more relevant then ever.

while other[6] have argued that the emerging of social network and media, have changed the traditional model. The activation is no longer only relying on neighbour nodes, but also an external influence. They found that large amount of information volume in Twitter is the result of network diffusion, while a small amount is due to external events and factors outside the network[6]. Another studies shows during the 2011 Egyptian Uprising, how larg amount of such a movement were "tweeted"[22].

As we mentioned in **??**, we mainly focus on 2 common information diffusion models, ICM and LTM. But there are different models too. One report[5] proposed several different problems with traditional models where each node is either *activated*(infected, influenced, '1') or *inactive*(healthy, not reached, '0'), and passes the *contagion*(information, data, infection, influence) to a neighbouring nodes through the edges. The report mentioned different assumptions that such models take. Among them is that a complete graph is provided, the spread of contagion is from a known source, and that the structure in the network is sufficient to explain the the behaviour[5]. The report propose an alternative model, *Linear Influence Model*(LIM), where the focus is on the global influence a infected node has on the rate of diffusion through the implicit network. This model takes the assumptions, that newly activated nodes is dependent on previous activated nodes. The LIM does not need explicit knowledge of the entire network, instead the model takes the newly activated nodes and model them as a *influence function*, which is used to find the global influence.

2.2   High Level Synthesis

High Level Synthesis as a concept have been around since the mid-1980s and early-1990s. Early tools, known as Carnegie-Mellon University design automation (CMU-DA)[26][27] was a pioneering early version of HLS tool. The tool gathered quickly considerable interest. A number of HLS tools were built in later year mostly for prototyping and research[28][29][30]. Some of these early tools was able to produce real chips, but the reason for lack of further development and adaptation, was that RTL synthesis was not a widely accepted and immature field. This often lead to suboptimal solutions.

In the 2000, new HLS tools was developed in academia and in the industry. These tools, used hihg level language, C and C++. Vivado HLS, designed by Xilinx [31], is one such HLS tool. The Vivado HLS became free during their 2015.4 update[32]. This resulted in an revived interest in HLS. The community around HLS is also evolving, on the Xilinx-forum, there are multiple anwsers and active members. We can see that the solution designed by HLS tools is close to traditional hand-crafted designs[33].

High Level Synthesis have getting more attention lately, more active forums and recently, Vivado HLS became free for the public.

2.3 Different optimization scheme

Chapter 3

Result

as we can see, the algorithm was able to finish a

Chapter 4

Conclusion

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