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Management of Distributed Services in MANETs



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Abstract

Nowadays mobile devices and wireless networks are becoming ubiquitous. Such a trend is leading to an emergence of new distributed services, that explore the natural interesting conditions offered by this type of networks. A prominent example is the case of multiplayer gaming.

However, granting service management functionality in these networks is a demanding task. The classical server-client and peer-to-peer approaches are not well suited to support distributed services in these environments. New approaches are appearing to cope with the demands and characteristics of mobile networks. One proposal is the zone-based architecture, where the network is divided into different zones and in each zone a zone server is selected. These special nodes are responsible for the management of the nodes that belong to their zone. PBS (Priority Based Selection) algorithm allows the selection and maintenance of the group of zone servers. This algorithm performs the selection of the zone servers based on their capacity or weight. In this thesis an algorithm called NWC (Node's Weight Computation) has been proposed to perform the weighting of the network nodes for the zone server selection. The algorithm computes the weight based on a group of node's properties and allows the adaptation of the computation as function of the service characteristics. To cope with the service context adaptation a framework has been proposed. The algorithm has been tested via simulations showing a good performance in the selection of the zone servers. The NWC algorithm has also been implemented in a real testbed.

Preface

At the end of this period, I feel that words are too simple to express feelings and experiences that I lived during this thesis time. I want to thank:

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Chapter 1

Task Description

This chapter presents the thesis task description. The first section provides a brief overview about the thesis topic, then a description of the report organization is given. In the last section the general regulations followed to complete the thesis are presented.

1.1 Introduction

MANETs (Mobile Ad hoc NETworks) are "plug-n-play" networks, consisting of a group of mobile devices (laptops, cell phones, PDAs, etc.), working in a distributed and cooperative environment, without any central administration. Each element of this self-organized network may act as a host and/or a router, forwarding packets to the other nodes.

With the emerging use of mobile wireless technologies, it is expected that MANETs and their applications will become popular soon. Real-time applications, such as games, multimedia applications, Voice-over-IP, are natural candidates to be used over MANETs [1]. But support these demanding applications under unreliable and dynamic conditions of a wireless network is a difficult task.

In [2] an overview of the technical challenges that are fundamental to support real-time applications in wireless and mobile networks is given. Service support or provisioning in MANETs asks for special functionality in order to cope with all the demands associated with this environment. The required functions can be divided in the following categories: service description, discovery, deployment and management.

Recently several solutions have appeared to support applications in infrastructureless environments, e.g. Konark [3], a middleware for service discovery and delivery in MANETs; or another proposal is SIRAMON (Service provIsioning fRAMework for self-Organized Networks) [4]. SIRAMON is a generic, decentralized service provisioning framework for self-organized networks, supporting the full life-cycle of the service by providing service specification, advertisement/lookup, deployment, management and environment monitoring mechanisms.

This thesis is integrated in the SIRAMON project, more concretely in the framework's Service Management functionality. In Appendix D, a brief presentation of the framework is given.

The management of distributed applications in MANETs presents some special issues. The traditional architectures (server-client and peer-to-peer, see section 2.1.1) do not completely cope with the dynamic conditions and limited resources of a MANET and so some alternative approaches have been proposed recently. An example is the 'Zone-Based Service Architecture' [5]. In this approach, the network is divided into separate zones. In every zone, a dedicated server node handles the client nodes belonging to the zone and synchronizes with the other zone servers. To select these server nodes a selection algorithm called PBS (Priority Based Selection) [6] is used. PBS compares the priorities of the different nodes and chooses the highly prioritized nodes as zone servers. Priority assignment is based on the node's weight, a metric which indicates the node's capability to act as a zone server.

In this thesis, an algorithm to compute and maintain the node's weight is proposed. The computation is based on the node's available resources (CPU, memory, and remaining battery power), the node's link connectivity (the quality of the links via the node can communicate with its neighbors) and the number of neighbors that a node has at each moment.

The aim of the proposed algorithm, NWC (Nodes Weight Computation), is to provide the weight computation functionality for the priority assignment of the PBS algorithm. This computation aims to perform the selection of robust and stable servers, and also provide a mechanism to adapt this computation according to the service requirements and characteristics.

To evaluate the algorithm's characteristics/performance simulation has been used. The algorithm has been implemented in a simulator (the Network Simulator NS-2 [7]) and to experiment it a generic service, based on *Real Time Multiplayer Games* characteristics, has been also simulated. Finally, the algorithm has been implemented in the SIRAMON framework.

In the next section the report organization is presented.

1.1.1 Report Organization

- Chapter 2 *Fundamentals* The Zone-based architecture and the PBS algorithm are presented in detail and an overview on related work is given.
- Chapter 3 *Node's Weight Computation Algorithm* The proposed algorithm, NWC algorithm, for node's weight computation is presented.
- Chapter 4 *Simulation and Evaluation* The NWC algorithm is tested via simulation using a generic service to test the behavior and features of the algorithm.
- Chapter 5 *Implementation* Presents the details about the implementation of the NWC algorithm in the simulator and in the Siramon framework.
- Chapter 6 *Conclusions and Outlook* Conclusions of the performed work and an outlook for future development/improvement of the proposed algorithm are given.

- Appendix A *Best Profile Computations* Presents and describes the best profile computation mechanism, to be used in the selection of the most appropriate group of servers for a given service.
- Appendix B *Finite Sate Machine (FSM)* Presents the Finite State Machine used in the PBS algorithm implementation, where the NWC algorithm is also implemented.
- Appendix C *NWC Algorithm NS-2 Installation* Presents the necessary steps for the NS-2 installation of the PBS algorithm with the NWC algorithm extension.
- Appendix D *Siramon Framework* Brief presentation of the Siramon framework.
- Appendix E *Used Abbreviations* List of the used abbreviations in this report.

1.2 General Regulations

This thesis work was guided by Károly Farkas (ETH), supervised by Bernhard Plattner (ETH) and Philippas Tsigas (Chalmers). At the end of the thesis, a written thesis report describing the work and the outcomes as well as the documentation of the implemented code had to be delivered. The master student understood and accepted the terms and regulations of ETH in regard to the developed code which would be published as open source under the terms of the GNU General Public License [8]. In the course of the work two intermediate and a final presentation had to be given. An accepted thesis report and successfully accomplished presentations were prerequisites of getting the final grade of the master thesis work.

Chapter 2

Fundamentals

This chapter gives a brief presentation of the thesis context.

In the first section an introduction to the concept of service provisioning and management in mobile ad hoc networks is given. This section also includes an overview of the basic architectures for service management in computer networks, including the zone-based architecture, which instead of a central server uses a group of zone servers distributed in the network. In the second section an algorithm for selection and maintenance of zone servers called PBS (Priority Based Selection) is presented. This algorithm classifies the nodes with a priority, based on the node's weight and uses it for the selection of the zone servers. In the last section an overview on related work is given.

2.1 Service Management in MANETs

A MANET is a self organized network composed by mobile devices, e.g. PDAs, cellular phones, laptops, see Figure 2.1. In this network there is no central administrative entity and every node is capable to perform as a router forwarding traffic to the other neighbor nodes. Nodes are free to move without "any" constraints and organize themselves without any pre-established agreement.

This type of networks is gaining some popularity nowadays mainly due to the increasing number of mobile devices and also their processing and communication capabilities, which opens doors for the creation of a new range of services and applications and explores new market areas. A very promising example is gaming. Networked games are gaining an enormous popularity [7], and it is expected that ad hoc networked games will become soon a major source of revenue in the entertainment area. Gaming is very attractive for MANETs, because it will allow to extend and explore the natural characteristics and capabilities of mobile ad hoc networks. There are a great number of scenarios where gaming may be introduced, for example, people playing a game at an airport waiting for their flight, or people playing a game when traveling on a highway.



Figure 2.1: A MANET

From the previous examples it is clear that the networks will have a high level of dynamicity and mobility with different types of devices coexisting in the same network, these conditions will increase the difficulties of deploying and manage the service.

To develop services/applications for a MANET, it is necessary to deal with the particular networking properties of these networks, which introduces a high complexity to the service itself and to the service developing task. A common approach to abstract the application layer from the network is the creation of a middleware layer capable of dealing with the lower layers and to provide an API (Application Programming Interface) to the higher layers. This allows the application developers to concentrate on the service development and abstract from the low level networking properties.

The middleware should cope with all the required network management and maintenance issues that the application services may require, providing service description, discovery, deployment and management functionality. An example of such a middleware is SIRAMON (Service provIsioning fRAMework for self-Organized Networks) [4].

SIRAMON is a framework that supports services in self-organized wireless networks. In Appendix D an overview of the SIRAMON framework is given.

In the next section, a close look at the basic architectures for network service management is given, presenting the characteristics of each one in the context of mobile networks.

2.1.1 Service Management Architectures

There are three basic architectures for service management:

- Server/Client Architecture (Centralized) There is a central administrative node that works as the network server and all the other nodes participating in the service will be its clients. The server owns the state/information of the service and also the rules to run it. It receives information from the clients, processes it, updates the new service state and (if it is necessary, for example in the case of multiplayer games) distributes the new service state to all its clients.
 - In MANETs: This architecture is not well suited for MANETs, basically due to the single central administrative point, which creates a single point of failure. It is also limited in terms of scalability.
- **Peer-to-Peer (Distributed)** In this architecture there are no central administrative authority. Every node is responsible and participates in the management and maintenance of the service state, by exchanging the service state information with all the other network nodes. The

service state is processed locally, in each node, based on the information that is exchanged between the nodes.

- In MANETs: The totally distributed design of this architecture imposes some difficulties for its usage in MANETs, mainly due to the required high overhead to perform the management of the services. The high dynamicity and also resource constraints, usually associated with these networks, makes this architecture a not very attractive solution for MANETs. There are even other disadvantages, for example in the gaming context this solution is not cheating proof because every node maintains locally the service state.
- Hybrid (Centralized/Distributed) A hybrid architecture consists of a combination of the two previously presented architectures. There is a group of servers that serve different groups of clients. In each group, the Server/Client architecture is implemented between each server and its clients. Between the servers to allow their synchronization a peerto-peer approach is implemented.
 - In MANETs: This architecture is very attractive for distributed service management in MANETs. It allows to combine the advantages of the two "basic" architectures in one: central and redundant administration points, or servers per zone, and peerto-peer communication to synchronize the servers. This reduces the complexity of the system in terms of overhead, increases fault tolerance and allows a service to be "always" accessible to the network clients.

In Table 2.1, 2.2 and 2.3 a brief summary/analysis of the most important characteristics of the three different architectures in the context of MANETs is presented.

Server/Client		
Pros Cons		
+ Secure	- Single point of failure	
	- Latency	

Table 2.1: Pros & Cons of a Server/Client Architecture in MANETs

Peer-to-Peer			
Pros	Cons		
+ Low latency	- Scalability		
+ Redundancy	- Security		

Table 2.2: Pros & Cons of a Peer-to-Peer Architecture in MANETs

Hybrid		
Pros	Cons	
+ Low latency	- Security	
+ Redundancy	- Server synchronization	
+ Scalability		

Table 2.3: Pros & Cons of a Hybrid Architecture in MANETs

2.1.2 The Zone-Based Architecture

The zone-based game architecture [5] is a hybrid architecture. It has been developed to be used in game service management, but its basic concept is also applicable in other services.

This architecture has been proposed because the traditional *server-client* and *peer-to-peer* architectures are not well suited for ad-hoc networks. There exists some similar approaches, namely Mirror Server Architecture [9], where there are special network nodes called mirrors that work as gateway points for the closest nodes. The main disadvantage of this architecture is that it does not handle fault tolerance, which is needed in a dynamic and unstable environment such as a MANET. Several other clustering techniques exist, but this thesis follows the zone-based architecture.

The basic idea behind the zone-based architecture is the concept of zone server and consequently the idea of division of the network into different zones. Some nodes are selected to perform as servers being responsible for the management of a group of nodes/clients (see Figure 2.2). The zone servers communicate between themselves to maintain the service state, which allows the service to be resistant to failures. For example, if a zone server loses connection or is shut down, its clients will continue to be served by another zone server, or a new zone server is selected.



Figure 2.2: Zone-Based Architecture - An Example Scenario

When a node "joins" the network, it will select the best zone server in its neighborhood. The joining node establishes a connection to the server and the server becomes responsible for the management of all the information concerning this client. The zone server has some special characteristics that distinguish it from the other nodes in the network.

The Zone Server

The zone-based architecture is based on the existence of special network nodes, which are designated as zone servers.

A MANET can be formed by different types of devices, with different properties, e.g. cellular phones with very tight resources, PDAs that are relatively more powerful or laptops that usually are quite powerful devices. This heterogeneity in type and capabilities (processing and communication) of the network nodes are important points in the selection of zone servers.

The aspects to be considered to perform the selection of zone servers are the following: basic capacity to run the service, energy levels, network position, communication capabilities, and some other more complicated, like the mobility characteristics of a node. Taking all these parameters into account, some type of benchmark has to be defined to decide whether or not a node is appropriate to act as a zone server. The basic aim of this thesis is to provide such a benchmark to allow the selection of the most appropriate nodes as zone servers.

To perform the selection and maintenance of the zone server nodes an algorithm called PBS (Priority Based Selection) [6] has been proposed. PBS is presented and explored in the next section.

2.2 The PBS Algorithm

The PBS algorithm treats the network as a graph and the aim of the algorithm is to select and maintain a weighted Dominating Set $(DS)^1$ of the graph. The selected DS will consist of the group of zone servers that will be responsible for the service management. For an introduction to the theory of Dominating Sets see [10].

There are several algorithms for DS selection/construction, e.g.: Largest ID [10], Local Randomized Greedy algorithm [11], Marking algorithm [12],

¹A dominating set is a subset of nodes, in a graph, such that all nodes have one neighbor in the DS or are itself in the DS.

LP-Relaxation algorithm [13], Dominator algorithm [14], Removing cycles algorithm [15], Steiner Tree algorithms [16].

All these algorithms were mainly developed to give routing functionality in ad hoc networks. Giving support for a zone-based architecture is faced with different requirements, therefore a different DS computation algorithm, called PBS (Priority Based Selection) has been proposed in [6] for the selection and maintenance of the zone servers.

The PBS algorithm performs the selection of zone servers in a MANET by assigning priorities to the network nodes, based on the nodes' properties and capabilities. The most prioritized nodes are selected as zone servers. PBS has initially been created to support real-time applications, namely multiplayer games, but it may be used in other types of services.

The algorithm grants a continuous maintenance of the DS even in dynamically changing network topologies, such as a MANET is.

In the next section, a brief presentation of the notation used in PBS and during the next chapters is given. The general ideas and definitions of this algorithm are also presented in the next section.

2.2.1 Notation and Definitions Used in PBS

A node in the network graph can be classified into four different states, depending on its priority and its neighbors' priorities:

- **DOMINATOR** Node is in the DS, will act as a zone server.
- **DOMINATEE** Node is not in the DS, but is covered by, at least, one DOMINATOR neighbor node.
- **INT_CANDIDATE** The node is an internal candidate, it wants to deploy the service, but still has not a defined state (DOMINATOR or DOMINATEE).
- **EXT_CANDIDATE** The node is an external candidate, it doesn't want to deploy the service, but even though it may be elected as DOMINATOR. This is a basic assumption in the algorithm, every node is collaborative in the deployment and management of a service.

Some other definitions used in PBS:

- **Span** The *span(v)* of a node *v* is the number of INT_CANDIDATE neighbors a node has, including itself.
- Fully connected node If a node has a link to every other node in the network it is considered fully connected.
- **Neighborslist** It is the list of all neighbors of a node containing all relevant information about a node's neighbors. The following information is stored in the neighborlist:
 - ID Unique ID of the neighbor node.
 - $\mathbf{Address}$ The network address of the neighbor node.
 - Node weight The weight of the neighbor node.
 - Span The span value of the neighbor node.
 - Status The status of the neighbor node.
 - Fullconnected Flag that indicates if a node is fully connected and a further DOMINATOR node is required.
- **Coverage** All the nodes that are directly connected to a DOMI-NATOR node are covered by this node. Every DOMINATEE node is covered by minimum one DOMINATOR node. The DOMINATOR neighbor that has the highest priority, and is closer, will be elected as the node's DOMINATOR and thus its zone server.

2.2.2 Dominating Set Computation

Figure 2.3 represents the flowchart of the basic behavior of PBS in the selection and maintenance of the DS.

The construction of the DS is based on the nodes' priorities, and the priorities of the neighbor nodes. As it can be observed in Figure 2.3, when a new "election" happens, the node sends its neighborlist to all its neighbors. After it waits for the similar message from its neighbors and defines its state based on the following set of priorities:



Figure 2.3: PBS - Flowchart

- 1. The node has a higher weight.
- 2. Tie breaker 1: the node has a higher span value.
- 3. Tie breaker 2: the node has more neighbors in DOMINATOR status.
- 4. Tie breaker 3: the node has a lower ID.

The node weight indicates the node's capability to act as zone server and the higher the weight the more powerful a node is. The span value is a measurement of how many nodes (in INT_CANDIDATE status) will be covered if the node becomes a DOMINATOR, which is a good measure if a minimal DS approximation is required. The second tie breaker is the number of DOMINATOR neighbors a node has, this decision is made because it is useful to have as less hops as possible between the DOMINATOR nodes for consistency and synchronization mechanisms required by the applications. As a final tie breaker, the node's ID is used. Every node has a different unique ID and the node with the lower ID is preferred. PBS allows the re-arrangement of the priority criteria, allowing to adapt the DS computation to different situations. For example, if the aim is to create a minimal DS, criterion 2 may be put as the first criterion for priority assignment.

The algorithm is performed in rounds. Every round consists of three steps: sending the own neighborlist to the neighbors, receiving the neighborlist from the neighbor nodes and recalculating the own status. Based on this, every node has knowledge about the surrounding (2-hop) network topology and is therefore able to determine the nodes with the highest priorities. These rounds are performed as long as there are neighbors in INT_CANDIDATE status including the node itself, within a 2-hop distance. Afterwards, a node starts a round again if it detects lost or new links.

2.2.3 Node's Weight

The first criterion for priority assignment of the PBS algorithm is the node's weight. The idea is to choose the nodes with the highest weights for the DS aiming to construct the most powerful and robust DS.

The weight should represent the capacity of a node in terms of processing, energy power and communication capabilities with the neighbors.

The proposed algorithm for node's weight computation for DS selection is presented in the next chapter. The next section presents related work with DS nodes selection, such as heuristics and algorithms used in the DS nodes selection process.

2.3 Related Work

The creation of a DS in a MANET may be a very complex task. First, the network is usually very dynamic, which requires a constant monitoring to maintain and adapt the DS to this dynamicity. Another issue is that the network is usually composed by a variety of different types of devices. These properties impose a higher difficulty on selection of the "most appropriate" nodes for the DS. In this section an overview on some DS selection algorithms is given, emphasizing the used metrics, computation procedures for the DS nodes selection and the applicability of the algorithm in the computation of the DS for different types of services.

The most common approach for DS nodes selection in context of the MANETs is to optimize the selection for a single objective. For example, minimize the power consumption to increase the network/DS life; minimize the number of DS nodes to decrease the message overhead and simplify the DS synchronization mechanism. These algorithms will be referred to as *single metric* DS selection algorithms, because the heuristic for weight assignment is based on a unique metric. There are other approaches that perform a combination of several metrics to make the DS nodes selection, these algorithms will be referred as *multiple metric DS selection*.

2.3.1 Single Metric DS Selection

Highest-Degree

The highest-degree algorithm is one of the most simple and common approaches in the construction of a DS, also known as *connectivity-based algorithm*. It was proposed in [17, 18].

Each node broadcasts its ID to its neighbors. A node x is considered a neighbor of y if it lies in the transmission range of y. The node with the maximum number of neighbors, or maximum degree, is chosen as clusterhead, and any tie is broken by the nodes' IDs, which are unique in the network. The neighbors of a clusterhead become members of that cluster, and can no longer participate in the election process.

Experiments demonstrate that the DS has a low rate of changes, but the throughput is also low. Typically, each cluster is assigned with some resources, which are shared by the members of a cluster (in a round-robin basis [17]). If the number of cluster members increases, the throughput decreases and a gradual degradation in the system's performance is observed.

This metric introduces an important question concerning the computation of Minimal DS (MDS). The use of a minimum number of nodes in the DS may introduce long delays and degrades the quality in the answers of the clusterheads, given that naturally the number of clients per DS node will increase. On the other hand if a large number of nodes is elected for the DS the throughput may be increased, but it will also lead to higher computational expenses and to the introduction of higher latencies. So the decision of a MDS approximation should be taken if the system needs

it, or is not sensitive to some of the drawbacks referred to above.

Lowest-ID

The lowest-ID, also known as *identifier-based clustering*, has been proposed in [19].

In this algorithm each node of the network is assigned with a unique ID and the nodes with the minimum ID are selected for the DS.

This algorithm does not have a very good adaptation mechanism, since a node is assigned with a given ID and remains with it during the whole network's life cycle. The solution of re-numbering the nodes' IDs is complex, given that a node has to assign some random ID, or alternatively it may be a function of the remaining battery power. But all the IDs have to be synchronized, and so neighborlists have to be exchanged, which, in case of renumbering, may generate a great amount of traffic overhead. On the other hand, if the nodes keep a constant ID, battery drainage may happen in the nodes with the lower ID, because they will always remain as clusterheads and so will have a higher battery consumption rate.

2.3.2 Multiple Metric DS Selection

Energy and Degree Aware DS Selection

In [20] an algorithm to select DS based on the node's degree and remaining battery level is proposed. The basic idea of this algorithm is the construction of the DS based on the combination of the two referred metrics to increase the network's life.

The justification for using such a combination of metrics is that the selection of the DS based on the remained energy tends to select more nodes in the DS than if the degree metric is used. It is also observed that the nodes frequently change their dominating status, thus balancing energy consumptions and prolonging network life. While there is a better balance with energy as the key metric, more nodes are also selected, and therefore the overall network energy consumption is increased. On the other hand, degree based metrics tend to reduce the size of dominating set thus reducing energy spending in each round. However, shifting roles between nodes is slow with degree based metrics, which makes higher degree nodes energy critical.

So, the idea of the algorithm is to provide an "equilibrium" combination with the two metrics to carry out the DS selection.

Several different combinations of both metrics are proposed:

- Using a balancing parameter a: $weight(u) = a \cdot degree(u) + (1 a) \cdot energy(u)$, where a is a parameter (between 0 and 1) that gives relative weights to both metrics. The aim of this approach is to increase the weight for degree up to a point where the overall consumed energy per round by all the DS nodes is better balanced than when the nodes with higher energy are selected for the DS. The best choice for the a parameter may depend on several network factors. It was experimentally shown that it was dependent on the network density, but not the number of nodes in the network. The selection of the a parameter is made under simulation and it has to be globally available for every node to perform its weight computation.
- To avoid having any parameter in the metric selected, a parameterless product and sum combinations are also proposed:
 - 1. $weight(u) = degree(u) \cdot energy(u)$, this combination is expected to balance the choice of nodes in the DS between those with high degree and high remaining energy giving importance to both equally.
 - 2. weight(u) = energy(u)/degree(u) + energy(u), gave the best results with respect to energy only metrics.

The used combinations of the metrics have shown that it increases the capacity of balance the energy consumption at the nodes, which consequently conducts to an increases of the network life.

Weight Clustering

In [21] a weighted clustering algorithm is proposed. This algorithm (Weight Clustering Algorithm - WCA) is a distributed clustering for multi-hop packet radio networks or ad-hoc networks.

Several parameters are taken into account to select the DS. The DS selection is performed on-demand, there are no periodic elections, new elections just happen if the current DS is not able to cover all the network nodes, aiming with this to reduce the computation and communication costs.

The main idea of the algorithm is to perform the weight computation according to the type of application/system requirements. In the following, the main characteristics of this algorithm are presented.

Parameters for weight computation

- *Ideal Degree* each clusterhead can ideally support only a given maximum number of nodes, to ensure an efficient medium access control (MAC) function. If a clusterhead tries to serve more nodes than it is capable of, the efficiency of the system may suffer some degradation.
- *Battery Power* the control of battery power consumption can be efficiently used in certain transmission range, meaning that it will take less power for a node to communicate with other nodes if they are within close distance.
- *Mobility* in order to avoid periodic DS changes, it is important and desirable to elect clusterheads that does not move very quickly.
- *Distance* a clusterhead is able to communicate better with its neighbors if they are in close distances. If nodes are moving away, the nodes communication may become difficult, due to the signal attenuation.

Clusterhead election

The procedure of clusterhead election consists of eight steps that should be accomplished by all the nodes. In the following, the procedure of clusterhead election of node v is presented:

- **Step 1** Find the number of neighbors that are in the node's transmission range: d_v .
- Step 2 Compute the degree difference $\Delta_v = |d_v \delta|$, where δ is the maximum number of nodes that a node can have.
- **Step 3** Compute the sum of the distances to the neighbors, D_v .
- **Step 4** Compute the running average of the speed for every node till the current time, this gives a measure of mobility denoted as M_v . (In the algorithm it is admitted that there are some mechanism of measurement of position in the network, which are not easily feasible in a MANET).
- **Step 5** Compute the cumulative time, P_v , that the node has been clusterhead, measuring the time a node has been consuming more battery than the ordinary network nodes.
- **Step 6** Calculate the combined weight, W_v , of the node v:

$$W_v = w_1 \cdot \Delta_v + w_2 \cdot D_v + w_3 \cdot M_v + w_4 \cdot P_v \tag{2.1}$$

Where w_i , i = 1...4, are weight factors, dependent on the type of service that is being deployed in the network.

- Step 7 Every node chooses the neighbor with the smallest weight as its clusterhead. Neighbor nodes that have chosen a clusterhead are not allowed to still participate in the election procedure.
- Step 8 Steps 2-7 are repeated until not all the network nodes are assigned with a clusterhead.

2.3.3 Summary

From the performed study it could be observed that in general the use of a single metric for the selection of the DS nodes may generate some constraints, mainly if it is aimed to perform the selection of DS with different purposes, or for different types of services.

When multiple metrics are used in the DS selection, an increase is observed on the accuracy and adaptability of the selection of the DS, according to the system needs.

The WCA algorithm is, from the literature exploration made in this thesis, the algorithm that uses the most representative group of metrics for node's weight computation, allowing the adaptation of the weight computation to the service/system requirements/characteristics.

The algorithm proposed in this thesis has some similarities to WCA algorithm, given that the idea is to create a procedure for node's weight computation that can be used in the selection of DS in different systems and services.
Chapter 3

Node's Weight Computation Algorithm

In this chapter the algorithm for node's weight computation (NWC - Node's Weight Computation), developed during this thesis work is presented. In the first section, the objectives of the designed algorithm are stated. Then, the assumptions followed to derive the algorithm are described in Section 3.2. The NWC algorithm design issues are explained in Section 3.3, and finally in Section 3.4 the algorithm is presented.

3.1 Algorithm Overview

3.1.1 Goals

The main goal of the algorithm is to provide a systematic procedure for node's weight computation to create a robust and powerful Dominating Set (DS) for different types of services in a MANET.

The majority of the existing algorithms for DS selection in MANETs use a simplified heuristic to classify/weight the network's nodes. Generally just one or very few node's properties are taken into account in the DS computation, aiming to find the best combination of nodes to achieve a specific objective, e.g.: Minimum Dominating Set (MDS) approximation, minimize the energy consumption.

When it is intended to perform the selection of a DS with different objectives, other properties should be taken into account, as well. For example, to select the most appropriate group of zone servers to serve a multiplayer game where, usually, good links and powerful nodes are required, other properties should be taken into account in the DS nodes selection, otherwise the selected nodes may not be able to serve the DS clients with good performance.

In the literature research, some algorithms can be found using a multiple metric heuristic to perform the DS selection (see section 2.3.2) but even these approaches are far too simple.

Different services have different requirements. To allow the NWC algorithm being used in construction of the DS for different types of services, a representative number of node's properties are used to accomplish the node's weight computation. Each type of service will select the most important properties for its DS selection. These are the basic propositions that conducted the development of the NWC algorithm.

The two main groups of parameters used in the algorithm computation are:

- Node's processing/energy capabilities
- Node's communication capabilities

The Node's processing/energy capabilities represent the local capabilities of a node in terms of processing power (CPU and memory) and battery power. The Node's communication capabilities represent the number of links a node has (node's degree) and the link quality of each link to the node's neighbors. The use of these two groups of parameters allows to measure the most important characteristics for a node's representation in the context of the zone server selection.

The selection of the "powerful" nodes into the DS is not simply a function of the referred parameters, it is also dependent on other factors, such as the type of service being deployed and the type of network. To address this context variation the referred parameters are affected by an importance, a weight factor, which reflects the importance of each single parameter in the context of a given election.

3.2 Assumptions

To design the NWC algorithm some assumptions were taken, basically related to the type of environment (MANETs), related to the framework where it will be implemented (Siramon [4]) and inherently related to the PBS algorithm where the NWC algorithm is used for the node's weight computation. The assumptions are divided in three basic groups: node, network and service context assumptions. The first two groups are concerning to the type of network, where the algorithm is used and also the type of devices in the network. The third group presents some assumptions related to the services that are deployed in the network, and requirements for the adaptation of the weight computation function to the service requirements.

Node:

- Heterogeneity The network is formed by different types of devices (laptops, PDAs, mobile phones, etc). It is assumed that all the devices can run the Siramon framework to grant all the required service provisioning functionality.
- Local parameter values It is assumed that every node is able to extract/collect all the necessary information of each of the local parameters (CPU, memory and battery) to perform the necessary computations to run the NWC algorithm.
- Link quality information It is assumed that the nodes can extract from their network interface devices the link quality information, measuring the SNR, of the links to each of their neighbors. This is necessary for the link quality parameter computation.
- **Neighbor knowledge** Every node keeps information about the neighbors, that are in its transmission range, in a neighborslist where all the relevant information about the neighbors is stored.

Network:

No central administration - The network has no unique central administrative entity, the management of the services is done in

a distributed manner by the zone servers. All the network and service management issues (service description, discovery, deployment and management (where the NWC algorithm is included)), is handled by the Siramon framework.

Mobility - The network elements are mobile devices and so they can move "randomly" without any constraints, entering and leaving the network without warning the other network elements. In the master thesis [22], a mobility prediction extension has been developed for PBS. It provides an estimation of link stability, thus links that are showing unstable behavior should not be considered in the DS selection.

Service Context:

Service Profile - A property of the algorithm is the adaptation of the weight computation to different types of services. This is achieved by assigning different importance levels to the different parameters, which is called service profile. We assume that the service description document, required for the service deployment in Siramon (see Appendix D) has an entry with the service profile, which is used for the weight computation.

3.3 NWC Algorithm Design

The design of the NWC algorithm follows a very simple approach:

Take a representative group of parameters into account and use them to compute the node's weight function of the service and network scenario.

3.3.1 Representativness of a Node

As discussed in section 3.1, the representation of a node's capabilities is achieved by using a group of parameters that reflect the characteristics a zone server needs to have for a given type of service. These characteristics are: the node's processing/power capabilities and the node's communication capabilities.

These characteristics are represented by a group of five parameters:

Processing Power:

- **CPU** Represents the available CPU capacity of a node, that should be enough to act as a zone server. This parameter is used because, for some services, it is aimed to have powerful nodes in the DS, and so nodes with enough resources should be preferred to act as zone servers.
- Memory The memory parameter, represents the memory capacity of a node. If it is important to have fast zone servers for a given service this parameter may play an important role, e.g., in real time services.

Energy:

Battery power - Represents the energy level of the node. This value will allow to look for the nodes that have a long life time, if such a requirement is important for the DS selection.

Connectivity:

- **Degree** The degree represents the number of neighbors a node has. This parameter may allow to look for the best MDS approximation, if it is important in the DS selection for a given service.
- Link quality The link quality parameter represents the quality of a connection between a node and its neighbors. This parameter allows to look for the nodes that have the best links to the neighbors, an important requirement for some types of services.

The presented parameters summarize the basic characteristics of the network nodes and their connections with the neighbors. Different services can have different demands for each of the presented parameters, but having such groups of parameters will allow the algorithm to be used in the deployment of different types of services in different contexts.



3.3.2 Algorithm Structure

Figure 3.1: NWC Algorithm - Parameters

The algorithm has five basic parameters (see figure 3.1): CPU, memory, battery, node degree and link quality. The other factor that influences the weight computation is the system's (Service/Network) context.

The formula for weight computation follows a linear combination of the referred parameters. Each parameter contributes with a given partial weight to the total node's weight according to the importance of the parameter in the DS selection.

The formula for node's weight computation is the following:

$$weight = \frac{cpu + mem + bat + deg + link}{number_parameters}$$
(3.1)

Where:

$$cpu = importance(cpu) \times repr(cpu)$$
(3.2)

$$mem = importance(mem) \times repr(mem)$$
(3.3)

$$bat = importance(bat) \times repr(bat)$$
(3.4)

$$deg = importance(deg) \times repr(deg)$$
(3.5)

$$link = importance(link) \times repr(link)$$
(3.6)

The first element (*importance*(*parameter*)) of the parameter's value computation represents the importance that is assigned to the parameter. This is dependent on the service/network context where the algorithm is applied (from now this value is referred to as parameter's importance). The second element (repr(parameter)) of the multiplication consists of the computed representation/value of the given parameter of the system based on measurements.

Parameters computation

In this section, the formulas to compute the parameter values are presented. A unitary scale for parameter value normalization and for the importance values has been used. The node's weight is also mapped in a unitary scale, see Formula 3.13.



Figure 3.2: NWC Value Scales

As can be seen in Figure 3.2, for parameter values a continuous unitary scale is used and for the importance values a discrete scale of values is used, to differentiate the level of importance of the parameter in the context of the service.

The formulas to compute the parameter values are the following:

$$repr(cpu) = \frac{capacity(cpu) - load(cpu)}{capacity(cpu)}$$
(3.7)

$$repr(mem) = \frac{capacity(mem) - load(mem)}{capacity(mem)}$$
(3.8)

$$repr(bat) = \frac{sys(bat)}{capacity(bat)}$$
(3.9)

$$repr(deg) = 1 - \frac{1}{node(deg)}$$
(3.10)

$$repr(link) = \frac{sys(link)}{max(link)}$$
(3.11)

CPU and **memory** parameter representations are inversely proportional to their load on the node. This approach allows to select the nodes that are not so loaded for the DS, which will naturally select the most powerful nodes, given that they are usually not so loaded.

If the CPU/memory load is very high, and they are assigned with some importance, an anomaly can be generated and a new election started to reflect the incapability of the nodes to act as zone server. This question is discussed further in Section 3.4 where the algorithm computation and maintenance procedures are defined.

The **battery** parameter is directly proportional to the battery level on the node. With this approach there is no differentiation between different types of device batteries (e.g.: PDA, and laptop), because usually the "less powerful" devices have longer battery life. However if the value for this parameter was computed directly based on the node's battery remaining time, this would favor the selection of weak nodes into the DS, which may be undesirable even if the aim is to increase the network's life.

If the energy level of the battery is too low, an anomaly may also be generated removing the node from the DS and generating a new election.

The **degree** parameter representation is directly proportional to the number of neighbors that a node has. This means that the higher the number of neighbors a node has, the higher the probability is to cover more clients, thus the number of DS nodes can be decreased. In other algorithms [21], a maximum number of clients is defined, but in this thesis such a limit is applied.

The **link quality** parameter is directly proportional to the measured link quality at every link to each neighbor obtained from the node's network interface device. An average of the observed link quality at each link is made and used to measure the node's link quality to its neighborhood.

After computing the parameter values for each node it is necessary to have the service profile (combination of parameters' importance values) to compute the node's weight. In the next section this profile is discussed.

3.3.3 Service Profile

The representation of the node's context and service requirements are reflected in the weight computation by the parameters' importance value. The aim of this importance value is to allow an adaptive selection of zone servers, which is function of the type of service and the network where it is being deployed and the objectives defined for the DS.

Given the list of parameters presented in Section 3.3.1, the problem to solve is: how should the parameters' importance value be defined?

The answer to this question is not simple. The definition of a dynamic functionality, responsible for this computation, would be theoretically the most appropriate approach. With such a functionality a given service would adjust the parameters' importance value as function of the service requirements and context where the service is being deployed. This would be "optimal", given that even knowing the type of service characteristics, some factors, like the scenario conditions and the number of nodes in the network, will affect the context of the service, meaning that the same service may have different requirements in different contexts.

But on the other side, such a dynamic approach would increase the complexity of the system given that higher complexity would be required in the system to grant a total synchronization of the importance values in all nodes.

A MANET is naturally a distributed and decentralized system and performing a dynamic profile adaptation would require a distributed computation, which would increase the message overhead and the complexity of the system. The increase of complexity may compromise this approach, due to the limitations/conditions of a MANET. In this thesis another approach is used to address this problem. The approach taken consists of the definition of static service profiles, thus each service will assign the most appropriate parameters' importance value for the service. This profile is used by all the nodes for the node's weight computation. Such approach allows a "correct" election process without increasing the system's complexity.

With a static approach it is not possible to completely predict the "context" where a given service will be deployed. This means that even knowing the basic characteristics of the service an optimal profile is not granted for the zone server selection.

To define the parameters' importance values, a framework has been created. It is based on simulation, and is presented in the next chapter.

The definition of the service profile can also be set directly according to the defined objective for the service DS. For example: if the aim is to get the best MDS approximation, the degree parameter's importance should be set to "1" (*high importance*) and all the other parameters' importance to "0" (*no importance*).

3.4 The NWC Algorithm

In this section, the steps of the NWC algorithm are presented.

When an election happens, the following steps are taken for the node x weight computation:

- **Step 1** Get the list of neighbors (neighborslist), that are in the node's transmission range.
- **Step 2** Compute the node's degree parameter value, degree(x), which consists of the number of nodes x has in its neighborslist.
- **Step 3** Get the link quality statistics to every neighbor node and compute the node's link quality parameter value, link(x):

$$link(x) = \frac{\sum_{i=1}^{N} link(i)}{N}$$
(3.12)

where N represents the number of neighbors (or degree).

- **Step 4** Get the CPU and memory load, battery energy level, and based on the equations 3.7, 3.8 and 3.9, compute the local parameter values: cpu(x), mem(x) and bat(x).
- **Step 5** Get the parameters' importance for the service, *i*(*parameter*).

Step 6 - Compute the node's weight:

 $weight(x) = \frac{i(cpu)cpu(x) + i(mem)mem(x) + i(bat)bat(x) + i(link)link(x) + i(deg)deg(x)}{num_param}$ (3.13)

The *num_param* represents the number of parameters that are not assigned with *not important* importance value.

NOTE: The computed node weight is a value that lies in the continuous scale of values between 0 and 1. A 0 weighted node means that the node does not have enough resources, and/or good links to the neighbors. On the other hand if a weight of 1 is obtained, the node is well suited to act as a zone server, given that it has enough free resources, and/or good links to the neighbors.

After the election, the node sends its weight and other information to and waits for the same information from all its neighbors. When these information are received, the node checks the neighbors' weights and decides its status (see section 2.2). At this point the election is finished. The steps taken to perform the election process are called "election rounds".

When an election round is finished the node enters a state where it will remain until a new election is started locally or by some neighbor.

When the node is in this state and its status is DOMINATOR, its resources are monitored to grant that they are in the limits of "good performance". Such monitor runs periodically and checks the three local parameters (CPU, memory and battery). If the values are not in the range of admittable values, an anomaly is generated. The default values used as basic demands for the local parameters are given in Table 3.1. If the values are not accomplished, the node will start a new election round.

The definition of maximum/minimum values for the parameters will grant that the DS nodes are always in good condition to perform their tasks. But, if for a given type of service a local parameter is assigned with *not important* importance level the parameter is not checked/monitored.

NWC Local Parameters Basic Demands			
Parameter	Demand		
CPU	load below 90%		
memory	load below 90%		
battery	energy level above 10%		

Table 3.1: NWC Local Parameter Default Demands

3.5 Summary

In this chapter an algorithm for node's weight computation called NWC has been proposed. The basic idea of the algorithm is to provide a node's weighting procedure, adaptable to the type of service that is being deployed in the network, giving a general and adaptable procedure to be used in the selection of DS nodes for different types of services.

To provide such a general procedure, a representative group of parameters is taken into account to measure the node's capabilities. The parameters are divided in two groups:

- *Processing/Power parameters* representing the local capabilities of the node: CPU, memory and battery.
- *Communication parameters* representing the number and quality of the node's connections to its neighbors: degree and link quality.

A linear combination of these five parameters is used to compute the node's weight. To adapt the weight computation to the type of service that is being

deployed, differentiating the importance of each parameter in the weight computation, each parameter is assigned with an importance value related to three preset levels.

To compute the most appropriate profile of importance values for a given service a framework has been constructed. It allows the simulation of a type of service, and the study of the most relevant parameters in some given scenario. The framework is presented and used in the next chapter.

The service profile may also be defined directly by the system/service designer according to the requirements of the service and the objectives for the DS computation.

The working of the algorithm can be divided into six basic steps (see section 3.4) that allows the computation of the node's weight. This value is assigned to the node and transmitted to the neighbor nodes, using the PBS algorithm. According to the assigned weight and the neighbors' weights a node will set its status and the DS will be constructed.

Chapter 4

Simulation and Evaluation

In this chapter the algorithm proposed in this thesis is evaluated using a simulation technique.

In the first section the goals of the simulation are stated and the used framework for the study of the best profile for a given type of service is presented. To evaluate the algorithm under simulation a test service is used, which is presented in section 4.2. A *Factorial Design* technique is used to conduct and evaluate the results of the simulation, this is presented in section 4.4. In section 4.5, the simulation settings are shown and in the last section a study and evaluation of the simulation results are made.

4.1 Simulation Goals

The basic idea of the simulation is to perform a study of the proposed algorithm performance, proving that the NWC algorithm allows the selection of a robust and stable DS.

The proposed algorithm is adaptable to the type of service and context where it is being deployed, by performing an adaptation of the parameters' importance according to the service needs. This flexibility allows the algorithm to be used in different systems for the deployment of different services.

Theoretically, a dynamic adjustment of the parameter importance values (profile), according to the context of the service/system, would conduct to

more optimized DS selection. This happens because even for the same type of service the system's context and requirements may vary. The drawback of such a dynamic profile adjustment would be the increase of the traffic overhead and more complex mechanisms to cope with the nodes synchronization, given that every node needs to have the same parameter importance profile for weight computation.

The NWC algorithm is based on a static assignment of parameter importance values, according to the service requirements and context of system. This approach is also followed in some other similar algorithms (see [21]) where different parameters are weighted according to the system's properties.

To study the best profile for a service an experimental design called *Factorial Design* and a *Multi-Objective Optimization* function of the objectives defined for the DS selection are used. This process is defined in the NWC framework. To test and evaluate the proposed algorithm, a real time multiplayer game service test has been used.

The simulations are performed in the Network Simulator NS-2 [7].

4.1.1 Simulation Constraints

There are three basic techniques to evaluate an algorithm performance: analytical modeling, simulation and implementation and measurement in a real system, [23]. To evaluate the NWC algorithm a simulation technique has been used.

Simulation has several advantages when compared with real test scenario measurements' techniques. For example, the number of service participants; the same scenario and traffic conditions, which in this case is important to compare the different importance values combinations in the same simulation conditions. Such conditions are difficult to achieve in a real testbed.

The problem when a simulation evaluation technique is used resides in the difficulty of truly representing a service/system. In the present case, for example, the service's characteristics (processing requirements, traffic and

users behavior) are difficult to model in a simulator. Some abstraction of the real system is required to allow its representation in simulation.

In the next section, the type of service used as test service in the simulation is explored and in section 4.3 its specifications are defined.

Multiplayer Games

4.2 Real Time Multiplayer Games

Figure 4.1: Multiplayer Games

Multiplayer games are classified in two categories, turn based and real time games.

Real time games are more demanding than turn based games because players often have a lot more information to transmit between each other or to the server than in a turn based game. For example, in a multiplayer game of chess the only data that needs to be transmitted may just be the current player's turn and the square the current player has moved to. Since this information does not need to be periodically updated, it is transmitted only after each player's turn. On the other side, real time games require higher amount of traffic to keep the game states synchronized and low delays to deliver all the messages. For example, in a car racing game the position of each car would have to be updated frequently in order to give the cars a smooth motion. The main problem with real time games that makes them difficult to implement, when compared with turn based games, is that the data need to be transmitted repeatedly at very short intervals and strict demands have to be granted to avoid interference in the users' perception of the game.

Multiplayer games are becoming popular in MANETs [1]. As has been referred before, PBS algorithm is a general algorithm to support the creation and maintenance of DS for different types of services in MANETs but was specially designed to support real time services. To test the NWC algorithm and see how it works in conjunction with PBS a real time multiplayer game is used as test service in the simulations.

In figure 4.1 the three basic types of real time multiplayer games are presented:

- First Person Shooter (FPS)
- Real Time Strategy (RTS)
- Sports

In this study focus on the two first type of services (FPS and RTS) is given, mainly because there are already several studies and models about their characteristics.

During the simulation it is assumed that all the devices participating in a game session are capable of deploying the service, thus every node has enough processing power resources (CPU and memory) to run the service.

Two types of service's properties are studied, networking and processing characteristics. In the next section the networking/traffic characteristics are studied, the processing characteristics will be explained in section 4.3, where the test service properties are defined.

4.2.1 Networking Properties

In the analysis of the networking properties of a real time multiplayer game, the following characteristics are the most important:

- Latency also referred as end-to-end delay, is the time taken for a packet to travel from its source to its final destination, between client and zone server and vice versa.
- Jitter variation of the latency in a given link.
- **Packet Loss** number of packets that are lost during a transmission between two network points, between client and zone server and vice versa.

The admissible latency for most of the real time multiplayer games is in the range of 100-150 ms. Usually FPS games are more sensitive to this factor than the RTS games, due to the fact that the player has direct control over his playing unit. Therefore any type of delays is a direct interference on his perception. On the other hand RTS games do not have such strict delay demands, because the player controls its units in an indirect way, by giving orders, which implies a higher insensitivity level of the player to higher delay values [24].

The variation of the latency, or jitter, affects the players' perception of the game. This variable is dependent on the type of game but also on the player, because playing reaction is different for each user. It is a fact that high level of jitter always degrades the game experience of the player so it should be kept as low as possible.

The packet loss in real time multiplayer games may be quite important and low levels should always be achieved to assure playability to the players. In [25], it has been proved that a level of 3% to 5% of packet loss is admissible in a FPS game.

4.2.2 Requirements

The general requirements in terms of latency and packet loss for real-time multiplayer games are summarized in table 4.1. From the performed study, the effect of jitter in the users' perception of the games is still not well defined.

	latency	packet loss
FPS	$50\text{-}150 \mathrm{\ ms}$	3-5%
RTS	$300\text{-}500~\mathrm{ms}$	3-5%

Table 4.1: Networking Requirements for Real Time Multiplayer Games

4.3 Game/Test Service Specifications

To perform the evaluation of the proposed algorithm a pseudo service has been created.

Performing service's behavior and requirements in a simulation is not an easy task so some simplifications have been made. The main characteristics of the service are:

Processing load generation - The service is characterized by a given processing activity. This is associated with the game playing activity but also in case of the zone servers with the serving tasks (processing clients' requests, performing state synchronization, distribute game state). So the notion of CPU and memory and the load generation as function of the service characteristics must both exist. For the performed simulations the following scale of values for CPU and memory capacity, for the network's nodes, has been considered:

CPU	500 - 1000 (MHz)
memory	250 - 500 (Mb)

Table 4.2: Processing Capacities Scale

It is admitted, based on the generated CPU and memory load, that some nodes may not be able in some moments to run as zone servers due to constraints on their processing resources.

Energy consumption - A simple mechanism of energy consumption is also considered in the simulator implementation. It is taken into account the required energy to send/receive packets. **Traffic generation** - A given service is characterized for some traffic pattern. In our case, considering the zone-based architecture, the generated traffic is between the clients and zone servers, and vice versa. To model the traffic patterns between clients and servers a constant bit rate (CBR) source is used.

Service Properties			
CPU	500 MHz		
memory	$250 { m ~Mb}$		
battery	f(traffic)		
traffic (server $\rightarrow client$)	$10 \mathrm{KBit/s} \ (\mathrm{CBR})$		
traffic (client $\rightarrow server$)	10 KBit/s (CBR)		

The service's properties are summarized in table 4.3.

 Table 4.3: Service Properties

Given the type of service, a real time service, the defined objectives for the DS are to assure a stable and powerful DS capable of serving all the network clients in good conditions. So in the definitions of the objectives for the *best profile* computation these objectives will have high priority.

4.4 Factorial Design

Factorial design is an experimental design technique specially useful to measure the effects of a group of factors in the output of an experiment. The more common and complete analysis is called *Full Factorial Design*, where a complete combination of all the factors that may influence the output is made and based on the results of the experiment it is possible to conclude about the importance/contribution of the factors in the output of the system. More details about this evaluation technique can be found in [23].

In the NWC algorithm, five factors may influence the output of the system, and three levels of variation have been chosen for it. Details about the factors and their levels are discussed in 3.3. Table 4.4 presents the possible

Weight Factors					
cpu memory battery link quality degree					
$0\ ;\ 0.5\ ;\ 1$	0 ; 0.5 ; 1	$0\ ;\ 0.5\ ;\ 1$	$0\ ;\ 0.5\ ;\ 1$	$0\;;0.5\;;1$	

Table 4.4: Parameters' Weight Factors

values for each factor, or parameter importance, (0 - not important; 0.5 - normal importance; 1 - high importance).

If a full factorial design is used $3^5 = 243$ different simulation runs would be required, not counting with the required simulation replication that is necessary to increase the confidence of the measurements in the simulation. Such an approach is not very practical nor feasible, due to the long simulation time. Therefore some simplifications are made to perform the experimental design.

Simplifications are based on the system's properties and allow to reduce the experimental design to a *Fractional Factorial Design* [23]. This simplification is very often used when the system complexity is too high to be evaluated with a full factorial design. In the next section this process is presented and explained.

4.4.1 Fractional Factorial Design

The basic idea of the fractional factorial design is the same as the full factorial design but, in this case, there is a simplification on the number of factors and/or the number of factor levels in the experiment.

In this concrete case, a test service representing a real-time multiplayer game is simulated in a MANET to study the most important factors in the selection of the DS and, based on this, evaluate the NWC influence on this selection. The aimed DS for the service, as referred in section 4.3, should be a stable and powerful DS, capable of serving all the network clients in good conditions, which means that DS should not experiment too many changes and should be made by powerful devices (processing power and good links to the neighbors). Based on the previous assumption, the following simplifications are performed to the full factorial design:

CPU - As it has been defined, the service aims at representing a *real time multiplayer game*. By analyzing the service properties (table 4.3), it can be concluded that it will require 500 MHz cpu processing power, which in the simulation is equivalent, in terms of cpu power, to the weakest node in the network (see table 4.2).

Based on this assumptions, it is clear that when selecting nodes for the DS the processing capacity should be taken into account, otherwise weak nodes may be selected for the DS; this may be a cause of problems (which in the algorithm are called *anomalies*), because if a node gets *overloaded* it will not serve properly its clients and this will not cope with the real-time demands of the service.

Hence, to avoid such problems in the node's weight computation, the nodes' processing power should be taken into account when performing the weight computation for the DS selection. To achieve this, two weight factor values are used for the factorial design, *Normal Importance* and *High Importance*.

- memory The same arguments used for the cpu parameter are valid here. So, this parameter is also assigned with two levels of importance, Normal Importance and High Importance.
- **battery** This is a very important factor for several types of services, mainly when it is aimed at keeping a service running for large amount of time without generating oscillations in the DS nodes. The nodes with high levels of battery energy should be preferred for the DS. In the case of a real-time multiplayer game this factor should not play a very important role. This type of services is high demanding in other aspects, requiring well located and powerful nodes to be selected for the DS to cope with such conditions.

To achieve such selection this parameter will be neglected when performing the selection of a DS to serve this service, meaning that *No Importance* level is always assigned.

- link quality As referred in section 4.2, real-time multiplayer games require good links between the servers and their clients due to the sensitivity that they have to packet loss, latency and jitter. Such conditions impose the selection of nodes with good link quality to their neighbors. To achieve such selection the link quality parameter will be assigned with normal importance and high importance levels when performing the factorial design. This will help the selection of nodes with good
 - link quality to their neighbors.
- degree The selection of a small DS has some advantages, mainly because this will decrease the necessary traffic overhead and also the delay required for synchronization of the DS nodes. But such MDS (Minimal Dominating Set) approximation can also cause some problems if nodes with bad connections, or without enough processing power, are selected for the DS.

It is clear that this parameter should be important in the DS nodes selection, so in the factorial design it will be assigned with *normal importance* and *high importance* levels to study which one will conduct to the best results.

Table 4.5 summarizes the weight factor combinations, or profiles, that are taken into account for the factorial design experience to select the best combination, or best profile, for the used test service. It also shows an evaluation of the algorithm performance.

4.5 The NWC Framework

In Figure 4.2, the NWC framework for computation of the best profile of parameter importance values, for a given service, is presented.

There are three basic variables that influence the simulation and more concretely the selected DS:

• The service specifications, which defines the characteristics of the simulated service (nodes' processing power load and traffic generation service patterns).

Weight Factors Combinations					
Combination	cpu	memory	battery	link quality	degree
1	0.5	0.5	0	0.5	0.5
2	0.5	0.5	0	0.5	1
3	0.5	0.5	0	1	0.5
4	0.5	0.5	0	1	1
5	0.5	1	0	0.5	0.5
6	0.5	1	0	0.5	1
7	0.5	1	0	1	0.5
8	0.5	1	0	1	1
9	1	0.5	0	0.5	0.5
10	1	0.5	0	0.5	1
11	1	0.5	0	1	0.5
12	1	0.5	0	1	1
13	1	1	0	0.5	0.5
14	1	1	0	0.5	1
15	1	1	0	1	0.5
16	1	1	0	1	1

Table 4.5: Weight Factor Combinations, for Factorial Design

- The Parameters' profile combinations, the different combinations of parameter importance values used to perform the node's weight computation (table 4.5).
- The scenario characteristics, nodes (number of nodes, movement and velocity and associated processing power and battery) and the scenario topology (dimension of the scenario).

In the simulation, the PBS algorithm is used to make the selection and maintenance of the DS. PBS is provided with node's weight computation and service monitoring functionality.

The implementation of the NWC algorithm was made as an extension of the PBS algorithm, so the NS-2 implementation of PBS developed in [6] was used as the basic algorithm for selection and maintenance of the DS,



Figure 4.2: Simulation Structure

with the NWC algorithm to *weight* the nodes. The same happened with the service monitoring functionality. In chapter 5 the details about the NS-2 implementation are presented. In the next section the most important implemented NWC functionality, in the simulator, are briefly introduced.

To evaluate the simulation, a group of five metrics has been used: Number of DS nodes, Number of DS changes, Number of Anomalies, Average packet loss and Average latency. This metrics are explained in section 4.5.4.

The last step of the simulation experiment consists of the analysis of the observed metrics for the different scenarios, with the different profile combinations. The aim is at finding the most appropriate group of parameter importance values, or best profile, based on the objectives defined for the DS selection, for the test service. To perform this evaluation a multi-objective evaluation technique has been used.

To perform the simulation the Network Simulator, NS-2 [7], with the wireless extensions developed at CMU [26] has been used.

4.5.1 NWC Simulation

To evaluate the performance of NWC algorithm in the DS nodes' selection and study how it performs in conjunction with the PBS algorithm the following extensions have been made to the base implementation of the PBS algorithm, in NS-2:

- **Node** The node is provided with three new extensions: *CPU*, *memory* and *battery*. These three values are used to generate, according to the definition of node's processing/energy capacity, service processing load and traffic, the parameters' representation values for the node's weight computation. So at the moment of the node's creation, all these values are *attached* to its properties.
- **Link quality measurement** Periodically each node performs measurements of the link quality, measuring the SNR, with its neighbors.
- Service requirements Defines the generated CPU/memory load and battery consumption during the simulation. The traffic generation properties are also defined, more concretely the traffic rate between the Zone Servers and its clients and vice versa.
- **Node's weight computation** Implements the NWC algorithm which allows to compute the node's weight each time a node is participating in an election.
- Service Monitoring During the simulation run, DS nodes are periodically checked to see if their local parameters are not out of the admissible limit values, cpu and memory overloaded or battery energy level very low.
- A full description of the NS-2 implementation is given in section 5.1.

4.5.2 Settings

To perform the simulation, each mobile device shares a 2Mbps radio channel with its neighbors, using the IEEE 802.11 MAC protocol with a two-ray ground [27] propagation model.

The mobility model used was the Random Way Point (RWP) model. This mobility model is often used to model movement of nodes in mobile networks and its performance are quite reasonable when comparing with other mobility models [28].

The basic dynamics of this model is: a node selects randomly a target location in the simulation scenario and moves to that point using an uniformly distributed speed. When it reaches the destination, it remains stopped for some instants (also a random value) and again moves to another point in the scenario. For the proposed scenarios, this model is reasonable but if a more particular simulation is aimed at, for example modeling the mobility of nodes inside a building, other mobility model should be selected. Some mobility models for MANETs are proposed in [29].

Two different scenarios have been used during the simulations to allow the evaluation of the algorithm in different conditions. The first is called *School Yard Scenario* and aims at describing a "real" gaming situation; a second scenario called *Test Scenario* is used to test the algorithm in a very demanding situation.

School Yard Scenario

This scenario consists of a group of people that stand in a school yard $500x500 \ m^2$, in line of sight, playing a game with each other. In such conditions it is assumed that nodes have a transmission range of 250 meters, a typical value for a WLAN in an open air area, free of obstacles. The movements of the nodes are modeled by the Random Way Point (RWP) mobility model, meaning that the nodes are moving freely in the yard. The velocity of their moves is uniformly distributed between 0 and 3 m/s, given that the idea is to model a real game session and so the players should not

move, or in case of movement it should be very few and smooth, but it is supposed that they can move during the game.

Table 4.6 summarizes the simulation settings for the school yard scenario.

School Yard Scenario				
Number of nodes	25 nodes			
Simulation time	600 seconds			
Mobility model	Random Way Point (RWP)			
Dimensions	$500 \mathrm{x} 500 \ m^2$			
Speed	0-3 m/s			
Pause time	180 seconds			
Traffic	CBR traffic (see section 4.3)			

Table 4.6: School Yard Scenario - simulation settings

Test Scenario

The test scenario is used as a limit/straining test for the performance of the algorithm, testing how the selection of nodes (based on their weights) influences the DS selection and stability under very demanding conditions. This scenario is characterized for its large dimensions, high number of nodes and also the speed of the nodes' movements. The movements are also modeled by the RWP model and the speeds are uniformly distributed between 0-10 m/s. Such conditions will obviously impose higher number of changes in the DS and also increase the number of required nodes in the DS.

Table 4.7 summarizes the simulation settings for the test scenario.

4.5.3 The Factorial Design

As has already been referred, the method used to test the algorithm has been an experimental design technique called *factorial design*. The design for the used test service has been made in section 4.4.1. The experimental design has been performed in both simulation scenarios to test the validity of the algorithm, and compare the results.

Test Scenario				
Number of nodes	35 nodes			
Simulation time	600 seconds			
Mobility model	Random Way Point (RWP)			
Dimensions	$800 \ge 800 \ge m^2$			
Speed	0-10 m/s			
Pause time	100 seconds			
Traffic	CBR traffic (see section 4.3)			

Table 4.7: Test Scenario - Simulation Settings

To accomplish valid and reliable results, each combination of parameter importance values has been tested during 10 game sessions per scenario, with different seeds (different scenarios, movements and users' behavior). Table 4.8 gives a brief summary of the simulation procedure.

Factorial Design				
Scenarios	School Yard and Test scenario			
Simulations per scenario	10			
Number of parameters' profile combinations	16			
Total number of simulation per scenario	160			
Total number of simulation	320			

Table 4.8: Factorial Design - Simulation Settings

4.5.4 Metrics for Algorithm Evaluation

Given the broad group of parameters the algorithm has and their different *nature* and aiming at studying the influence of them in the DS selection, the following group of metrics is used:

Number of DS nodes - The number of nodes in the DS, or the number of DOMINATOR nodes or Zone Servers, represents the number of the required nodes to cover all the client nodes in the network. PBS assumes that every node/client has to have at least one neighbor in the

DOMINATOR state.

This value is aimed at being as small as possible. In the limit this should be the MDS but, in our approach, we look for the best approximation preferring first to have a stable and robust DS.

Number of DS changes - This metric gives a measure of stability of the selected DS by measuring how often the DS changes (new nodes are selected for the DS) during the service run.

In the simulation, it was admitted that a DS node will remain in the DS even if all its clients are covered by other DOMINATOR nodes. This problem was already studied in [9]. It was observed that if a node is removed from the DS because it is redundant, oscillation problems may be observed in the DS. Removing redundant nodes may "free" some nodes from the DS, but in general other nodes have to be selected after some small amount of time, which can cause extra operations to be done, new elections and service state transferring/synchronization for the new DOMINATOR nodes. This may not be a good trade because the price of reducing the DS can cause instabilities and an increasing the message overhead.

- Number of observed anomalies As has been introduced in section 3.1.1, one of the aims of the proposed algorithm is at allowing the selection of a powerful DS. If it is required to deploy a given service, the DS nodes should have sufficient resources (processing, energy) to serve, in "good" conditions, their network clients. This metric measures the number of times that DOMINATOR nodes are not able to support properly the service that is being deployed and/or the clients that they are serving. In concrete it counts the number of times CPU and/or Memory load is above a threshold value (the value considered is 90%) for some period of time, or the energy level is below a threshold value (the value considered is 10%). If such conditions are observed, an anomaly is generated and a new election takes place.
- **Packet Loss** This metric gives a measure of the number of packets that are lost in the communication between clients and zone server nodes.

Latency - Measure the average time that a packet requires to be transmitted between a DOMINATEE and a DOMINATOR, or vice versa. Latency and the packet loss allow extracting conclusions about the quality of the links between DOMINATOR nodes and their DOMINA-TEE nodes. These two values will be mainly affected by the network congestion, which is mainly caused by the density of the nodes' transmitting information and the quality of the links between the nodes.

4.6 Evaluation and Best Profile Computation

In the next points the results of the designed experimental design are going to be presented. First a global picture of the results is depicted, performing an analysis of the metrics, taking all the used simulated combinations into account (the used importance values combinations are presented in section 4.4.1). In a second point, the combination of importance values that conduct to the best results, referred as the best profile for the used test service, is computed and studied.

4.6.1 Global Algorithm Performance

Results

Figure 4.3 represents the observed Number of DS nodes, Number of DS changes and Number of anomalies in the 16 performed simulations. Figure 4.4 represent the Average packet loss (in percentage) and Average latency (in milliseconds) observed in all the factorial design for the school yard scenario. The equivalent results for the test scenario are presented in Figure 4.6 and Figure 4.6, respectively.

Table 4.9 and table 4.10 give a brief summary of the statistics observed in both scenarios' simulations.

Analysis

In this section a global analysis of each metric is done, explaining the evolution of it in the performed simulations and analyzing the most important



Figure 4.3: School Yard Scenario - DS nodes, DS changes and Anomalies



Figure 4.4: School Yard Scenario - Packet loss and Latency

factors in the output of each observed metric.

Number of DS nodes



Figure 4.5: Test Scenario - DS nodes, DS changes and Anomalies



Figure 4.6: Test Scenario - Packet loss and Latency

The high "priorized" nodes are selected for the DS but the number of DS nodes is dependent on how the different parameters are weighted. If the degree is assigned with a *high importance* value the natural

	School Yard Scenario						
	DS nodes	DS Changes	Anomalies	PKT loss	Latency		
min	3	0	0	0.012	5.23		
max	6	9	4	1.54	7.23		
avg	3.90	1.99	0.76	0.30	6.84		
σ	0.58	1.99	0.21	0.05	0.5		

Table 4.9: Global Statistics - School Yard scenario

Test Scenario						
	DS nodes	DS Changes	Anomalies	PKT loss	Latency	
min	4	0	0	0.56	5.12	
\max	17	18	5	2.65	7.12	
avg	8.29	5.18	1.21	1.19	6.51	
σ	1.48	4.57	0.26	0.14	0.6	

Table 4.10: Global Statistics - Test scenario

conclusion would be that the number of DS nodes would decrease but, since the algorithm is not just dependent on this parameter, different results were also observed as was expected.

In the school yard scenario it is always observed a quite small DS, Figure 4.3, which is basically explained by the scenario properties: the network has 25 nodes and the scenario has $500\times500 \ m^2$, which allows a complete coverage of all nodes by a DS of 3 to 6 nodes - in average 3.9 DS nodes are required. The best performance was observed when the parameters' importance combination 6 was used. In this combination the node's degree is assigned with a *high importance* value.

In the **test scenario** different results were observed, figure 4.5. The main difference, compared with the other scenario, is the increasing number of required DOMINATOR nodes. This scenario is larger than the school yard scenario and has a bigger number of nodes. Another factor that influences the number of servers is the higher speed of the nodes (this is also observed in the number of DS changes). These results were even enforced by the fact that redundant servers remain in

the network. In average a reasonable value of DOMINATOR nodes (8.29 nodes) was achieved even in such conditions. The best case was observed when the parameters' importance combination 10 was used, where the node's degree is also assigned with a *high importance* value.

Number of DS changes

It can be deduced that the number of DS changes varies a lot when the two scenarios are compared. The increasing number of DS nodes could naturally make a selection of not so powerful nodes for the DS, but, based on the observed number of anomalies in each scenario, this is not the most important cause for the considerable differences in the DS changes in both scenarios. In both scenarios was observed that, when the more powerful nodes were selected for the DS, a more stable DS was achieved. This may be observed in the decreasing number of changes in the combinations above 13, when processing power parameters are assigned with *high importance* value.

The school yard scenario shows a very stable DS, figure 4.3. In some cases very few or no changes occurred during the whole game session. In average a value of about 2 changes has been observed. The best case (mean of 1.43 changes) was observed with the parameters' importance combination 13, where cpu and memory are assigned with *high importance* and link quality and degree are assigned with *normal importance*.

In the **test scenario**, 4.3, the best case was observed when, as in the school yard scenario, the parameters' importance combination 13 was used. This enforces the idea that selecting powerful nodes for the DS, may be very important in the selection of a stable DS, but it does not mean that it increases the performance in other points, as the number of DS nodes, or link quality.

Number of observed Anomalies

In average the number of anomalies does not vary too much between
the two scenarios. In both an average of about 1 anomaly is observed. The observed variations are mainly explained by the increasing number of nodes in the network for the test scenario which increases the probability of selection of not so powerful nodes for the DS. However this result is very dependent on the type of devices that are in the network, their capacity and also the service's generated processing load. It is clear that when the processing power parameters are assigned with *high importance* value, a reduction on the observed number of anomalies occurred. This result was expected, given that the number of anomalies is only influenced by this parameters and hence when their importance is increased a reduction on the number of anomalies should always be expected.

The school yard scenario shows an average of 0.76 anomalies, figure 4.3. This value reflects that in general nodes with sufficient power resources are selected for the DS. Such result is mainly explained by the fact that *normal importance* or *high importance* is assigned to the processing power parameters and this allows to enforce the selection of the more powerful nodes for the DS. The best results were obtained when the weight factor combination 13 was used, similarly to the case of the observed minimal DS changes.

In the **test scenario** an average of 1.21 anomalies were observed, figure 4.5. As it was theoretically expected, the increasing of network number of nodes provoked also an increase on the observed number of anomalies. Best results were observed when the weight factor combination 13 was used. This result is coherent with the one obtained in the schoolyard scenario.

In figure 4.7 and figure 4.8 two examples of the DS number of nodes variation are given for both scenarios, when profile 5 was used.

Packet Loss

In both simulations a very low packet loss rate was observed. This



Figure 4.7: Number DS nodes - School Yard scenario



Figure 4.8: Number DS nodes - Test scenario

rate is mainly dependent on the network congestion and the quality of the links between the network nodes. The results are very close in all the simulations.

The school yard scenario has a very low average packet loss rate,

0.30%. The best situation was observed in the case of the parameters' importance combination 5. This combination assigns *high importance* value to the link quality and degree parameters. Theoretically the best results would occur when just link quality would be assigned with *high importance* weight factor, but the fact is that there is a very small difference in the results for all the combinations. Such result emphasizes the idea that it is sufficient, in this case, to assign a *normal importance* value to the link quality parameter to achieve good results.

In the **test scenario** similar results to the school yard scenario were obtained. The best case was also obtained with the same parameters' importance combination, 5, was used. In this case an higher packet loss rate was observed, which makes sense because the network is more congested due to the higher number of nodes in the network.

Latency

Small values for latency were obtained (for both simulation about 6 ms) which means that this issue is not very relevant in the selection of the DS nodes, given that in all the different combinations the DS nodes position is near to their clients.

The best results were obtained, in both scenarios, for the combination 13. This combination has *high importance* value assigned for cpu and memory parameters and *normal importance* value for the link quality and degree weight factors. This result seems at the first glance not coherent, but as also observed for packet loss it may be explained by the fact that, for this service, it is enough to have a *normal importance* assigned to this weight factors to allow the selection of appropriate nodes, with good link quality and thus low packet loss and latency.

4.6.2 The Best Profile

After performing a global analysis of all used combinations, in this section the selection of the best parameters' importance values combination, or the best profile, is made. The criterion to define the best profile, for a given service and system, is not a straight operation given that the system has several requirements and objectives to accomplish. Thus an optimization of several "objectives" is required to select the most appropriate profile for the system.

In the DS selection there are objectives that cannot, in general, be satisfied at the same time. As an example get the best MDS approximation and the most stable DS. These two metrics are, in general, not possible to achieve at the same time.

The simulated service, a real time multiplayer game, has some demands in terms of processing power and link quality, see section 4.2. In the decision of the best profile the most stable DS is preferred to the MDS approximation. This is also a coherent decision with some previous assumptions (remark that redundant servers are not removed from the DS). As already has been argued, to serve some types of services, including *real-time applications*, in general a stable DS will allow to achieve a more suitable DS than the MDS approximation.

In the next section the best profile is derived. A *Multi-objective Optimization* technique is used to compute it since the best profile is dependent on several objectives.

The best profile

The main aim of the NWC algorithm is at computing the most robust and stable DS given the service requirements/characteristics. In the used test service, the objective function may be defined as: get a stable and powerful DS with good links between DOMINATOR nodes and their DOMINATEE ones, with the best MDS approximation.

The optimization of multi-objective functions is not a trivial problem given that several objectives are aimed at for a system and some objectives may not be possible to accomplish at the same time. To perform the computation of the objective function, usually, the problem is reformulated, approaching it to a single-objective problem. Such approximation is achieved by forming a weighted combination of the different objectives or else replacing some of the objectives by constraints.

In the approach followed, an experimental design, based on simulation, has been done, using several metrics to evaluate the algorithm performance. Based on these metrics and from the study developed in the previous section, it can be deduced that for the studied system the most relevant variables are the *Number of DS nodes*, *Number of DS changes* and the *Number of Anomalies*. This is explained by the fact that *Packet loss* and *Latency* have shown a constant behavior for all the used parameters importance combinations. This result means that these two metrics will not influence the search of the best profile given that in all the profiles very close, almost constant, results were obtained.

In the search of the best profile a minimization technique is used and the objective function is defined as follows:

```
Obj.Func. = w(Nodes) \cdot f(Nodes) + w(Changes) \cdot f(Changes) + w(Anomalies) \cdot f(Anomalies) (4.1)
```

As it can be seen in equation 4.1, the objective function depends on the observed number of DS nodes, the number of DS changes and the number of observed anomalies. Each of these variables has an associated weight, which defines the importance of each of them in the global Objective Function. The definition of these weights depends on the importance of each of the factor/objective in the context of the system/service. It was referred before that the aim in the definition of the best profile is to look at the combinations that revealed the more stable DS selection. This objective is mainly influenced by the number of DS changes and the number of observed anomalies. Therefore, these two objective function variables are assigned with higher weights when compared with the remaining variable, the number of DOM-INATOR nodes.

Definition of the weights for the objective function computation:

Number of DS changes - Higher weight, because this metric measures the DS stability and given that the main objective is to select the profile that gives the most stable DS, this is assigned with the highest weight.

Number of Anomalies - It measures the number of times that DS nodes have to be removed from the DS. This is also a measure of DS stability given that is aimed at having the smallest number of anomalies to grant a stable and powerful DS. Based on this, this metric is assigned with the second highest weight.

Number of DS nodes - The number of DS nodes measures the number of required nodes to form the DS and assure a full coverage of all nodes in the DS. The MDS approximation allows increasing the system's performance but, on the other side, having a small number of DS nodes does not assure the stability of the DS. In the objective function computation this metric is assigned with the lower weight, given that it important and is an objective for the system to minimize this metric. However stability is preferred for the analyzed system.

Objective Function Weight Factor				
Metric	Weight			
Number of DS changes	5/10			
Number of Anomalies	3/10			
Number of DS nodes	2/10			

Table 4.11: Objective Function Weights

In appendix A, the detailed computation of the best profile is done, using the referred *multi-objective* computation approach. Figures 4.9 and 4.10 represent the results of the objective function computation, for each parameters' importance value combination.

From the performed computations it may be deduced that the profile 13 is the one that best fulfills the defined objective function for both scenarios.

Table 4.12 presents the profile that gives the best results for the analyzed system and Tables 4.13 and 4.14 summarize the observed metrics in the school yard and test scenario, respectively.



Figure 4.9: Objective Function - School Yard Scenario

The Best Profile					
cpu memory battery link quality degree					
1	1	0	0.5	0.5	

Table 4.12: The Best Profile, in Both Scenarios

School Yard Scenario						
Combination DS nodes DS Changes Anomalies PKT loss Latency						
13	3.87	1.43	0.43	0.35	6.75	

Table 4.13: Best Profile Results - School Yard Scenario

As it can be observed, this is the profile that gives the best performance in both scenarios, according to the desired objectives for the system. The minimum number of DS changes, the minimum number of anomalies and the minimum values for latency are observed when this profile is used, in both



Figure 4.10: Objective Function - Test Scenario

Test Scenario						
Combination DS nodes DS Changes Anomalies PKT loss Latency						
13	8,2346	4,4286	0,8571	1,0735	6,4214	

Table 4.14: Best Profile Results - Test Scenario

scenarios. The observed number of DS nodes is also close to the minimum observed in both scenario simulations.

4.7 Summary

In this chapter the proposed algorithm has been tested in the selection of the DS for a test service in two different scenarios. The definition of the best profile, for this service, has also been made using the NWC framework. Simulation is a very useful tool to evaluate algorithms, mainly when the complexity of real implementation is an issue. However transposing the real conditions of a system to a simulator always requires some grade of abstraction.

The test service used during the simulation is constructed based on the properties of real time multiplayer games, section 4.2. The service properties are presented in section 4.3. Given the properties of the service, the main objectives were to select a stable and robust DS with possible good approximation to the MDS.

The service has been tested under two different scenarios, with different properties. This was used to compare and evaluate the performance of the selection under different conditions, for the same type of service.

The algorithm allows an adaptation to the type of service/system according to its requirements by the adjustment of the parameters' importance values. In the simulation the best profile for the simulated service was made using the NWC framework.

In section 4.6.1 the obtained results are shown for both scenarios.

The best profile was computed using a multi-objective optimization technique, given that more that one objective was aimed at for the DS selection. It was observed, for both scenarios, that the best profile for the simulated service will consist of powerful nodes (in terms of processing power). Such results are explained by the service's characteristics. The service processing load was quite high compared to the devices' processing power characteristics, which may generate some problems if devices with low processing power capabilities are selected for the DS.

The same best profile was observed for both scenarios. This confirms that defining a constant profile for a given DS selection and type of service may be a good approximation. However based on the performed experiments it may not be affirmed that this is a universal conclusion. But based on the observed results it may be concluded that, even varying the scenario conditions, the service has pattern characteristics associated to it, which influences the DS selection and changes in the same way. This enforces the fact that using a static assignment of parameter importance values for a given service may be a good approach.

Chapter 5

Implementation

This chapter gives an overview of the implementation of the NWC algorithm in NS-2 [7] and in the Siramon framework [4]. In both implementations the NWC algorithm is implemented as an extension of the basic PBS algorithm developed in [6]. The code of both implementations can be found in the Siramon project webpage 1 .

5.1 NS-2 Implementation

The implementation of the NWC algorithm in NS-2 has two main objectives:

- Implement the NWC algorithm (to compute the node's weight) for the DS selection.
- Create a framework (*NWC framework*) that allows the study of the best profile of parameters' importance for a service, based on the defined objectives for the DS selection.

5.1.1 The NS-2 Network Simulator

The network simulator NS-2 [7] is a discrete event simulator targeted at networking research. It provides support for simulation of TCP, routing and multicast protocols over wired and wireless networks.

¹www.siramon.org

NS-2 is written in C++ and OTcl. In the C++ space protocols and all functionality that is not changed often are implemented. In the OTcl space the scripting functionality to control the simulations and perform the required interface with the C++ space is placed.

The reason to use two different spaces, with two different languages, is explained by the complexity of each space. C++ is better suited for write complex operations. For example, implementation of protocols and data control. On the other hand OTcl allows a fast and simple construction of simulation scripts, which allows an easy creation/change of the necessary scripts to perform the simulations.

In [7] a collection of material about NS-2 can be found.

5.1.2 NWC Algorithm Implementation

The NWC algorithm has been implemented as an extension of the PBS algorithm [6] in NS-2 and also in the Siramon framework.

PBS Algorithm Implementation

In NS-2, the main functionality of PBS has been implemented as an agent (ZSSpbsAgent). This was implemented as part of the core (C++ space) of the NS-2 simulator to allow the different network nodes to attach this functionality in the configuration scripts (OTcl space). Basic zone server selection functionality is implemented in another agent (ZSSAgent) to allow the extension to implement different zone server selection algorithms. The basic communication between agents is made by a dedicated protocol (PT_ZSS) and packet format that consists of a content and content length field, and can be used by any zone server selection implementation. The algorithm implementation follows a Finite State Machine (FSM) which defines states and transitions between states, controlled by incoming events. The use of the FSM allows an easy control of the algorithm flow. The FSM used for the PBS algorithm is in Appendix B, it also already contains the developed NWC algorithm.

NWC Algorithm Implementation

The aim of the NWC algorithm NS-2 implementation is to provide the PBS algorithm with the nodes weighting functionality but also allow the study of the best profile of parameters' importance for a given service. Figure 5.1 shows the NWC functionality implemented in NS-2, as an exten-



Figure 5.1: NWC NS-2 Implementation

As it can be observed in Figure 5.1, the implementation is divided by the two NS-2 spaces, OTcl and C++ spaces.

In the OTcl space all the necessary variables for the simulation are assigned and then passed to the C++ space where the NWC algorithm is implemented as an extension of the PBS algorithm. The three basic groups of variables that are passed to the C++ space are:

• Service Specifications - It is used to characterize the type of service

that is simulated. Variables representing CPU/memory load generation, battery consumption and also traffic (rate) between zone servers and its clients are defined and passed to the PBS agent, in the C++ space, to be used in the service simulation.

- Nodes Nodes' local parameters capacity (CPU, memory and battery) which are used to simulate the CPU/memory load generation and battery consumption, are assigned to every node of the network that participates in the service deployment. This will allow the computation of the parameters' values for the node's weight computation.
- **Parameters' Importance** Represent the importance values assigned to the different parameters of the NWC algorithm. When computing the best profile these values are changed and the results of the simulation are analyzed to select the most appropriate combination, according to the simulated service and the objectives defined for the selected DS.

In the C++ space some extensions to the PBS algorithm have also been made. The basic NWC algorithm implementation has been included in the PBS Agent. This node's weight computation functionality is called by the PBS algorithm each time an election takes place, as it presented in the FSM in Appendix B.

In the C++ space the following functionality has been added:

• Service Monitor - This functionality is responsible for monitoring and checking the local parameters values. It also controls the traffic generation between zone servers and clients, which is performed according the service specifications. This functionality is called at each election time to update the parameters values for the node's weight computation. Periodically statistics about the parameters are taken and the local parameters (CPU, memory and battery) values are check, if they are not in the range of admissible values an anomaly is generated, this question is discussed in more detail in section 3.4. • Local Parameters - This functionality is responsible for the generation of the local parameters values according to the node's capacity and the service characteristics.

In table 5.1 a summary of the file structure of the PBS algorithm with the NWC algorithm extension is given. More details can be found in the programming documentation in Siramon project webpage ². In Appendix C the steps for installation of the PBS/NWC algorithm are given.

PBS/NWC implementation in NS-2			
File	Description		
$zss{.h.cc}$	The main class for any ZSS implementation.		
$zss_pbs{.h.cc}$	Implements the PBS/NWC algorithm sup-		
	port.		
$timer_pbs\{.h.cc\}$	Timer functionality used by the PBS/NWC		
	algorithm.		
neighborlist $\{.h.cc\}$	Stores and handles the relevant information		
	about the neighbors.		
$node_information.h$	Contains the structure of a node entry in		
	the Neighborlist and the structure of the sent		
	packets by the PBS algorithm.		
monitor{.h .cc}	Monitors the 1-hop neighborhood.		
service_monitor{.h .cc}	Implements the service monitor functionality.		
timer_monitor{.h .cc}	Timer functionality used by the monitor, and		
	the service monitor.		
local_parameters{.h .cc}	Local parameter values generation.		
debugger{.h .cc}	Debugger used to write outputs.		
utils/fsm $\{.h.cc\}$	Implementation of a Finite State Machine		
	(FSM).		
utils/state{.h .cc}	State of the FSM.		

Table 5.1: Files Used for PBS/NWC Implementation in NS-2

²www.siramon.org

5.2 Siramon Implementation

The implementation of the PBS algorithm in the Siramon framework follows the same approach of the NS-2 implementation, it is an extension of the PBS algorithm, and this implementation is presented in this section.

Siramon is implemented in Java and has five basic modules:

- Service Specification
- Service Indication
- Service Deployment
- Service Management
- Environment Observer

A more detailed description of the Siramon framework is given in Appendix D.

The NWC algorithm is implemented in three of the basic Siramon's framework modules: *Service Specification*, *Service Management* and *Environment Observed*:

• Service Specification - This module is responsible for the service description document definition and manipulation. This document is written using the XML language [30]. To support the NWC functionality the document has been extended with the profile of parameters' importance values (represented by the fields *cpu_i*, *mem_i*, *bat_i*, *link_i* and *deg_i*). This information is added under the DEMANDS element (path SERVICE/IMPLEMENTATION/CODE/ENVIRONMENT/DEMANDS) of the document. These fields are optional and will just be checked if the service requires the zone server selection functionality. An extension on the specification module has also been made to allow the extraction of the service profile.

- Service Management This module is responsible for the maintenance, reconfiguration and termination of the services that use the framework. Since the PBS algorithm is responsible for the selection and maintenance of the zone servers, for a given service, it has been implemented in this module. The NWC algorithm has also been implemented in this module as an extension of the basic PBS algorithm. To compute the node's weight the necessary values of the parameters and the parameters' importance values are obtained using the specification module and the Environment Observer module.
- Service Observer The Environment Observer module has the functionality of monitoring the device resources and the service context on the network. To compute the node's weight the NWC algorithm has to monitor the CPU and memory load, the battery energy level and the link quality of the connections to its neighbors. This functionality has been included in this module. Four external monitors have been created to monitor the four parameters, because Java does not have native functionality to support the extraction of the required information for the node's weight computation.

Tables 5.2 and 5.3 give a resume of the file structure of the PBS algorithm with the NWC algorithm extension in the Siramon framework. In Appendix C some details about the implementation and the used external monitors are given.

Package	Description
Siramon.Management.zss	Contains the main classes needed by
	the ZSS functionality.
Siramon.Management.zss.utils	Contains some utils for the ZSS
	functionality like a basic debugger
	and the Finite State Machine (FSM)

Table 5.2: Zone Server Selection Functionality in Siramon Framework

File	Description
ZSS.java	The main class for any ZSS implementation.
ZSS_PBS.java	Implements the PBS/NWC algorithm.
ZSS_PBSfsm.java	Implements the FSM used by the PBS/NWC
	algorithm.
ZSS_PBSpacket.java	Used packet format to be sent in the content
	part of the SIRAMON packet.
TimerPBS.java	Timer functionality used by the PBS/NWC
	algorithm.
Neighborlist.java	Stores and handles the relevant information
	about the neighbors.
NodeInformation	Used by the Neighborlist to store the infor-
	mation about a neighbor.
utils/Debugger.java	Debugger used to write outputs.
utils/FSM.java	Basic implementation of a Finite State Ma-
	chine (FSM).
utils/State.java	State of the FSM.

Table 5.3: Used Files for the PBS/NWC Implementation in SIRAMON

Chapter 6

Conclusions and Outlook

In this chapter conclusions about the achievements of this Master's Thesis are depicted and directions for further developments of the work are given.

6.1 Conclusions

New ways of communication are gaining popularity nowadays. MANETs are one of these new trends, so new services and business areas are emerging for these networks. To support such networks new protocols and functionality are required to cope with the demanding and especially dynamic conditions associated to it.

The support of distributed services deployment in MANETs has different demands when compared to wired networks. The classical "server-client" and "peer-to-peer" architectures are not well suited for the deployment of distributed services in a MANET. New service management architectures are being proposed for MANETs. An example of such architecture is the Zone-based architecture [5]. It is a hybrid architecture that combines the "good" features, in the MANETs context, of the two classical architectures. This architecture is based on the notion of zone servers, which are special network nodes that work as servers of a given network zone managing the nodes that belong to its zone. Previously an algorithm called PBS (Priority Based Selection) [6] has been proposed for zone server selection and maintenance. PBS treats the network as a graph and selects the DS based on node's priorities. The first criterion for the node's priority assignment is the node's weight, which measures the capacity of a given node to perform as zone server for a given service.

Several algorithms exist to perform the selection of the DS nodes. The most common approach for such selection is based on a small group of nodes' characteristics aiming to optimize the DS selection for a unique objective. In this thesis a different algorithm is proposed. The NWC (Node's Weight Computation) algorithm performs the selection of the most appropriate DS as function of the service/system requirements. The algorithm, in conjunction with the PBS algorithm, allows the selection of DS nodes for non-specific types of services, meaning that different objectives may be defined for the DS selection. The algorithm is used in the Siramon framework [4] providing the *weighting* functionality to the PBS algorithm to select the DS.

A group of parameters is taken into account to quantify the weight of the network nodes, in total five parameters are used. The parameters are divided into two classes, node's processing/power parameters and node's communication parameters. The first group contains the parameters which measure the processing power capabilities of the nodes (CPU and memory) and the energy capabilities of the nodes (battery). Concerning to communication parameters the node's degree and the link quality parameters are used, to measure the communication capabilities of a node with its neighbors.

To use the algorithm in the deployment of different services, the notion of parameter importance level has been introduced. These importance values are assigned to the parameters as function of the service/system requirements. For example, if the service requires good connections and fast processing nodes, the processing parameters (CPU and memory) and the link quality parameter are assigned with a *high importance* value.

To allow the differentiation of the parameters' importance a scale of three values is used (*no importance, normal importance* and *high importance*), see section 3.3.1 for more details.

The definition of the parameter importance values for a service, or service profile, is made in a static way. This means that it is not adaptable to the current service context. To study the most appropriate combination of importance values for a given service a framework has been created, the NWC framework.

The NWC framework allows, by performing a factorial design based on the service properties, to study the contribution of each parameter in the selection of the DS. To find the most appropriate profile a multi-objective optimization technique is used. Such technique is used because a multiple objectives are usually defined for the DS, such minimum number of DS changes and minimum number of DS nodes.

The definition of the service profile may also be defined directly by the service/system designer, according to the goals defined for the DS.

The NWC algorithm has been tested using simulation. A test service was used modeling a real service (real time multiplayer game) characteristics. It was simulated in two different scenarios. The NWC framework has been used to evaluate and search for the best profile for the service. Good results were obtained and it was clear that in all the simulated scenarios one profile was the more appropriated, given the defined objectives for the DS. Such conclusions confirmed that the proposed static assignment will be, in general, consistent in the selection of the DS nodes for a given service, in different conditions. This is mainly explained by the fact that a service is characterized for a set of properties and characteristics that generate, even in different contexts, a pattern behavior and this is reflected in the requirements for the DS selection.

The NWC algorithm has been implemented in the Siramon framework [4]. Given the constraints of the testbed, it could not be observed and studied the importance of the use of the NWC algorithm in the selection of the DS nodes. However it was observed that the more powerful nodes were selected for the DS when the algorithm was used.

6.2 Outlook

From the performed work, some future lines of investigation and development of the proposed algorithm are here depicted:

- Different scenarios During the performed evaluation of the algorithm two different scenarios have been used. A random way point mobility model was used in these simulation. To study the NWC algorithm in more realistic conditions other scenarios should be also tested. An example that is nowadays having some interest is the vehicular ad hoc networks. Some simulation scenarios are being proposed for this cases, for example an *highway scenario* where the movement of cars is simulated in a high way. Such tests will allow to conclude about the performance of the algorithm in more realistic scenarios.
- Node's weight computation formula The NWC algorithm formula for node's weight computation follows a linear combination of the parameters values assigned with a given importance value. Every parameter value is normalized into an unitary scale and the weight for a node is then computed. Given the characteristics of the parameters and also the importance that they have in the DS selection for a given service, different combinations may conduct to a more appropriate selection of the DS nodes. Some more research in the weight computation formula should be done, studying different combinations of the parameters and their importance values.
- Testbed The test and evaluation of an algorithm is usually performed under simulation due to the complexity of construct a real test environment. But it is impossible to transpose all service/scenario/system properties for a simulation. It is suggested to test the NWC algorithm in a real scenario, testbed. A representative testbed containing a large number of different types of devices and reproducing real service deployment situation will allow to perform measurements and depict conclusions about the algorithm performance. This will allow to study and develop the algorithm further.

Appendix A

Best Profile Computations

In this chapter the computation for the best profile using a multi-objective optimization technique is performed.

A.1 Factorial Design Results

In this section the results of the performed factorial design are presented. In Table A.1 the importance values, or profile, combinations to perform the experimental design are presented. After, Tables A.2 and A.3 present the results of the factorial design for the school yard scenario and test scenario, respectively.

A.2 The Best Profile

There are five metrics that measure the performance of the system and the objective is to minimize all these metrics, for the best profile computation. Such a goal is not possible because some of the objectives cannot be optimize at the same time.

To accomplish the optimization a multi-objective optimization technique is going to be used. This is used in the computation of the best profile for the service in both scenarios.

Importance Values Combinations					
Combination	CPU	memory	battery	link quality	degree
1	0.5	0.5	0	0.5	0.5
2	0.5	0.5	0	0.5	1
3	0.5	0.5	0	1	0.5
4	0.5	0.5	0	1	1
5	0.5	1	0	0.5	0.5
6	0.5	1	0	0.5	1
7	0.5	1	0	1	0.5
8	0.5	1	0	1	1
9	1	0.5	0	0.5	0.5
10	1	0.5	0	0.5	1
11	1	0.5	0	1	0.5
12	1	0.5	0	1	1
13	1	1	0	0.5	0.5
14	1	1	0	0.5	1
15	1	1	0	1	0.5
16	1	1	0	1	1

Table A.1: Parameters' Importance Combinations for Factorial Design

A.2.1 Multi-Objective Optimization

A multi-objective problem consists of the search of the values for the design variables which optimizes a set of objective functions. The set of variables that produces the optimal outcome is designated as the optimal combination. It yields a set of possible answers from which the designer may choose the desired values of the design variables.

The usual approach in the search of the *global* optimization consists of weighting the several objective functions, or decreasing the number of used objective functions by setting constraints. In the following point a *weighting objectives method* is presented.

School Yard Scenario						
Combination	DS nodes	DS Changes	Anomalies	PKT loss	Latency	
1	3.98	2.14	0.71	0.30	6.94	
2	3.82	2.57	0.86	0.25	6.82	
3	3.98	1.71	0.71	0.34	6.91	
4	3.88	1.86	0.71	0.23	6.77	
5	3.80	2.29	0.86	0.28	6.82	
6	3.71	2.43	1.00	0.25	6.84	
7	4.05	2.86	1.29	0.27	6.86	
8	3.77	2.43	0.86	0.29	6.88	
9	4.19	2.43	0.86	0.25	6.88	
10	3.98	1.57	0.57	0.30	6.77	
11	3.95	1.57	0.71	0.36	6.85	
12	4.01	2.00	0.86	0.30	6.89	
13	3.87	1.43	0.43	0.35	6.75	
14	3.77	1.43	0.57	0.36	6.85	
15	3.96	1.71	0.57	0.36	6.84	
16	3.77	1.43	0.57	0.36	6.83	

Table A.2: Factorial Design Results - School Yard scenario

Weighting Objectives Method

This method takes each objective function and multiplies it by a "weighting coefficient", which is represented by w_i . The modified functions are then added together to obtain a single cost function. The objective functions, f_i , must be optimized before the multi-objective optimization.

$$f(x) = \sum_{i=1}^{k} w_i f_i(x)$$
where: $0 \le w_i \le 1$ and $\sum_{i=1}^{k} w_i = 1$
(A.1)

In this method the weighting coefficients are assumed beforehand. The designer is expected to pick the values of the variables from this set of solutions.

Appendix A

Test Scenario						
Combination	DS nodes	DS Changes	Anomalies PKT loss		Latency	
1	8.0079	5.7143	1.4286	1.3603	6.4327	
2	8.4862	6.0	1.4286	1.0546	6.5939	
3	8.1992	5.1429	0.8571	1.122	6.6504	
4	8.4653	5.1429	1.1429	0.9305	6.5207	
5	8.4902	5.7143	1.2857	1.176	6.5328	
6	8.6763	5.1429	1.2857	1.2371	6.4607	
7	8.4076	5.0	1.2857	1.3829	6.5003	
8	8.3879	4.4286	1.0	1.0782	6.5117	
9	8.2385	5.1429	1.2857	1.1768	6.5258	
10	7.8952	5.8571	1.7143	1.0728	6.4482	
11	8.0917	4.7143	1.1429	1.4099	6.5436	
12	8.1953	5.8571	1.5714	1.3021	6.4536	
13	8.2346	4.4286	0.8571	1.0735	6.4214	
14	8.1009	4.5714	1.1429	1.0894	6.5254	
15	8.3709	5.1429	1.0	1.2234	6.5112	
16	8.4535	4.8571	0.8571	1.3612	6.5603	

Table A.3: Factorial Design Results - Test scenario

A.2.2 Best Weight Profile Computation

In the designed experiment $f_i(x)$ represent the measured (mean) values of the metrics used for the system evaluation:

- Number of DS Nodes
- Number of DS Changes
- Number of Anomalies
- Packet Loss
- Latency

From the performed study in 4.6.1, it has been concluded that packet loss and latency metrics are not relevant in the evaluation of the system's performance. This happens because they show a constant behavior for all used combinations of weight factors.

Hence, the multi-objective function will have three basic objectives:

- Minimize the Number of DS nodes
- Minimize the Number of DS changes
- Minimize the Number of Anomalies

As it has been presented, the global objective for the evaluated system is first to guarantee a stable and robust DS selection and second to the best MDS approximation. Based on such assumptions, the combinations of weights presented in Table A.4 were used to find the best profile.

Objective Function Weight Factors				
Metric	Weight			
Number of DS changes	5/10			
Number of Anomalies	3/10			
Number of DOMINATOR nodes	2/10			

Table A.4: Objective Function Weights

In both scenarios, for each combination the defined multi-objective function (MOF) can be then described as:

$$MOF = 2/10 \cdot Num_Nodes + 5/10 \cdot Num_Changes + 3/10 \cdot Num_Anomalies$$
(A.2)

In the next sections the results of the MOF computation for all the parameters' importance values combinations on the school yard scenario and test scenario are presented. Before the computation of the MOF a normalization of all the metrics into a unitary scale has been made.

A.2.3 School Yard Scenario

Table A.5 represents the MOF computations. It can be observed that the best profile, minimal value of the MOF, occurs when the combination 13

is used. This combination has *high importance* assigned to CPU and memory parameters and *normal importance* assigned to link quality and degree parameters.

School Yard MOF Computation						
Combination	Num_Nodes	Num_Changes	Num_Anomalies	MOF		
1	0.56	0.50	0.33	0.52		
2	0.23	0.80	0.50	0.60		
3	0.57	0.20	0.33	0.31		
4	0.35	0.30	0.33	0.32		
5	0.19	0.60	0.50	0.49		
6	0.00	0.70	0.67	0.55		
7	0.70	1.00	1.00	0.94		
8	0.11	0.70	0.50	0.52		
9	1.00	0.70	0.50	0.70		
10	0.55	0.10	0.17	0.21		
11	0.49	0.10	0.33	0.25		
12	0.61	0.40	0.50	0.47		
13	0.32	0.00	0.00	0.06		
14	0.13	0.00	0.17	0.08		
15	0.52	0.20	0.17	0.25		
16	0.13	0.00	0.17	0.08		

Table A.5: MOF computation, School Yard Scenario

A.2.4 Test Scenario

Table A.6 represents the MOF computations for the test scenario experimental design. It can be observed that the best profile is obtained, as in the case of the school yard scenario, when combination 13 is used. This combination has *high importance* weight factor associated to CPU and memory, and *normal importance* associated to link quality and degree parameters.

Test Scenario MOF Computation					
Combination	Num_Nodes	Num_Changes	Num_Anomalies	MOF	
1	0.14	0.90	0.67	0.68	
2	0.76	1.10	0.67	0.90	
3	0.39	0.50	0.00	0.33	
4	0.73	0.50	0.33	0.50	
5	0.76	0.90	0.50	0.75	
6	1.00	0.50	0.50	0.60	
7	0.66	0.40	0.50	0.48	
8	0.63	0.00	0.17	0.18	
9	0.44	0.50	0.50	0.49	
10	0.00	1.00	1.00	0.80	
11	0.25	0.20	0.33	0.25	
12	0.38	1.00	0.83	0.83	
13	0.43	0.00	0.00	0.09	
14	0.26	0.10	0.33	0.20	
15	0.61	0.50	0.17	0.42	
16	0.71	0.30	0.00	0.29	

Table A.6: MOF computation, Test Scenario

Appendix B

Finite State Machine (FSM)

This chapter presents the Finite State Machine (FSM) used in the PBS algorithm NS-2 and Siramon implementations. As the NWC algorithm is implemented as an extension to the PBS algorithm, here an updated version of the FSM is provided.

B.1 Introduction

A FSM is a model of behavior composed of states, transitions and actions. Between different states of the machine some transitions are defined that perform the defined actions based on incoming events. The specification of the used FSM in the PBS/NWC algorithm implementation is shown in Fig. B.1.

B.2 The States

The FSM consists of four states: "idle", "msgsent", "roundfinished" and "finished" state. The "idle" state is the initial state of the FSM. PBS algorithm performs in rounds. When the node goes off the initial state ("idle") and joins the service deployment, computes its weight and sends its neighborslist, with the computed weight, to its neighbors. In every round it sends



Figure B.1: Finite State Machine (FSM)

the neighborlist to its neighbors and waits in the "msgsent" state for the neighborlists from the neighbors. If all required neighborlists have arrived, or the timeout timer has expired, it goes to the "roundfinished" state. In this state it determines its own status based on the its weight and the information available from the neighborhood. If there are still INT_CANDIDATE neighbors, the neighborlist is resent. At this point the FSM is again waiting in the "msgsent" state. If all INT_CANDIDATE neighbors have switched to DOMINATEE or DOMINATOR status, the FSM goes to the "finished" state. It is waiting in the "finished" state unless there are some changes in the network topology detected (some new INT_CANDIDATE neighbors arrive) or if some anomaly in the local parameters is observed. If this is the case, a new round of the PBS algorithm will be started and the FSM returns to the "msgsent" state. In this transition the node computes its weight to start a new election.

B.3 The Transitions and Actions

In Table B.1 shows all transitions/events and respective actions.

From State	To State	Event	Action
idle	msgsent	join	Node joins the service, computes
			node's weight and starts sending out
			neighborlists.
idle	msgsent	receivedmsg	Node in idle state, but received a
			neighborlist. Computes its weight and
			starts sending out neighborlists.
msgsent	msgsent	timerRESEND	Resend timer expired and not all
			neighbors sent a neighborlist back. So,
			the already sent neighborlist will be re-
			sent.
msgsent	roundfinished	receivedmsg	All required neighborlists have arrived.
			Node determines its own status, based
			on its weight and its neighbors weights.
msgsent	roundfinished	timerTIMEOUT	Not all required neighborlists arrived,
			but the timeout timer expired, so the
			node determines its own status.
roundfinished	msgsent	resend	There are still INT_CANDIDATE
			neighbors, a new round of PBS needs
			to be started.
roundfinished	finished	finished	All nodes determined their status, no
			INT_CANDIDATE neighbors left.
finished	finished	receivedmsg	Handles incoming neighborlist. If re-
			quired it sends a neighborlist back.
finished	msgsent	resend	Anomaly in the local parameters,
			changes in the network and/or some
			INT_CANDIDATE neighbors have
			been detected. Computes node's
			weight and starts a new round of the
			PBS/NWC.

Table B.1: Transitions of the Finite State Machine (FSM) for PBS/NWC.

Appendix C

NWC algorithm NS-2 Installation

In this chapter a tutorial for installation of the implemented functionality in NS-2 is given.

C.1 Installation

The following steps have to be taken, to install NS-2 and the PBS/NWC functionality:

- 1. Download the Ns-Allinone package from the official NS-2 homepage and install it according to the instructions.
- 2. Copy the zoneserver directory containing the implementation into the ns-2 main directory (e.g., ns-allinone-2.29/ns-2.29).
- 3. Some changes in some of the ns source files are required to add the new agent, especially because a new packet format is used:
 - Add to the file 'common/packet.h' the new packet protocol ID 'PT_ZSS' to the second last line of the 'enum packet_t {}'. In the same file you have to add the line 'name_[PT_ZSS]="zss";' to the 'p_info()' function and the line 'double snr_;' to the 'struct hdr_cmn', to support the SNR monitoring functionality.

- The file 'tcl/lib/ns-default.tcl' has to be edited, too. This is the file where all default values for the Tcl objects are defined. Insert the line 'Agent/ZSS set weight_ 0' to set the default weight for Agent/ZSS.
- The new packet has also to be added in the file 'tcl/lib/ns-packet.tcl' in the 'foreach prot' loop with the entry 'ZSS'.
- Add the following code to the 'mac/mac-802_11.cc' file, in the 'recv(Packet *p, Handler *h)' function:

```
double snr;
if(rx_state_ == MAC_IDLE) {
    snr = 10*log10(p->txinfo_.RxPr) -10;
} else {
    snr = 10*log10(p->txinfo_.RxPr/pktRx_->txinfo_.RxPr) - 10;
}
hdr->snr_ = snr;
```

- The last change is a change that has to be applied to the 'Makefile'. All sourcefiles has to be added after the 'OBJ_CC=' list: zoneserver/zss.o zoneserver/zss_pbs.o zoneserver/debugger.o zoneserver/timer_pbs.o zoneserver/timer_monitor.o zoneserver/monitor.o zoneserver/neighborlist.o zoneserver/utils/fsm.o zoneserver/utils/state.o zoneserver/link_process.o zoneserver/local_parameters.o zoneserver/service_monitor.o.
- 4. Recompile ns by using make clean; make depend; make.
- 5. After recompilation the Tcl scripts can be used to run simulations with the command ns example.tcl.
Appendix D

Siramon Framework

In this chapter a brief description of the Siramon framework is given.

D.1 Introduction

SIRAMON [4] (Service provIsioning fRAMework for self-Organized Networks) is a generic, decentralized and modular service provisioning framework for self organized networks. It integrates all the required service provisioning functionality to support the whole life-cycle of a service, integrating service specification, advertisement/lookup, deployment, management and environment monitoring mechanisms.

SIRAMON structure consists of five different modules. Each module has different functionality, but they cooperate with each other to give a complete service provisioning framework to the application layer services. Figure D.1 represents the SIRAMON structure.

D.2 The Structure

Next, a brief description of each of the modules is given:

Service Specification - Defines a universal service description language to describe the services that will be supported by the framework. XML



Figure D.1: SIRAMON architecture

Information Set [30] is used for the service description. The services are defined in a tree structure based on the type of service. The description is included in a XML document that is handled by the framework to allow the support of it.

- **Service Indication** It is responsible for advertisement and service lookup procedures on the network. Due the lack of a central infrastructure this procedures are completely distributed.
- **Service Deployment** This functionality includes all the necessary steps for the deployment of a service. It may have several operations associated:
 - Requesting/Downloading software.
 - Discovering/Gathering resources.
 - Mapping the service specifications to the resources.
 - Configuring the resources and installing/configuring the service software.

- Activating the service, synchronized with the other service participants.
- Handling the control on the management module.
- **Service Management** Service Management contains all the functionality of service maintenance, reconfiguration and termination functions of the running services.
 - The service maintenance keeps track on the service context, adapting dynamically the service to its context (resource variations, new nodes) in order to optimize user's perceived service quality hiding all the scenario conditions changes to the node.
 - The service reconfiguration is responsible for the reconfiguration of the service, globally or locally. Local reconfiguration is transparent to the other users. It happens when the context of the device changes or the service user intends to modify the running service session. When global reconfiguration happens, all the service users will perceive the change and a global reconfiguration/synchronization occurs.
 - The service termination can be seen as a special case of service reconfiguration, when a device stops running its service instance. This occurrence is informed to all other participants, the resources have to be released and the participant nodes will then reconfigure their service instances to the new service scenario.
- **Environment Observer** This module is responsible for monitoring the device resources and the service context on the network. It collects information about the context of the service locally/globally and transforms it into an appropriate form to be used as input for the deployment and management modules.

Siramon is a platform-independent framework, which is an imperative requirement due the heterogeneity associated to MANETs. To achieve this independence it is written in Java. For more details about the Siramon framework [4] should be consulted.

Appendix E

Used Abbreviations

- **ad hoc** for this (latin); for a particular case without any form of centralized administration.
- **DS** Dominating Set; A dominating set of a graph, is a subset of nodes such that all nodes are neighbor of a node in the DS or are itself in the DS.
- **FSM** *Finite State Machine*; A FSM is a model of behavior composed of states, transitions and actions.
- MAC Medium Access Control; Handles access to a shared medium.
- MANET Mobile Ad hoc Network; Temporary network in which devices want to communicate with each other, with a continuously changing network topology and without any form of centralized administration. MANET is also the name of an IETF working group, that is working in the field of ad hoc networks.
- **MDS** Minimum Dominating Set; A Dominating Set (DS) is called a MDS if the number of nodes in the DS is minimal.
- **NS-2** Network Simulator 2; NS-2 is a discrete event driven simulator to support networking research.
- **OTcl** *MIT Object Tcl*; An extension to Tcl/Tk for object-oriented programming.

- **P2P** *Peer-to-peer*; Network that does not have fixed clients and servers, but a number of peer nodes that function as both clients and servers to the other nodes on the network. Any node is able to initiate or complete any supported transaction.
- **PBS** *Priority Based Selection*; An algorithm that selects the Zone Servers supporting a zone-based architecture based on comparing priorities of the nodes.
- **SIRAMON** Service Provisioning Framework for Mobile Ad-hoc Networks; A proposal of a generic, decentralized service provisioning framework for mobile ad hoc networks [2].
- WLAN Wireless Local Area Network; In a WLAN the device (e.g. a laptop, PDA, etc.) communicates via a wireless connection with a WLAN Access Point which is connected (just like a normal computer) via a cable to the Internet or the local network. As the devices are not wired up the users are mobile. This is the advantage of a WLAN. The indoor range depends on structural factors and is considerably lower than outdoors, where WLAN connections are possible over more than 200 metres.
- XML Extensible Markup Language; Simple, very flexible text format for electronic publishing and data exchanging, standardised by the World Wide Web Consortium (W3C).
- **ZSS** Zone Server Selection; The selection procedure for choosing the Zone Servers supporting a zone-based architecture.

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