JOINT ROUTING, GATEWAY SELECTION, SCHEDULING AND POWER MANAGEMENT OPTIMIZATION IN WIRELESS MESH NETWORKS

A THESIS SUBMITTED TO THE DEPARTMENT OF INDUSTRIAL ENGINEERING AND THE GRADUATE SCHOOL OF ENGINEERING AND SCIENCE OF BILKENT UNIVERSITY IN PARTIAL FULFILLMENT OF THE REQUIREMENTS FOR THE DEGREE OF MASTER OF SCIENCE

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July, 2011

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ABSTRACT

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The third generation (3G) wireless communications technology delivers user traffic in a single step to the wired network via base station; therefore it requires all base stations to be connected to the wired network. On the other hand, in the fourth generation (4G) communication systems, it is planned to have the base stations set up so that they can deliver each other's traffic to a small number of base stations equipped with wired connections. In order to improve system resiliency against failures, a mesh structure is preferred.

The most important issue in Wireless Mesh Networks (WMN) is that the signals that are simultaneously transmitted on the same frequency channel can interfere with each other to become incomprehensible at the receiver end. It is possible to operate the links at different times or at different frequencies, but this also lowers capacity usage.

In this thesis, we tackle the planning problems of WMN, using 802.16 (Wi-MAX) protocol, such as deploying a given number of gateway nodes along with operational

problems such as routing, management of power used by nodes and scheduling while maximizing the minimum service level provided. The WMN under consideration has identical routers with fixed locations and the demand of each router is known. In order to be able to apply our results to real systems, we work with optimization models based on realistic assumptions such as physical interference and single path routing. We propose heuristic methods to obtain optimal or near optimal solutions in reasonable time. The models are applied to some cities in Istanbul and Ankara provinces.

Keywords: Wireless Mesh Networks, Integer Programming, Gateway Selection

ÖZET

ÇOKGEN BAĞLANTILI KABLOSUZ AĞLARDA ROTALAMA, AĞ GEÇİT DÜĞÜMÜ SEÇİMİ, ÇİZELGELEME VE GÜÇ KONTROLÜ PROBLEMİNİN BÜTÜNLEŞİK ENİYİLEMESİ

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Üçüncü nesil (3G) kablosuz haberleşme teknolojisi, kullanıcı trafiğini tek adımda baz istasyonu üzerinden kablolu şebekeye aldığı için bütün baz istasyonlarının kablolu şebekeye bağlı olmasını gerektirmektedir. Kurulum maliyetlerini düşürmek için, dördüncü nesil (4G) kablosuz haberleşme sistemlerinde her bir baz istasyonunun şebekeye kablo ile bağlanması yerine, birbirlerinin trafiğini şebekeye bağlı az sayıda baz istasyonuna ulaştırabilecek şekilde kurulmaları planlanmaktadır. Bozulmalara karşı sistem direncini arttırmak için çokgen bağlantılı bir yapı tercih edilmektedir.

Çokgen Bağlantılı Kablosuz Ağlarda (ÇBKAlarda) karşılaşılan en önemli problem, aynı frekans kanalında ve aynı anda gönderilen sinyallerin ortamda fiziksel olarak birbirleriyle etkileşime girerek alıcı tarafında anlamsız bir hal alabilmesidir. Bağlantıları farklı zamanlarda veya farklı kanallarda çalıştırarak etkileşim önlenebilmekte ancak bu denetimin niteliği, kaynak kullanımındaki verimliliği de etkilemektedir.

Bu çalışmada, 802.16 protokolü kullanan ÇBKAlarda belirli sayıda ağ geçit düğümünün yerlerinin seçimi gibi planlama problemlerinin yanısıra, verilen en kötü hizmeti en iyileme amacıyla, bu seçimden kaynaklanan rotalama, düğümler tarafından kullanılan güç kontrolü ve çizelgeleme gibi operasyonel problemlere birlikte bakılmaktadır. Ele alınan ÇBKA, yerleri belirli ve trafik miktarları bilinen özdeş düğümlerden oluşmaktadır. Sonuçların gerçek sistemlerde uygulanabilmesi için fiziksel etkileşim ve tek rotalı erişim gibi gerçekçi eniyileme modelleri ile çalışılmaktadır. Daha kısa sürede 'iyi' sonuç elde etmek amacıyla bazı sezgisel yöntemler geliştirilmiş ve modeller İstanbul ve Ankara'daki bazı bölgelere uygulanmıştır.

Anahtar Kelimeler: Çokgen Bağlantılı Kablosuz Ağlar, Tamsayılı Programlama, Ağ Geçidi Seçimi

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Chapter 1

Introduction

Wireless Networking Technologies are being used more frequently in recent years. After the Wireless Ad Hoc Networks, relatively new wireless technology, Wireless Mesh Networks (WMNs), have emerged as a cheap, easy to implement, efficient and reliable networking solution. As the number of users increased, the need for a better planned, faster, self-healing and flexible network also increased and WMNs propose an effective alternative for this need. Rather than delivering user traffic in a single step to the wired network, WMNs use a multi-hop structure to deliver user traffic. Hence, WMNs need less wired connection which decreases the deployment cost significantly. WMNs do not only lower the deployment cost but also lower the operational cost. It has been showed by M. Chee (2003) that using a mesh network to interconnect 133 existing hotspots in the Toronto downtown core will decrease the total cost of running hotspots by 70%.

A typical WMN consists of Mesh Clients (MCs), and Mesh Routers (MRs). A Mesh Client is actually a user, can be either mobile or stationary, trying to send/receive data to/from the Internet. Notebook users, smart phone users, PDA users are typical examples of MCs. MRs, on the other hand are static devices deployed in a deterministic way. Zhou et al., (2010) define MRs as powerful devices without constraints of energy, computing power, and memory and are usually distributed in a static and deterministic manner. Some of the MRs with special bridge functionalities, called gateways, play an important role within the WMNs. They connect WMNs to actual Internet with wire thus they are expensive to deploy.



Figure 1.1: Working Principle of a Mesh Router

The basic working principle of WMNs can be summarized as follows; as it can be seen in Figure 1.1, an MR gathers data from the MCs within its covering range and delivers it to the gateways. If the MR is within the Line-Of-Sight (LOS) transmission range of the gateway, the MR delivers its traffic through a single-hop link, however if the MR is not in the LOS transmission range, then it uses other MRs to reach the gateway and uses a multi-hop structure which increases the capacity usage in non-

LOS environments. According to Cao et al. (2006), the mesh topology not only extends the network coverage and increases capacity in non-LOS environments, but it also provides higher network reliability and availability when node or link failures occur, or when channel conditions are poor. A typical WMN can be seen in Figure 1.2.



Figure 1.2: A Typical WMN

According to Akyildiz et al. (2004), WMNs have a wide range of application areas, including broadband home networking, community and neighborhood networking, enterprise networking, metropolitan area networks (MAN), transportation systems, building automation, military communications and surveillance systems etc.

In addition to all these advantages, WMNs have some issues. The major issue that drew the attention of researches is the Interference Problem. Different from wired networks, when a transmission occurs in a wireless network, signals trace a circular pattern as shown in Figure 1.3. Since the transmission signals of the sender will also be received by other nodes within the transmission range, this situation will prevent them from receiving other signals. This is undesirable for wireless networks and the solution methods used to prevent interference decreases the network capacity significantly. A simple illustration of interfering signals can be seen in Figure 1.4.



Figure 1.3: Wired Transmission vs. Wireless Transmission



Figure 1.4: Interfering Signals

In Figure 1.4, MR B is in the transmission range of both MR A and MR C. The red and blue circles show the transmission signals of MR A and MR C, respectively. Suppose MR A is trying to send data to MR B and MR C is trying to send data to MR D at the same time, using the same frequency. MR B will not only receive the signals sent from MR A to itself but also the signals sent from MR C to MR D. Both of the signals will interfere with each other and become meaningless at the MR B and the transmission from MR A to MR B will fail and transmitted data will be lost.

In the literature, there are two main interference models; protocol model and physical model as defined by Gupta and Kumar (2000).

1. The Protocol Model: Suppose node *i* located at X_i transmits over the *b*th subchannel to a node *j* located at X_j . Then this transmission is successfully received by node *j* if

$$\left|X_{k} - X_{j}\right| \ge (1 + \Delta)\left|X_{i} - X_{j}\right| \tag{1.1}$$

for every other node k simultaneously transmitting over the same subchannel. The quantity $\Delta > 0$ models the situation where a guard zone is specified by the protocol to prevent a neighboring node from transmitting on the same subchannel at the same time.

2. The Physical Model: Let S be the set of nodes that are simultaneously transmitting at some time over a certain subchannel and let P_i be the power level used by node *i*. Then the transmission between node *i* and node *j* is successful if

$$\frac{\frac{P_i}{\left|X_i - X_j\right|^{\alpha}}}{\gamma + \sum_{\substack{k \in S \\ k \neq i}} \frac{P_k}{\left|X_k - X_j\right|^{\alpha}}} \ge \theta$$
(1.2)

where θ is the corresponding signal-to-interference-noise-ratio (SINR), α is the path loss exponent and γ is the ambient noise power level. Since the signal strength between two nodes decreases as the distance between these nodes increases, we can calculate the loss between node *i* and node *j*, l_{ij} , as follows;

$$l_{ij} = \frac{1}{|X_i - X_j|^{\alpha}}$$
(1.3)

In the end, the physical interference model can be viewed as;

$$\frac{P_i l_{ij}}{\gamma + \sum_{\substack{k \in S \\ k \neq i}} P_k l_{kj}} \ge \theta \tag{1.4}$$

The physical interference model is much more realistic than the protocol interference model as it considers the effect of all other transmissions occurring simultaneously on the same subchannel. To successfully transmit the data, the MRs have to use enough power to reach beyond a certain threshold SINR value. Suppose node i transmits data to node j in a transmission slot and also there are some other transmissions occurring in the same transmission slot. For node j to successfully receive the data of i, the signal to interference ratio at the node j has to be over the certain threshold value. Thus for a successful transmission, the environment noise ratio and the effect of other ongoing transmissions in the same slot on node j should not be high to effect the quality of the transmission negatively.

Since, interference is the major issue, many schemes and channel access methods are proposed to carefully control the interference for multiple users while trying to use the available capacity efficiently. According to Kumar et al. (2006), these schemes can be classified into two: contention based and non-contention based. In the contention based schemes, no controller is needed and each terminal transmits data in a decentralized way. ALOHA and Carrier Sensing Multiple Access (CSMA) are typical examples of contention based schemes. In the non-contention based schemes on the other hand, a logic controller is needed. Time Division Multiple Access (TDMA), Frequency Division Multiple Access (FDMA) and Orthogonal Frequency Division Multiple Access (OFDMA) are among these multiple access schemes. According to Garcia and Widjaja (2006), in TDMA, users all transmit using the same frequency and yet, their signals do not interfere with each others because different users transmit at different time slots as depicted in Figure 1.5. In TDMA, the time is divided into time slots that are assigned to the different users such that all users are using the same frequency band for transmission. Suppose there are N users sharing a single frequency band for transmission. In each slot, a user transmits data one after another and after the Nth user, the process is repeated again and again. The period in which all users are assigned one transmission time slot is called a frame, and it contains N time slots as shown in Figure 1.6.



Figure 1.5: Illustration of TDMA Scheme



Figure 1.6: Frame Structure in TDMA

FDMA on the other hand, uses non-overlapping frequency bands for each user as depicted in Figure 1.7. Different than TDMA, users can transmit all the time.



Figure 1.7: Illustration of FDMA Scheme

Finally, OFDMA which is a combination of both TDMA and FDMA divides the time into equal frames and allows transmissions using different frequencies as depicted in Figure 1.8.



Figure 1.8: Illustration of OFDMA Scheme

In these schemes, only one transmission can occur in a certain transmission slot however spatial reuse allows more than one transmission to occur in the same transmission slot if the SINR threshold value for these transmissions is satisfied. By using spatial reuse, available capacity is used more efficiently.

There are protocols for wireless networking defined by Institute of Electrical and Electronics Engineering (IEEE) to control the operations like routing, media access control etc. These protocols are 802.11 Wireless Fidelity (Wi-Fi) and 802.16 Worldwide Interoperability for Microwave Access (Wi-Max).

802.11 protocol, which was originally defined for Wireless Local Area Network (WLAN), uses 2.4 GHz or 5 GHz unlicensed radio bands. 802.11a, 802.11b and 802.11g protocols are the most popular Wi-Fi protocols used. The data transmission rate for this technology differs between 11-54 Megabits per second (Mbps). The most common application areas of Wi-Fi are buildings, campuses, airports etc. Although 802.11 technology offers great advantages in terms of wireless networking, it is a decade old and was not designed for mesh networks (Djukic and Valaee, 2007)

802.16 Wi-MAX protocol on the other hand is designed for mesh technology and uses 2-11 GHz and 10-66 GHz radio bands. The 802.16, 802.16a and 802.16e are the most popular Wi-Max protocols. The original version of the standard was released in December 2001 and addressed systems operating in the 10-66 GHz frequency band. However, this system needed LOS environment which increases the deployment cost significantly. The 802.16a technology uses in 2-11 GHz band and operates in non-LOS environment. The data transmission rate can be increased up to 100 Mbps. Different than Wi-Fi, systems using Wi-MAX can be used to cover wider areas and the applications of the technology are cities, metropolitan regions etc.

In this thesis, we are jointly considering the Gateway Selection, Interference, Routing, Fairness, Scheduling, Power Management and Throughput problem of WMNs using 802.16 Wi-Max protocol. In our study Wi-MAX protocol is used since it offers great advantages in terms of providing next generation wireless networking solutions and has Metropolitan Area Network (MAN) applications.

In the literature, there are many studies addressing different problems in wireless mesh networks, however most of them do not consider these problems jointly. Since these problems affect each other, to provide an effective solution joint consideration of these aspects is crucial. Using adjustable power level is also another contribution of the thesis. By controlling the power usage ratio in each transmission slot, more effective use of the network capacity is provided.

The remainder of the thesis is organized as follows:

In Chapter 2, considered problems will be outlined and relevant studies in the literature will be given. In Chapter 3, a Mixed Integer Linear Programming (MILP) model will be proposed to obtain exact solutions for the problems defined in the Chapter 2. The proposed model picks the predefined number of gateways among the placed MRs, finds a tree structured routing to the gateway nodes, assigns transmission slots to each MR and determines the power usage ratio of each node while maximizing the minimum service level. In Chapter 4, a special version of p-median problem is proposed to obtain 'good' solutions for larger networks in reasonable time. The heuristic will also be tested using numerical examples in terms of running time and quality. In Chapter 5, some real life applications will be given and the proposed system will be applied to some chosen areas in major cities of Turkey. In Chapter 6, some extensions will be provided and a MILP model which uses flexible path routing will be proposed. Finally in Chapter 7, we will conclude the thesis by briefly summarizing our efforts and contributions of the thesis. We will also present future research opportunities in this chapter.

Chapter 2

Problems in WMNs and Problem Definition

2.1 Problems in WMNs

Although WMNs offers great advantages in terms of providing fast and reliable networking solution, WMNs have some deployment and operational problems. These problems of WMNs need a careful and logical planning. The major problems that drew the attention of researchers are Gateway Selection Problem, Routing Problem, Scheduling Problem, Covering Problem, Channel Assignment Problem, Clustering Problem, Fairness and Power Management Problem.

Gateway selection problem (GSP) is one of the most commonly studied problems of WMNs in the literature. It is the problem of determining the MRs with bridge functionalities among the placed ones. Different than a usual MR, gateways are much more expensive and complex devices. As they are connected to the actual network with wire, deployment of these devices are also costly. Since the gateway

selection affects the throughput, routing, scheduling and capacity usage; these aspects of WMNs should be jointly considered.

The GSP is shown to be NP hard by He et al. (2006) by using a reduction from the Capacitated Facility Location Problem (CFLP). In this study, a mixed integer linear program with two objectives is proposed. One objective is to use minimum number of gateways and the other objective is to use minimum number of hops for each MR to reach the gateway. Also, two heuristic approaches are proposed and the efficiency of heuristics is tested using simulation technique. However, in this study interference, which is the most challenging issue in WMNs, and scheduling are not considered.

Another research on Gateway Placement is done by Zhou et al. (2010). An innovative gateway placement scheme is proposed which considers the number of MRs, MCs and gateways. Protocol interference model is used and the throughput performance has been tried to be enhanced. Two throughput metrics are defined with this manner: maximizing aggregate throughput of WMNs and maximizing the worst-case-per client throughput in the WMN.

In addition, Routing Problem in WMNs is to find the paths that will carry the data packets to the selected gateway while considering the interference. There are also alternative methods used for routing MR traffic to the gateway. Using a tree structure rooted at gateway node, using multiple paths to carry data by dividing it and using flexible paths are among these methods. The Figure 1.5 depicts an illustration of these methods. In this figure, gateway node is depicted as black node with number 1. In the first picture, traffic division is not allowed and the final routing is a tree structure rooted at gateway. In the second picture, traffic division is allowed and data proportions follow different paths to reach the gateway. The traffic of 7th node is divided into two proportions. One proportion follows 7-4-2-1 to reach the gateway while other uses 7-8-9-5-3-1. Finally in the third picture, without allowing traffic

division, nodes are allowed to follow flexible paths. The 4th node gathers the data of 6^{th} and 7th node. Although it sends the traffic of 7th node and itself by using 4-2-1, it is allowed to send the traffic of 6^{th} node via another path, 4-5-3-1.



Figure 2.1: Different Routing Methods

As mentioned before, routing the traffic of a node to a gateway also assigns the node to that gateway. If there are more than one gateway, then the MRs will be clustered so that each cluster has a gateway and MRs in each cluster are routed to that specific gateway. To satisfy the desired objective function, these clusters should also be formed in a logical manner. Figure 1.6 illustrates these clusters.

Aoun et al. (2006) have proposed an Integer Linear Programming (ILP) formulation to cluster MRs in such a way that the maximum number of hops in each cluster is bounded above by R while ensuring relay load and cluster size constraints. To cope with larger networks, an algorithm, based on recursively computing minimum Dominating Sets has been proposed. However, neither the traffic demand of each router nor the interference has been taken into account which are among the most crucial aspects of the WMNs.



Figure 2.2: An Illustration of Clusters

Locating the MRs to cover all the MCs is also another problem in WMNs. Since an MR can cover all the MCs within its transmission range, MRs should be located in a logical manner to cover the MCs and provide service.

Channel assignment problem is also an issue for radios using multiple channels. Determining which transmission to use which channel while considering interference has also been another hot topic for researchers.

Although power consumption is not a significant problem for WMNs, as the power used by MRs affect both the transmission rate and the interference caused on other nodes, it is also quite crucial to determine power levels used by MRs. Power levels should not only be adjusted to provide connectivity within the network, but also transmit minimum interference to the environment. If the power levels are too low, this will cause a disconnected network, whereas if they are too high, this will cause extra interference to the environment and prevent effective capacity usage. The problem of determining the minimum range between two nodes for a disk area of *A* with n nodes is studied by Gupta and Kumar (1998). This study is further developed by Kawadia et al. (2001) and an algorithm is proposed to find minimum power level for a connected graph.

Delivering the MCs a fair service is yet another aspect of WMNs. One cannot let a certain MR use the whole capacity while restricting others. Thus, defining a fairness criterion and trying to satisfy it is quite crucial. According to Pioro et al. (2004), an intuitive way to approach this problem is to assign as much volume as possible to each demand, at the same time keeping the assigned volumes as equal as possible. This intuitive requirement leads to the assignment principle called Max-Min Fairness (MMF), known also as equity or justice in other applications. In our problem, we are using the same criterion and trying to maximize the minimum provided service level to each MR.

In our problem, similar to Cao et al. (2006) we have focused on the centralized scheduling of the IEEE 802.16 mesh mode, where a scheduling tree rooted at the gateway node is constructed for the routing path between each MR and the gateway, and the gateway acts as the centralized scheduler that determines the transmission or reception of every MR in each minislot.

2.2 Problem Definition

In our problem, we are given the number and the locations of mesh routers, and our aim is to maximize minimum service level using a predefined number of gateways and to use the available capacity as much as possible.

Each mesh router gathers data from the mesh clients within its transmission range and tries to deliver gathered data to the mesh routers with gateway functionalities to reach the Internet. In this problem, only uplink traffic (from MRs to the Internet) is considered. It is supposed that each router is equipped with two radio interfaces and one radio is used to control the local traffic (from MCs to MRs) while the other one is used to control the backbone traffic (among MCs) as in Zhou et al. (2010). Since we are not dealing with the local traffic, we assume each MC is served by the closest MR.

As only one radio is used for the transmissions between two MRs, a node can only transmit or receive in a single slot and cannot do both simultaneously. Transmission between two MRs occurs if the SINR threshold for communication is satisfied. As the location of each MR is known, one can easily determine the propagation loss between each node pair. This determination helps us to understand whether a node pair can communicate with each other or not.

Traffic division is not allowed in our problem as it creates some problems for the users. Especially for the Voice-Over-IP (VoIP) applications, receiving data simultaneously is a quite significant performance measure. In case of traffic division, the variation of arrival intervals for each packet may vary causing some problems for the user. Thus, traffic division is not allowed.

The network under consideration consists of nodes and arcs. Each MR is represented by the nodes in the network with predefined locations. As mentioned before, these MRs are static and each has demands to be delivered to the gateway node. Transmissions between two nodes are represented by edges. Suppose node *i* can transmit data with a maximum power defined by P_i^{max} and the propagation loss between node *i* and node *j* is defined by l_{ij} . There exists an arc between node *i* and node *j* if $P_i^{max} l_{ij} \ge \theta$ where θ is the SINR threshold value. For simplicity, we define $g_{ij} = P_i^{max} l_{ij}$.

As mentioned before, maximizing the minimum service level is one of the most crucial aspects of our study. We define a service ratio and aim to satisfy all the MRs with the same service ratio to provide fairness among MRs. Suppose d_i is the amount of flow that node *i* has to deliver to the gateway node and *m* is the ratio defined as minimum capacity allocated to node *i* / the demand of node *i*. In this study, md_i is the minimum capacity allocated to node *i*. Our aim is to provide a service to all customers while maximizing the defined service level.

The most related work to our study is proposed by Targon et al. (2010) which is an extension of the GSP including joint routing and scheduling aspects. Similar to our assumption, traffic division is not allowed in this study. An ILP formulation is presented which tries to minimize gateway deployment cost. The model finds the location of the gateway nodes along with routes from each MR to the found gateways and a schedule for each MR. Although most of our assumptions are similar to those, our model is an extended version of this study. Rather than trying to find a schedule and routing for the given traffic demands of each MR, we are trying to provide a service level to each MR and trying to maximize the minimum service level. Also, in our study we have power control option so that transmitting MRs can adjust their power level to increase the use of the available capacity. In addition, we use physical interference model which is more realistic than the interference model used in this study. Rather than trying to deploy a given number of gateways, this study involves with finding minimum gateway deployment cost to provide service.

Chapter 3

Model Formulation

In this chapter, we will formulate a Mixed Integer Linear Programs (MILP) to solve joint scheduling, routing, gateway selection and power control problem, using physical interference model in OFDMA based single rate wireless mesh networks while maximizing minimum service level provided.

In the first part of the chapter, we will define the model *WMN*1, which uses the treerouted structure and then add some valid inequalities and upper bounds to improve the running time of the model.

3.1. MILP Formulation

3.1.1 Assumptions

The assumptions listed below are used to model the problem defined earlier

- WMN under consideration uses 802.16 Wi-MAX standards
- OFDMA is used as multiple access scheme
- There is a finite number of MRs operating
- All the MRs used are identical
- All the locations of these MRs are known
- Multi rate transmission is not allowed
- Each MR is equipped with two radio interfaces
 - One for the local communications (among MCs and MRs)
 - One for backbone communications (among MRs)
- Traffic division is not allowed
- Tree-routed structure allowed
- Each MC is served by the closest MR
- Omni-directional antennas are used
- Ambient noise power is static

3.1.2 Sets and Parameters

To model the problem, following sets and parameters are used;

- N denotes the set of Nodes i.e. MRs where $N = \{1, ..., N\}$

- *A* denotes the set of Arcs i.e. available transmissions. We can say that the directed link $(i, j) \in A$ if $g_{ij} = P_i^{max} l_{ij} \ge \theta$ and the propagation loss is calculated as follows;

$$l_{ij} = \frac{1}{\left|x_i - x_j\right|^{\alpha}}$$

where α is the path loss exponent and $|x_i - x_j|$ is the distance between node *i* and node *j*.

-T denotes the number of non-interfering transmission slots. This can be either transmissions being active in different time slots or transmissions using non-interfering frequencies in the same slot.

-c denotes the capacity of an existing link. If two nodes can communicate with each other, than they can send data with a rate defined as c.

-G denotes the number of gateways to be deployed.

-a denotes the capacity vector of a gateway. If a gateway is placed at a node, then it will have a capacity equal to a.

 $-d = (d_1, ..., d_N)$ denotes the traffic vector of the MRs in the WMN. A node *i* collects the traffic of all MCs within its coverage range and tries to deliver it to a gateway.

 $-g = [g_{ij}]$ denotes the signal strength of node *i* at node *j* while node *i* transmits data using maximum power available.

 $-\gamma$ denotes the noise value in the environment.

 $-\theta$ denotes the corresponding SINR threshold value required for a successful transmission

3.1.3 Mixed Integer Linear Programming Model

3.1.3.1 Variables

The following variables are needed to model the problem.

$x_{ij}^t = \begin{cases} 1, \\ 0, \end{cases}$	if the link (i, j) is active in transmission slot t otherwise	
$y_i = \begin{cases} 1, \\ 0, \end{cases}$	if node i is used as a gateway otherwise	
$z_{ij} = \begin{cases} 1, \\ 0, \end{cases}$	if the link (i,j) is used for transmission otherwise	
$f_{ij} = the modelship modelship$	aximum amount of flow carried on link(i,j)	
fout _i = the maximum amount of flow leaving system with gateway i		
$fin_i = the c$	amount of flow originating from the clients of node i	
$P_i^t = the \ ra$	tio of the transmitted power from node i in transmission slot t	
$m = minimum \ service \ level$		

3.1.3.2 Formulation of MILP

The following MILP, WMN1, is constructed to solve the problem defined earlier. The model jointly considers the gateway location, scheduling, routing and power control aspects of the problem using a tree routing structure. The aim is to maximize the minimum service level provided to each customer and use the capacity of the network efficiently. WMN1

max m

s.t.

$$f_{ij} \leq \frac{c}{T} \sum_{t=1}^{T} x_{ij}^{t} \qquad \forall (i,j) \in A \qquad (3.1)$$

$$fin_i = md_i \qquad \forall i \in N \tag{3.2}$$

$$fout_i \le ay_i \qquad \forall i \in N \tag{3.3}$$

$$\sum_{i=1}^{N} y_i = G \tag{3.4}$$

$$fin_i + \sum_{s:(s,i)\in A} f_{si} = fout_i + \sum_{k:(i,k)\in A} f_{id} \qquad \forall i \in N$$
(3.5)

$$\sum_{j:(i,j)\in A} z_{ij} = 1 - y_i \qquad \forall i \in N$$
(3.6)

$$x_{ij}^t \le z_{ij} \qquad \forall (i,j) \in A, t \in T$$
(3.7)

$$\sum_{s:(s,i)\in A} x_{si}^t + \sum_{k:(i,k)\in A} x_{id}^t \le 1 \qquad \forall i \in N, t \in T$$
(3.8)

$$P_i^t \le 1 \qquad \forall i \in N, t \in T \tag{3.9}$$

$$x_{ij}^{t} \leq \begin{cases} 1, & \text{if } \frac{P_{i}^{t}g_{ij}}{\gamma + \sum_{s=1, s\neq i, j}^{N} P_{s}^{t}g_{sj}} \geq \theta \\ 0, & \text{otherwise} \end{cases} \qquad (i,j) \in A, t \in T \qquad (3.10)$$

$$x_{ij}^t \in \{0,1\} \qquad \forall (i,j) \in A, t \in T$$
(3.11)

$$y_i \in \{0,1\} \qquad \forall i \in N \tag{3.12}$$

$$z_{ij} \in \{0,1\} \qquad \forall (i,j) \in A \tag{3.13}$$

$$f_{ij} \ge 0 \qquad \forall (i,j) \in A \tag{3.14}$$

$$f_{out_i} \ge 0 \qquad \forall i \in N \tag{3.15}$$

$$fout_i \ge 0 \qquad \forall i \in N \tag{3.15}$$
$$fin_i \ge 0 \qquad \forall i \in N \tag{3.16}$$

$$fin_i \ge 0 \qquad \forall i \in N \tag{3.16}$$

$$P_i^t \ge 0 \qquad \forall i \in N, t \in T$$

$$m \ge 0 \qquad (3.17)$$
Our objective is to maximize the minimum service level provided to all customers while ensuring the constraints.

Constraint (3.1) is the capacity constraint. If a link is active in a period *t* then it has $\frac{c}{T}$ available capacity to transmit the data in this period. Thus, the traffic flow on a link cannot be more than the available capacity.

Constraint (3.2) ensures that the actual traffic given to the system by node i equals to the provided service level times the traffic of that node. Similarly, constraint (3.3) is added to make sure that the amount of traffic flow using node i as gateway cannot be larger than the gateway capacity of that node. Constraint (3.4) implies that the total number of gateways to be deployed is equal to G.

(3.5) is the flow conversation constraint. The traffic flow from other nodes reaching node i and the traffic flow of node i equals to the leaving traffic flow from node i. Constraint (3.6) ensures that the traffic of each node has to follow a single route to reach the gateway.

Constraint (3.7) implies that there can be a transmission between node i and node j if and only if the link under consideration is used. Since each MR is equipped with a single radio used for transmission, then in a single transmission slot, it can only send or receive data and it cannot do them simultaneously. The single radio issue is handled by (3.8).

Constraint (3.9) is the power usage constraint. The power ratio used by node i in period t cannot be larger than 1.

The most crucial and restrictive constraint of the model is the interference constraint. As mentioned earlier, different than most of the studies, a more realistic physical interference model is used in our model. Constraint (3.10) is the physical interference constraint.

Non-linear constraint (3.10), is linearized as follows;

N 7

$$P_i^t g_{ij} + M_{ij} \left(1 - x_{ij}^t \right) \ge \theta \gamma + \sum_{s=1, s \neq i, j}^N \theta P_s^t g_{sj} \qquad \forall (i, j) \in A, t \in T \quad (3.10a)$$

where M_{ij} is a sufficiently large, positive number.

For the linearized constraint (3.10*a*), if $x_{ij}^t = 1$, then this means node *i* transmits data to node *j* in transmission slot *t* and the constraint becomes;

 $P_i^t g_{ij} \ge \theta \gamma + \sum_{s=1, s \ne i, j}^N \theta P_s^t g_{sj}$ which means the minimum SINR value for this transmission have to be greater than θ plus any other transmissions occurring in this time slot *t*.

If there is no transmission from node *i* to node *j*, i.e. $x_{ij}^t = 0$, then the constraint becomes; $P_i^t g_{ij} + M_{ij} \ge \theta \gamma + \sum_{s=1, s \ne i, j}^N \theta P_s^t g_{sj}$ which becomes redundant because of the choice of M_{ij} .

 M_{ij} used in this constraint is calculated as follows;

$$M_{ij} = \theta \gamma + \sum_{\substack{s=1\\s \neq i,j}}^{N} \theta g_{sj}$$
(3.19)

which is always greater than the right hand side (RHS) of the constraint to make it redundant when $x_{ij}^t = 0$.

Finally, the constraints (3.11), (3.12) and (3.13) imply that these variables are binary and similarly (3.14), (3.15), (3.16) and (3.17) are added to define positive variables.

In the end WMN1 becomes;

max m

s.t. (3.1) – (3.9), (3.10*a*), (3.11) – (3.18).

3.1.3.3 Additional Constraints for WMN1

The following additional constraints are added to the WMN1 to narrow down the solution space and help the solver find the optimal solution faster, without changing the final solution. These kinds of inequalities are called as valid inequalities. According to Cornuejols (2006), an inequality is said to be valid for a set if it is satisfied by every point in this set and a cut with respect to a point $(\bar{x}, \bar{y}) \notin conv(S)$ is a valid inequality for conv(S) that is violated by (\bar{x}, \bar{y}) where conv(S) is the convex hull of S.

Although the valid inequalities may narrow down the solution space, they do not always improve the running time of the solver. There is a tradeoff between adding valid inequalities and the number of constraints. As the number of valid inequalities increases, the solution space will narrow down however as the number of constraints increases, the model will contain too many constraints which may cause clumsiness in terms of the running time of the model. An illustration of unnecessarily added valid constraints is given in Figure 3.1.

In the figure below, ABCD defines a polyhedra and the red point is the optimal integer solution. The valid inequalities are added and they narrowed the solution space down, however they do not cut off the fractional solutions around the optimal solution and thus do not help the solver with the running time. Adversely, they may increase the number of constraints and may increase the running time of the model.



Figure 3.1: An Illustration of Unnecessarily Added Constraints

Proposition 1. The following inequality is valid for the polyhedra defined by the constraints of WMN1.

$$\sum_{t=1}^{T} x_{ij}^{t} \ge z_{ij} \qquad \forall (i,j) \in A$$
(3.20)

Proof. Suppose $z_{ij} = 1$, which means that the link $(i, j) \in A$ is used for transmission. Then (3.20) becomes $\sum_{t=1}^{T} x_{ij}^t \ge 1$ which implies the link $(i, j) \in A$ has to be active in at least one of the transmission slots which is what we desire.

Similarly, suppose $z_{ij} = 0$. Then (3.20) becomes $\sum_{t=1}^{T} x_{ij}^t \ge 0$ which is valid because the constraint (3.7) forces no link to be active in any of the transmission slots. Thus $\sum_{t=1}^{T} x_{ij}^t = 0$, implies. Hence the constraint (3.20) is valid for the polyhedra defined by the constraints of *WMN*1.

The effect of the added valid inequality can be seen in Appendix B.

3.1.3.4 Improving the Upper Bound of WMN1

In addition to the added constraints, adding some lower and upper bounds on the objective value may improve the running time of the solver. The basic working principle of a typical MILP solver can be summarized as follows; the solver tries to find feasible solutions using heuristic methods. For a maximization problem, each feasible solution found is a lower bound on the objective value and conversely, the relaxation of the model yields an upper bound on the objective value. The solver terminates when the upper bound and the lower bound get the same value.

With this point of view the following upper bounds are added to the model;

Proposition 2. The following inequality is an upper bound on the objective value, m^* , of the WMN1.

$$m \le \frac{\sum_{i=1}^{N} a y_i}{\sum_{i=1}^{N} d_i} \tag{3.21}$$

Proof. The constraint (3.21) uses the available gateway capacity as an upper bound on the provided service level. Since each MR is provided with the same service level *m*, and all the traffic uses the gateways to reach the Internet.

Suppose (3.21) is not an upper bound on m^* , i.e. $m^* > \frac{\sum_{i=1}^N ay_i}{\sum_{i=1}^N d_i}$, then the total throughput of the system is $m^* \sum_{i=1}^N d_i$. However, since the throughput is bigger than the total gateway capacity, $m^* \sum_{i=1}^N d_i > \sum_{i=1}^N ay_i$, such an m^* is not feasible for the system, Thus (3.21) is an upper bound for the optimal value of WMN1.

Proposition 3. The following inequality is an upper bound on the objective value, m^* , of the WMN1.

$$m \le \frac{Gc}{k(G)} \tag{3.22}$$

where $k(G) = \sum_{i=1}^{N} d_i - \sum_{i=1}^{G} d'_i$ and $d' = (d_{(1)}, d_{(2)}, \dots, d_{(N)})$ such that $d_{(1)} \ge d_{(2)} \ge \dots \ge d_{(N)}$.

Proof. The constraint (3.22) uses the available link capacity as an upper bound on the provided service level. Since $G \in \{1, 2, ..., N\}$, and the traffic of the nodes with gateway functionalities will directly reach the Internet without using any links, then the traffic to be transmitted is $\sum_{i=1}^{N} d_i - \sum_{i \in G} d_i$. The available link capacity for the transmission of these data is at most Gc. In each transmission slot, the links have $\frac{c}{T}$ available capacity. Since all the MRs have single radio, a gateway can only receive data from one node in a transmission slot, the maximum number of transmissions that a gateway can have equal to T. Thus, for a gateway the maximum link capacity is $\frac{c}{T}T = c$ and since there are G gateways, the total available link capacity to transmit data is G * c.

Thus, because of the link capacity constraint, the service level provided can be at most; $\frac{Gc}{\sum_{i=1}^{N} d_i - \sum_{i \in G} d_i}$ even the traffic division is allowed. Since we do not know

which nodes will be chosen as gateway nodes, we can write;

$$m \le \frac{Gc}{\sum_{i=1}^{N} d_i - \sum_{i \in G} d_i} \le \frac{Gc}{\sum_{i=1}^{N} d_i - \sum_{i=1}^{G} d_i'} = \frac{Gc}{k(G)}$$
(3.23)

Since $\sum_{i \in G} d_i \leq \sum_{i=1}^G d'_i$, then $\sum_{i=1}^N d_i - \sum_{i \in G} d_i \geq \sum_{i=1}^N d_i - \sum_{i=1}^G d'_i$ and the inequality (3.23) holds.

After adding the valid inequalities and constraints on the objective value, WMN1 becomes;

max m

s.t. (3.1) - (3.9), (3.10a), (3.11) - (3.18), (3.20) - (3.22).

The effect of these bounds on the objective value is shown in the Appendix C by using LP relaxation.

In the end, the number of variables and constraints that *WMN*1 has is listed in the following table.

# of binary variables	# of other variables	#of constraints	
N+A(T+1) N(T+2)+A+1		N(2T+4)+A(2T+2)+3	

Table 3.1: Number of Variables and Constraints of WMN1

In addition to adding valid inequalities and improving the upper bound, we needed to relax some binary variables and try to figure out the effects of these relaxations in the remaining part of the chapter.

3.2 LP Relaxation

*WMN*1 has three sets of binary variables x_{ij}^t, z_{ij} and y_i . By relaxing each combination of these sets, we get following sub models shown in Table 3.2.

MODEL	EXPLANATION
WMN1	None of the variables are relaxed
WMN1_x	x_{ij}^t relaxed, z_{ij} and y_i are binary
WMN1_y	y_i relaxed, z_{ij} and x_{ij}^t are binary
WMN1_z	z_{ij} relaxed, x_{ij}^t and y_i are binary
WMN1_xy	x_{ij}^t and y_i are relaxed, z_{ij} binary
WMN1_xz	x_{ij}^t and z_{ij} are relaxed, y_i binary
WMN1_yz	y_i and z_{ij} are relaxed, x_{ij}^t binary
WMN1_xyz	All the variables are relaxed

Table 3.2: The Names of the Relaxed Models

After running the models separately, we see that $WMN1, WMN1_y, WMN1_z$ and $WMN1_yz$ yield the same solution even though some of the binary variables are relaxed. In spite of relaxing the variables, the constraints on the model still forces them to get binary values. A comparison of these models is given in the Appendix D in terms of running time. According to that comparison, $WMN1_z$ performs better in terms of running time than the others thus it is used in this study to solve the defined problem. From now on, whenever we say $WMN1_z$.

It is also seen that both $WMN1_x$ and $WMN1_xy$ yield the same solution. Even though we additionally relax y_i in $WMN1_xy$, the constraints force it to have binary values. In these models, as x_{ij}^t is relaxed, the interference constraint no longer works. Although interference is not considered by these models, an advantage of these models is that, they pick the gateways and determine the single routes from each MR to these gateways.

WMN1_xz, on the other hand, picks a gateway and allows multipath routing without considering the interference. Since the constraint (1.23) is found by the same manner, there is no doubt that the solution of *WMN1_xz* will be equal to $\frac{Gc}{k(G)}$.

Since the model, *WMN*1, has too many binary variables, it takes excessive amount of time to obtain the optimal solution as the network size increases. Thus, we needed to devise some solution methods to obtain 'good' solutions in reasonable time for larger networks. The running time and the objective value of each run can be seen in Appendix E.

Chapter 4

Solution Methods

In this chapter, we will define some basic methods to solve the problem defined in Chapter 2, for larger networks. Since the problem we are dealing with is a joint consideration of more than one NP hard problem, the running time of the solver increases exponentially as the input size increases.

In our study, we have seen that pre-defining the gateway nodes for the WMN, decreases the running time as it cuts off the feasible solution space. However, we may lose the chance of finding the optimal solution. So, there is a tradeoff between obtaining a solution in reasonable time by defining the gateways and finding the optimal solution in excessive amount of time.

With this manner, we have tried to determine the gateway node in a logical manner to obtain a near-optimal solution in reasonable time. The problem is simplified by reformulating it as a special case of p-median problem and some additional constraints are added. Then the solution of this formulation is used in *WMN*1 to obtain a solution in reasonable time.

4.1 Heuristic Approach

The p-median problem, defined by Hakimi (1964), consists of locating p facilities on a network, so that the sum of the shortest demand weighted distances from each of the nodes to its nearest facility is minimized. When we start considering the capacities of the facilities, the p-median problem becomes the capacitated p-median problem (CPMP) which is defined by Maniezzo et al. (1998) and can be modeled as:

CPMP

$$\min\sum_{i=1}^{N}\sum_{j=1}^{N}q_{i}h_{ij}r_{ij}$$

s.t.

$$\sum_{j=1}^{N} r_{ij} = 1 \qquad \forall i \in N$$
(4.1)

$$\sum_{j=1}^{N} r_{jj} = \nu \tag{4.2}$$

$$\sum_{i=1}^{N} q_i r_{ij} \le Q x_{jj} \qquad \forall j \in N$$
(4.3)

$$r_{ij} \in \{0,1\}$$
 (4.4)

where h_{ij} is the distance from node *i* to node *j*, *v* is the number of facilities to be deployed, q_i is the demand of node *i* and finally *Q* is the capacity of a facility. The binary variable r_{ij} equals to 1 if the demand of node *i* is satisfied by facility *j* and 0 otherwise.

Constraint (4.1) implies that every node will be assigned to a facility. (4.2) implies that the number of facilities to be opened is exactly *v*. Constraint (4.3) on the other hand is the capacity constraint. The demands assigned to a facility cannot be larger than the capacity of that facility.

By making the following changes we can easily get a solution for our problem: $h_{ij} = the \ minimum \ number \ of \ hops \ needed \ to \ be \ traversed \ from \ node \ i \ to \ j$ $v = the \ number \ of \ gateways \ to \ be \ deployed \ in \ the \ WMN$ $q_i = the \ traffic \ demand \ of \ each \ node$ $Q = the \ capacity \ of \ a \ deployed \ gateway$

We also add following constraints to *CPMP* and obtain *CPMP*'. The additional constraints are;

$$\sum_{i=1,i\neq j}^{N} q_i r_{ij} \le c x_{jj} \qquad \forall j \in N$$

$$h_{ij} r_{ij} \le H \qquad \forall i, j \in N$$

$$(4.5)$$

where *c* is the link capacity and *H* is the hop count value.

Constraint (4.5) ensures that the total demand assigned to a gateway node cannot be larger than the total link capacity and (4.6) implies that the number of hops that each MR uses is bounded by H. In this approach, H is the minimum number that satisfies a feasible assignment for the *CPMP*'.

By solving *CPMP'*, we are trying to give weighted penalties to each node. The penalty given to a node can be calculated as number of nodes traversed times the traffic demand of that node which is $h_{ij}q_i$ as shown in Figure 4.1.



Figure 4.1: Assigning penalties

In the figure above, 6^{th} node is the gateway node and the traffic of node 1 is carried along the path (1-2-4-6) where it uses 3 hops to reach the gateway. Thus the penalty of node 1 is $3q_1$ where q_1 is the traffic demand of the 1st node. The overall penalty of the system above is $3q_1 + 2q_2 + 3q_3 + 1q_4 + 1q_5 + 0q_6$. Since we want to minimize the total penalty for all nodes, we are using the objective function of *CPMP* and trying to find the deployment with minimum penalty. The model tends to pick nodes with high traffic demands as gateway nodes.

In the latter case, we solve the *WMN*1 by fixing the gateways found by *CPMP*. To easily express;

Heuristic

- Step 1. Solve *CPMP'* and obtain the gateway set
- Step 2. Solve *WMN*1 by fixing the gateway set found in the step 1.

In Appendix G, our heuristic is compared with the heuristic approach proposed by Zhou et al. and a basic approach defined as picking the nodes with highest traffic demand as gateway nodes. The red colored lines show that the chosen heuristic performs better than the others.

Zhou et al. define a metric R_g , called as the gateway radius. The R_g is calculated as;

$$R_g = round\left(\frac{\sqrt{N_r}}{2\sqrt{N_g}}\right)$$

where N_r and N_g are the number of mesh routers and the number of gateways in the network, respectively. For each node in the system, Multihop Traffic-Flow Weight (MTW) is calculated as:

$$MTW(j) = (R_g + 1)D_j + R_g(traffic demand on all 1 - hop neighbors of j) + (R_g - 1)(traffic demand on all 2 - hop neighbors of j) + \cdots$$

Then the node with maximum *MTW* value is picked as the gateway node. If there are more than one gateway to be deployed, some adjustment on the traffic demand of the routers are needed.

Another heuristic used to compare our approach is Busiest Router Placement (BRP) which consists of picking the nodes with highest traffic demand values as the gateway nodes.

In Appendix G, the outlined heuristics will be compared using the largest network we have.

Chapter 5

Numerical Examples

In order to evaluate the performance of the model and the efficiency of the proposed heuristic, we have generated some numerical applications. For this, some cities in Istanbul and a city in Ankara are chosen. We have deployed a grid topology to the selected areas since Robinson and Knightly (2007) have shown that grid topologies are more realistic in delivering the desired network performance.

First, we have located the grid topology with 1 mile distance between each MR. As we want to have high capacity on the links, we have placed them as close as possible as shown in the Figure 5.1.



Figure 5.1:Placed Grid Topology on Kadikoy

Then we have pointed the centers of each MR as shown in Figure.



Figure 5.2: Placed MRs on Kadikoy

Finally, we have pointed the districts in the chosen cities and we have assigned each district to its nearest MR as shown in Figure. The same colors represent the assignment relations between each MR and the district.



Figure 5.3: Assigning Districts to the Placed MRs

We needed to remove some of the MRs from the system for two reasons:

- 1. If there is no district to be covered within the coverage range of the MR,
- 2. If the geographical conditions do not allow a MR deployment.

To use the advantages of mesh topology, we needed to make sure that each MR has at least 2 neighbors. For this reason, sometimes we needed to keep the MRs even if they are supposed to be removed from the system because of the conditions above.

The traffic demand of each MR is determined as follows: The total population assigned to the MR / 5000. In other words, we have assumed that 5000 people will need 1 mbps traffic load to be carried to the gateway node in each frame. Since the provided service level can be scaled, the number 5000 is not so important in this assumption.

In order to have a grid topology, the ambient noise ratio (γ) and the SINR threshold value (θ) are chosen accordingly. The following table shows the parameters used in the examples.

Parameters	Values	
Capacity of an existing link <i>c</i>	100 mbps	
Capacity of each gateway <i>a</i>	1000 mbps	
Maximum Usable Power P_i	1 mW	
The ambient noise power δ	1	
SINR threshold value θ	0.8	
Path loss exponent α	3	
Number of transmission slots	5,6,7,8,9,10	

Table 5.1: The Parameters used in the Numerical Examples

The cities that are picked are: Kadikoy, Sariyer, Maltepe-Kartal, Pendik, Uskudar-Umraniye and Cankaya. For simplicity, we will use KAD, SAR, MAL, PEN, USK and CAN respectively. To define the number of gateways and the used transmission slots, we will use KAD_X_Y which corresponds exactly X gateways deployed and Y transmission slots are used in Kadikoy. The final networks of all areas can be seen in Appendix A, where each node represents a MR and the dashed lines represent the links between the MRs.

The table 5.2 shows the total traffic demand of each node deployed in the application areas.

The results in the Appendix are obtained by using GUROBI solver 4.5.0 on a 2.27 GHz, i5-M430 computer. We have limited the running time of each run with 3600 seconds.

Node	KAD	SAR	PEN	MAL	USK	CAN
1	9	6.3	14.1	9.7	20.9	10.5
2	11.3	3	8.6	10.9	4.5	11.8
3	4.2	9.6	5.3	12.4	6.4	2.2
4	12.9	4.9	8.2	11.7	7.8	8
5	7.6	0.7	11.1	15.3	4.6	12.9
6	8.2	0.7	28.2	5.7	12.2	7.6
7	15.5	0	6.7	22.6	8	4.6
8	16.3	5.5	2.4	2.9	7.6	5.7
9	13.5	3.1	3.9	10.4	7.1	5.5
10	7.5	2.4	7.2	8	11.8	2.1
11		5.8	2.2	4.5	17.9	13.4
12		5.2	0	7.9	7.3	3.4
13		2.5	1.4	7.9	19	4.6
14		1	2.9	6.6	10.7	7.1
15		2.2	0.7	3.7	7.6	4.7
16			6.6	3.4		2.6
17			0.9	12		11.2
18				14.3		2.6
19						2.5
20						5.1
21						5
22						13

Table 5.2: Traffic demands of each node deployed in the application areas

Let us consider the solution shown in the table below. For this table, we have used 5 transmission slots and up to 3 gateways are deployed to provide service. The first objective value is obtained by using one gateway in the area. This means if we deploy one gateway we can cover the 61.16% of the whole demand. In other words, rather than giving 5000 people 1 mbps in each frame, in the worst case we can give only 0.61 mbps. If we deploy 2 gateways in the area, then we can cover 132.45% of the demand which means they can transmit data much faster and they can increase their traffic demand up to 1.32 mbps.

	m	time (sec)
KAD_1_5	0.611621	1.86
KAD_2_5	1.3245	6.66
KAD_3_5	2.43902	1.78

Table 5.3: Solutions obtained for KAD

A solution found by WMN1 is illustrated in Appendix F. In this solution, node 10 is picked as the gateway node and the traffic demand of each node is carried to that node in each transmission slot by activating non-interfering links.

Chapter 6

Extension

The routing structure of *WMN*1 is commonly used in the literature and in the applications. However, we wanted to see the effect of a flexible routing rather than a tree-structured routing. Like the Multi Protocol Label Switching (MPLS) technology used in wired connections, each MR again has a single route to reach the gateway node but this time a node is allowed to send data to more than one node without dividing the traffic. Since each tree-structured routing is also feasible for the flexible routing, there is no surprise that the solution of the flexible routing will be at least as good as the solution of tree-routed structure.

In the remaining part of the chapter, we will formulate the model which uses flexible routing, define some valid inequalities for the model and compare it with the *WMN*1

6.1 Mixed Integer Linear Programming Model Using Flexible-Routing

6.1.1 Variables

The following variables are needed to model the problem.

$x_{ij}^t = \begin{cases} 1, \\ 0, \end{cases}$	if the link (i,j)is active in transmission slot t otherwise
$y_i = \begin{cases} 1, \\ 0, \end{cases}$	if node i is used as a gateway otherwise
$z_{ij}^s = \begin{cases} 1, \\ 0, \end{cases}$	if the traffic of node s is transmitted using the link(i,j) otherwise
$zout_i^s = \begin{cases} 1, \\ 0, \end{cases}$	if node s uses node i as a gateway node other
$f_{ij}^s = the modelse model$	aximum amount of flow originated from node s,carried on link(i,j)
$fout_i^s = the$	maximum amount of flow originated from s, leaving system with gateway i
$P_i^t = the \ ra$	tio of the transmitted power from node i in transmission slot t
m = minim	um service level

6.1.2 MILP Formulation Using Flexible-Routing

The following model is constructed for the problem defined earlier however rather than tree-routing, flexible-routing is used.

WMN2

max m

$$m \le \frac{\sum_{i=1}^{N} a y_i}{\sum_{i=1}^{N} d_i} \tag{6.1}$$

$$m \le \frac{Gc}{k} \tag{6.2}$$

$$\sum_{s=1}^{N} f_{ij}^s \le \frac{c}{T} \sum_{t=1}^{I} x_{ij}^t \qquad \forall (i,j) \in A$$

$$(6.3)$$

$$f_{ij}^{s} \le md_{s}z_{ij}^{s} \qquad \forall (i,j) \in A, \forall s$$
(6.4)

$$fout_i^s \le md_s zout_i^s \qquad \forall i, s \in N \tag{6.5}$$

$$\sum_{s=1}^{N} fout_{i}^{s} \le ay_{i} \qquad \forall i \in N$$
(6.6)

$$\sum_{i=1}^{N} y_i = G \tag{6.7}$$

$$\sum_{i=1}^{N} zout_{i}^{s} = 1 \qquad \forall s \in N$$
(6.8)

 $zout_i^s \le y_i \qquad \forall i, s \in N$ (6.9)

$$\sum_{i:(i,j)\in A} z_{ij}^s = \sum_{k:(j,k)\in A} z_{jk}^s + zout_j^s \qquad \forall s,j\in N, s\neq j$$
(6.10)

$$\sum_{j:(i,j)\in A} z_{ij}^i = 1 - zout_i^i \qquad \forall i \in N$$
(6.11)

$$\sum_{j:(i,j)\in A} z_{ij}^j = 0 \qquad \forall j \in N$$
(6.12)

$$x_{ij}^t \le \sum_{s=1}^N z_{ij}^s \qquad \forall (i,j) \in A, t \in T$$
(6.13)

$$\sum_{s:(s,i)\in A} x_{si}^t + \sum_{k:(i,k)\in A} x_{id}^t \le 1 \qquad \forall i \in N, t \in T$$
(6.14)

$$P_i^t \le 1 \qquad \forall i \in N, t \in T$$

$$P_i^t g_{ij} + M_{ij} (1 - x_{ij}^t) \ge \theta \gamma + \sum_{s=1, s \neq i, j}^N \theta P_s^t g_{sj} \qquad \forall (i, j) \in A, t \in T$$

$$x_{ij}^t \in \{0, 1\} \qquad \forall (i, j) \in A, t \in T \qquad (6.17)$$

$$y_i \in \{0,1\} \qquad \forall i \in N \tag{6.18}$$

$$z_{ij}^{s} \in \{0,1\} \qquad \forall (i,j) \in A, s \in N$$

$$zout_{i}^{s} \in \{0,1\} \qquad \forall i,s \in N$$

$$(6.19)$$

$$(6.20)$$

$$f_{ij}^s \ge 0 \qquad \forall (i,j) \in A, s \in N$$
 (6.21)

$$fout_i^s \ge 0 \qquad \forall i, s \in N \tag{6.22}$$

$$P_i^t \ge 0 \qquad \forall i \in N, t \in T$$

$$m \ge 0 \qquad (6.23)$$

Our objective is again to maximize the fair service level provided to all customers while ensuring the constraints.

Constraint (6.1) and (6.2) are the upper bounds found earlier. Since the routing aspects have no impact on the calculation of these bounds, these upper bounds are still valid for WMN2.

(6.3) is the capacity constraint which is same as the constraint (3.1) used in WMN1.

Constraint (6.4) means that if the traffic of node *s* is carried on the (i, j) link, then that traffic can not be larger than the provided service level to all nodes times the traffic of node *s*. As one can easily see, constraint (6.4) is non-linear. Following constraints, (6.4*a*), (6.4*b*)and (6.4*c*) are used to linearize (6.4);

$$f_{ij}^{s} \le md_{s} \qquad \forall (i,j) \in A, s \in N \tag{6.4a}$$

$$f_{ij}^{s} \le Rd_{s}z_{ij}^{s} \qquad \forall (i,j) \in A, s \in N$$
(6.4b)

$$f_{ij}^{s} \ge \left(R\left(z_{ij}^{s}-1\right)+m\right)d_{s} \qquad \forall (i,j) \in A, s \in N$$
(6.4c)

where R is a number greater than m.

If $z_{ij}^s = 1$, then the 3 constraints above becomes $f_{ij}^s \le md_s$, $f_{ij}^s \le Rd_s$ and $f_{ij}^s \ge md_s$ respectively. Since *R* is greater than *m*, then the second inequality becomes redundant and we get $f_{ij}^s = md_s$ which is what we desire.

If $z_{ij}^s = 0$, then the 3 constraints above becomes $f_{ij}^s \le md_s$, $f_{ij}^s \le 0$ and $f_{ij}^s \ge (-R+m)d_s$ respectively. Since *R* is greater than *m*, (-R+m) is negative and d_s is greater than 0, than the third inequality becomes redundant. Similarly, since md_s is nonnegative, then the first inequality also becomes redundant. In the end as f_{ij}^s is nonnegative and $f_{ij}^s \le 0$, we get $f_{ij}^s = 0$ which is again what we desire.

Similar to constraint (6.4), (6.5) also ensures that if the traffic of node s reaches Internet by using node i, then the traffic flow from node s using gateway i cannot be greater than the provided service level to all nodes times the traffic of node s. The constraint is also nonlinear and can be linearized using the following additional constraints;

$$fout_i^s \le md_s \qquad \forall i, s \in N \tag{6.5a}$$

 $fout_i^s \le Rd_s zout_i^s \qquad \forall i, s \in N \tag{6.5b}$

$$fout_i^s \ge (R(zout_i^s - 1) + m)d_s \qquad \forall i, s \in N$$
(6.5c)

where R is the same number, greater than m, used to linearize (6.5).

If $zout_i^s = 1$, then the constraints above implies $fout_i^s = md_s$ with the same logic defined earlier and if $zout_i^s = 0$, then the constraints implies $fout_i^s = 0$.

Constraint (6.6) implies that the total traffic flow, using node i to reach the Internet cannot be larger than the capacity of the gateway located at node i. (6.7) implies that the total number of gateways to be deployed have to be equal to G. (6.8) is added to

make sure that traffic of each node reaches the gateway node. Since traffic division is not allowed, a node can use at most one node as a gateway node.

Constraint (6.9) ensures that node *s* can only reach Internet using node *i* if the node *i* is equipped with gateway functionalities. (6.10) on the other hand is added to conserve the traffic flow of node *s*. If the traffic of node *s* reaches node *j* using $(i, j) \in A$ then it needs to leave node *j* using $(j, k) \in A$ unless *s* reaches Internet using node *j*. (6.11) implies that the traffic of each node has to follow a single route to reach the gateway. Similarly, (6.12) ensures that the source of the traffic of node *j* is again node *j* itself.

Constraint (6.13) implies that there can be a transmission between node i and node j if and only if the link under consideration is active in a transmission slot. Constraints (6.14), (6.15) and (6.16) are the same constraints as (3.8), (3.9) and (3.10a).

Finally, the constraints (6.17), (6.18), (6.19), and (6.20) imply that the variables are binary and similarly (6.21), (6.22), (6.23), and (6.24) are added to define nonnegative variables.

In the end, constraints (6.4) and (6.5) are removed and these removed constraints are linearized by adding additional (6.4a), (6.4b), (6.4c), (6.5a), (6.5b) and (6.5c), respectively.

6.1.3 Additional Constraints for WMN2

Proposition 4. The following inequality is valid for the polyhedra defined by the constraints of WMN1.

$$zout_{i}^{s} \leq \sum_{j:(j,i)\in A} z_{ji}^{s} \qquad \forall i,s \in N, i \neq s$$
(6.25)

Proof. Suppose $\sum_{j:(j,i)\in A} z_{ji}^s = 0$, this means the traffic of node *s* does not reach node *i*. Thus node *s* cannot use node *i* as a gateway node and *zout*_i^s has to be equal to 0 which is satisfied by the constraint (6.25).

Now suppose $\sum_{j:(j,i)\in A} z_{ji}^s = 1$, this means the traffic of node *s* reaches node *i*. Then the constraints becomes $zout_i^s \le 1$ which does not restrict $zout_i^s$.

Since traffic division is not allowed, there is no other alternative and the constraint (6.25) is valid for the polyhedra.

Proposition 5. The following inequality is valid for the polyhedra defined by the constraints of WMN1.

$$\sum_{t=1}^{T} x_{ij}^t \ge z_{ij}^s \qquad \forall (i,j) \in A, s \in N$$
(6.26)

Proof. Suppose $z_{ij}^s = 1$, which means the traffic of node *s* is transmitted using the link $(i,j) \in A$. Then (6.26) becomes $\sum_{t=1}^{T} x_{ij}^t \ge 1$ which implies the link $(i,j) \in A$ has to be active in at least one of the transmission slots which is what we desire.

Similarly, suppose $z_{ij}^s = 0$. Then (6.26) becomes $\sum_{t=1}^T x_{ij}^t \ge 0$ which is valid because the link under consideration may need to be activated to transmit the traffic of another node. Thus, the constraint (6.26) is valid for the polyhedra.

After adding the valid inequalities, *WMN2* becomes; *WMN2*

max m

s.t. (6.1) - (6.3), (6.4a) - (6.4c), (6.5a) - (6.5c), (6.6) - (6.24), (6.25), (6.26). In the end, the number of variables and constraints that WMN2 has is listed in the following table.

# of binary variables	# of other variables	#of constraints	
N(N+1)+A(T+N)	N(A+N+T)+1	N(4A+6N+2T+2)+A(2T+1)+3	

Table 6.1: Number of Variables and Constraints of WMN2

6.2 WMN1 vs. WMN2

Consider a randomly generated network as depicted in Figure 6.1. In this topology, there are 16 MRs and 78 links and we are trying to deploy one gateway. The nodes represent the locations of MRs and the dashed lines are the possible links.



Figure 6.1: Randomly Generated Topology

After solving both *WMN*1 and *WMN*2, we get the following results as depicted in Figure 6.2 and Figure 6.3, respectively. The outcomes of these solutions can be seen in Table 6.2.



Figure 6.2: The Outcome of WMN1

In the figure above, each MR follows a single path to reach the gateway node, the 11th node depicted as black, and the resulting topology has a tree structure rooted at gateway node. Each MR can transmit data to at most one node.



Figure 6.3: The Outcome of WMN2

In the figure above, again each MR follows a single path to reach the gateway node however an MR is allowed to transmit more than one node without traffic division. To illustrate MR 14 sends the data of itself and 9 to node 10, whereas it sends the data of node 4 and node 16 to node 6.

Below is the table comparing *WMN*1 and *WMN*2 in terms of the objective value and running time for the randomly generated data.

Model	Objective Value	Running Time	
WMN1	1.42857	3,725 seconds	
WMN2	1.6667	267,792 seconds	

Table 6.2: Comparison of WMN1 and WMN2 for the Randomly Generated Data

For this data, *WMN*2 computes a solution that is 16.67% better than the solution of *WMN*1 however finding the solution takes even days.

Chapter 7

Conclusion

In this study, we have worked on the WMNs which propose an effective way of providing fast and reliable networking solutions. We have jointly considered the problems of gateway selection, routing, scheduling and power management while trying to maximize the worst-case scenario.

First, we have briefly described the general characteristics of WMNs and then focused on ones which use 802.16 protocol. We have described our problem and showed some of the related studies in the literature.

We have proposed a MILP to obtain exact solutions for our problem. However, as the number of nodes and arcs increased in the network, our model began to need excessive amount of time to find the optimal solution.

Further, we needed to improve a heuristic approach to obtain 'good' solutions in reasonable time. We have generated data by deploying grid topologies for some

cities in Istanbul and Ankara. We have used the generated data to evaluate the running time and performance of both our model and the proposed heuristic.

In addition, we wanted to know the effect of using flexible paths instead of a treerouted structure and formulated another model. We have obtained the solutions for this model and compared it with the performance of the first model in terms of running time and the objective value.

As a future direction, more sophisticated heuristics are needed to be developed which take the interference into account. Similarly, the upper bounds found on the objective value are too loose to help the solver to obtain the optimal solution in reasonable time. A way of finding a good bound on the objective value is to use the Lagrangian relaxation. By relaxing the interference constraint, a tighter bound can be found.

Although 802.16 Wi-Max protocol allows multi-rate transmission, we have assumed single-rate transmission in our model. The model can be extended to cover multi-rate case. However this will affect the overall model and increase the number of variables excessively.

The buffer capacities of the MRs are not considered in this study. The model can be extended to cover the buffer capacities and a more realistic approach can be proposed.

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Appendix A



Figure A.1: The network of KAD



Figure A.2: The network of SAR



Figure A.3: The network of MAL



Figure A.4: The network of PEN



Figure A.5: The network of USK



Figure A.6: The network of CAN

Appendix B

	Running Time			
	with Valid Inequality	without Valid Inequality		
MAL_1_5	24.88	20.44		
MAL_1_6	19.53	14.35		
MAL_1_7	21.48	20.2		
MAL_1_8	33.46	26.33		
MAL_1_9	31.11	32.82		
MAL_1_10	32.63	68.44		
MAL_2_5	57.58	69.01		
MAL_2_6	101.2	261.32		
MAL_2_7	456.16	475.9		
MAL_2_8	460.53	2315.46		
MAL_2_9	1085.89	1430.26		
MAL_2_10	1265.35	1964.27		
MAL_3_5	141.77	152.66		
MAL_3_6	95.57	107.16		
MAL_3_7	92.56	82.48		
MAL_3_8	251.91	281.49		
MAL_3_9	1588.47	865.82		
MAL_3_10	1123.89	406.52		

Table B.1: The Effect of Valid Inequality for MAL

Appendix C



Figure C.1: LP Relaxation vs. LP Relaxation with added upper bounds for KAD



Figure C.2: LP Relaxation vs. LP Relaxation with added upper bounds for SAR



Figure C.3: LP Relaxation vs. LP Relaxation with added upper bounds for PEN



Figure C.4: LP Relaxation vs. LP Relaxation with added upper bounds for MAL



Figure C.5: LP Relaxation vs. LP Relaxation with added upper bounds for USK



Figure C.6: LP Relaxation vs. LP Relaxation with added upper bounds for CAN

Appendix D



Figure D.1: The Running Times of Each Model for KAD_1



Figure D.2: The Running Times of Each Model for KAD_2



Figure D.3: The Running Times of Each Model for KAD_3



Figure D.4: The Overall Running Times of Each Model for KAD



Figure D.5: The Running Times of Each Model for SAR_1



Figure D.6: The Running Times of Each Model for SAR_2



Figure D.7: The Running Times of Each Model for SAR_3



Figure D.8: The Overall Running Times of Each Model for SAR



Figure D.9: The Running Times of Each Model for PEN_1



Figure D.10: The Running Times of Each Model for PEN_2



Figure D.11: The Running Times of Each Model for PEN_3



Figure D.12: The Running Times of Each Model for SAR_1

Appendix E

	m	time (sec)		
KAD_1_5	0.611621	0.75		
KAD_2_5	1.3245	2.12		
KAD_3_5	2.43902	0.92		
KAD_1_6	0.757576	0.73		
KAD_2_6	1.47493	2.28		
KAD_3_6	2.15054	2.11		
KAD_1_7	0.688468	1.22		
KAD_2_7	1.66113	2.00		
KAD_3_7	2.21484	2.9		
KAD_1_8	0.852273	1.28		
KAD_2_8	1.64474	2.82		
KAD_3_8	2.41935	2.82		
KAD_1_9	0.803213	2.5		
KAD_2_9	1.47167	3.99		
KAD_3_9	2.58398	2.92		
KAD_1_10	0.775194	4.34		
KAD_2_10	1.74419	9.86		
KAD_3_10	2.58065	3.79		

Table E.1: Objective Value and Running time of KAD



Figure E.1: The objective value of KAD using 5,6 and 7 transmission slots



Figure E.2: The objective value of KAD using 8,9 and 10 transmission slots

	m	time (sec)
SAR_1_5	1.10497	5.55
SAR_2_5	2.75862	5.77
SAR_3_5	4.44444	5.66
SAR_1_6	1.31234	5.76
SAR_2_6	3.0303	17.5
SAR_3_6	5.20833	6.51
SAR_1_7	1.4881	7.96
SAR_2_7	2.91545	21.23
SAR_3_7	5.19481	20.94
SAR_1_8	1.38122	9.45
SAR_2_8	2.95276	28.97
SAR_3_8	5.51471	29.2
SAR_1_9	1.5015	9.81
SAR_2_9	3.47222	26.75
SAR_3_9	5.74713	22.67
SAR_1_10	1.5748	16.04
SAR_2_10	3.1746	50.98
SAR_3_10	6.25	41.78

Table E.2: Objective Value and Running time of SAR



Figure E.3: The objective value of SAR using 5,6 and 7 transmission slots



Figure E.4: The objective value of SAR using 8,9 and 10 transmission slots

	m	time (sec)
PEN_1_5	0.70922	22.22
PEN_2_5	1.36054	25.35
PEN_3_5	2.32558	26.68
PEN_1_6	0.734214	37.28
PEN_2_6	1.5015	22.96
PEN_3_6	2.52525	35.9
PEN_1_7	0.789266	23.49
PEN_2_7	1.30463	52.64
PEN_3_7	2.69542	53.18
PEN_1_8	0.722543	26.07
PEN_2_8	1.64474	66.19
PEN_3_8	2.90698	47.77
PEN_1_9	0.773395	39.34
PEN_2_9	1.57604	92.46
PEN_3_9	2.849	166.64
PEN_1_10	0.828729	27.66
PEN_2_10	1.79372	128.93
PEN_3_10	2.7027	175.72

Table E.3: Objective Value and Running time of PEN



Figure E.5: The objective value of PEN using 5,6 and 7 transmission slots



Figure E.6: The objective value of PEN using 8,9 and 10 transmission slots

	m	time (sec)
MAL_1_5	0.337268	24.88
MAL_2_5	0.684932	57.58
MAL_3_5	0.970874	141.77
MAL_1_6	0.361533	19.53
MAL_2_6	0.737463	101.2
MAL_3_6	1.23457	95.57
MAL_1_7	0.407	21.48
MAL_2_7	0.716076	456.16
MAL_3_7	1.36054	92.56
MAL_1_8	0.442478	33.46
MAL_2_8	0.816993	460.53
MAL_3_8	1.24585	251.91
MAL_1_9	0.422476	31.11
MAL_2_9	0.835422	1085.89
MAL_3_9	1.26743	1588.47
MAL_1_10	0.443787	32.63
MAL_2_10	0.877193	1265.35
MAL_3_10	1.3289	1123.89

Table E.4: Objective Value and Running time of MAL



Figure E.7: The objective value of MAL using 5,6 and 7 transmission slots



Figure E.8: The objective value of MAL using 8,9 and 10 transmission slots

	m	time (sec)		
USK_1_5	0.48193	15.07		
USK_2_5	0.78431	26.35		
USK_3_5	1.63934	25.9		
USK_1_6	0.4845	17.05		
USK_2_6	0.87719	60.92		
USK_3_6	1.41243	48.47		
USK_1_7	0.50839	20.81		
USK_2_7	0.89847	83.12		
USK_3_7	1.443	140.38		
USK_1_8	0.48263	44.43		
USK_2_8	0.98039	84.75		
USK_3_8	1.67785	74.32		
USK_1_9	0.48876	49.81		
USK_2_9	0.98328	115.94		
USK_3_9	1.65017	247.11		
USK_1_10	0.5305	31.65		
USK_2_10	1.05263	80.76		
USK_3_10	1.63934	190.33		

Table E.5: Objective Value and Running time of USK



Figure E.9: The objective value of USK using 5,6 and 7 transmission slots



Figure E.10: The objective value of USK using 8,9 and 10 transmission slots

	m	time (sec)		
CAN_1_5	0.394477	695.89		
CAN_2_5	0.909091	1012.38		
CAN_3_5	1.42857	1765.39		
CAN_1_6	0.485909	547.19		
CAN_2_6	1.06838	631.84		
CAN_3_6	1.29199	3580.32		
CAN_1_7	0.519481	243.13		
CAN_2_7	1.0661	1148.37		
CAN_3_7	AN_3_7 1.38696			
CAN_1_8	0.55066	941.56		
CAN_2_8	0.992063	2450.11		
CAN_3_8	1.4881-1.86567 (25.4%)	3600		
CAN_1_9	0.555556-0.66985 (20.6%)	3600		
CAN_2_9	0.996512-1.17766 (18.2%)	3600		
CAN_3_9	1.49477-2.05144 (37.2%)	3600		
CAN_1_10	0.59701-0.75358 (26.2%)	3600		
CAN_2_10	1.03093-1.44635 (40.3%)	3600		
CAN_3_10	1.45631-2.80899 (92.9%)	3600		

Table E.6: Objective Value and Running time of CAN



Figure E.11: The objective value of CAN using 5,6 and 7 transmission slots



Figure E.12: The objective value of CAN using 8,9 and 10 transmission slots

Appendix F



Figure F.1: 10th node is picked as the gateway node



Figure F.2: Transmissions occurring in the first transmission slot



Figure F.3: Transmissions occurring in the second transmission slot



Figure F.4: Transmissions occurring in the third transmission slot



Figure F.5: Transmissions occurring in the fourth transmission slot



Figure F.6: Transmissions occurring in the fifth transmission slot



Figure F.7: Transmissions occurring in the sixth transmission slot



Figure F.8: Transmissions occurring in the seventh transmission slot

	CPMP'		Zhou		BRP		WMN1	
	Performance	Time	Performance	Time	Performance	Time	Performance	Time
CAN_1_5	0.22805	207.73	0.22805	207.73	0.22805	207.73	0.394477	695.89
CAN_1_6	0.47893	33.91	0.47893	33.91	0.47893	33.91	0.485909	547.19
CAN_1_7	0.51948	23.62	0.51948	23.62	0.51948	23.62	0.519481	243.13
CAN_1_8	0.55066	34.6	0.55066	34.6	0.55066	34.6	0.55066	941.56
CAN_1_9	0.55556	55.86	0.55556	55.86	0.55556	55.86	0.555556	3600
CAN_1_10	0.59701	3600	0.59701	3600	0.59701	3600	0.59701	3600
CAN_2_5	0.68729	40.36	0.63942	48.86	0.62305	117.64	0.909091	1012.38
CAN_2_6	0.93109	24.96	0.8547	56.69	0.59952	191.71	1.06838	631.84
CAN_2_7	1.02041	29.06	0.91575	75.01	0.72516	83.66	1.0661	1148.37
CAN_2_8	0.93284	73.27	0.93284	81.29	0.77882	532.85	0.992063	2450.11
CAN_2_9	0.95785	117.54	0.99651	79.64	0.82305	130.18	0.996512	3600
CAN_2_10	0.95238	3600	0.95238	162.8	0.8658	3600	1.03093	3600
CAN_3_5	0.89686	35.22	0.90909	42.04	0.97561	71.45	1.42857	1765.39
CAN_3_6	1.24378	25.79	1.15741	24.84	1.19048	180.33	1.29199	3580.32
CAN_3_7	1.10742	32.04	1.0661	51.31	1.21065	245.06	1.38696	2226.13
CAN_3_8	1.19048	125.39	1.05932	277.70	1.28866	694.86	1.4881	3600
CAN_3_9	1.36054	71.45	1.10011	274.89	1.49477	53.56	1.49477	3600
CAN 3 10	1.43541	165.76	1.18343	284.36	1.42857	188.81	1.45631	3600

Appendix G

Table G.1: Comparison of the Proposed Heuristic