

Faculdade de Engenharia da Universidade do Porto



FEUP

Routing Optimization for Delay Tolerant Networks in Intelligent Environments

Constantino Antunes

Tese submetida no âmbito do
Mestrado Integrado em Engenharia Electrotécnica e de Computadores

Major de Telecomunicações

Orientador: Ricardo Morla

Junho de 2008

A Dissertação intitulada

“Routing Optimization for Delay-Tolerant Networks in Intelligent Environments”

foi aprovada em provas realizadas 15/Julho/2008

o júri

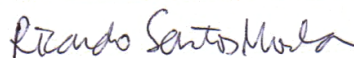
Presidente Professor Doutor Manuel Alberto Pereira Ricardo
Professor Associado da Faculdade de Engenharia da Universidade do Porto



Professor Doutor Carlos Miguel Ferraz Baquero Moreno
Professor Auxiliar da Escola de Engenharia da Universidade do Minho



Professor Doutor Ricardo Santos Morla
Professor Auxiliar da Faculdade de Engenharia da Universidade do Porto



O autor declara que a presente dissertação (ou relatório de projecto) é da sua exclusiva autoria e foi escrita sem qualquer apoio externo não explicitamente autorizado. Os resultados, ideias, parágrafos, ou outros extractos tomados de ou inspirados em trabalhos de outros autores, e demais referências bibliográficas usadas, são correctamente citados.

Autor - Constantino Rocha Antunes



Faculdade de Engenharia da Universidade do Porto

Resumo

Este trabalho apresenta um estudo sobre routing DTN e possíveis otimizações. São propostas e testadas três melhorias ao protocolo PROPHET, mostrando aumentos de entrega de mensagens em torno de 1% e redução de gasto de recursos em cerca de 1%. É também proposto um protocolo híbrido que mostrou ser capaz de entregar uma quantidade de mensagens acima de 99% do valor máximo possível para a rede, ao mesmo tempo que usa apenas mais 0.3% e 0.2% de buffer e largura de banda por mensagem entregue do que o protocolo PROPHET.

Abstract

This work presents a study on DTN routing and possible optimizations. Three improvements for the PROPHET protocol are proposed and tested, showing message delivery increase around 1% and resource saving around 1%. A hybrid protocol is evaluated and observations show its capability of delivering an amount of messages above 99% of the maximum delivery possible for the network while exhibiting buffer occupancy and bandwidth usage cost only 0.3% and 0.2% above that of PROPHET, which is a reference regarding these two metrics. These results were obtained from data collected in real usage situations by third-party projects.

Acknowledgements

I would like to thank my supervisor, Ricardo Morla, for his countless advices and for his contributions for this work. I would also like to thank Nathan Eagle for providing the Reality Mining Database.

Contents

1	Introduction	1
1.1	Context	1
1.2	Goals	2
1.3	Structure	2
2	Related Work	5
2.1	DTN protocols	5
2.2	DTN Applications	13
2.3	User’s Context	14
2.4	Other related work	17
3	Methodology	21
3.1	Simulator	21
3.1.1	Addresses	22
3.1.2	Links and Connections	23
3.1.3	Messages	23
3.2	Traffic models	23
3.3	Networks	23
3.3.1	UMass	24
3.3.2	Reality Mining	24
3.4	Evaluation metrics	24
3.4.1	Arrivals	25
3.4.2	Buffer	25
3.4.3	Xmit	25
3.4.4	Cost	25
3.4.5	Metrics discussion	27
3.5	Simulations	28
3.6	Work Organization	28
4	Comparison Protocols	31
4.1	Epidemic	31
4.1.1	Results	32
4.1.2	Discussion	33
4.2	DTLSR	34
4.2.1	Results	34
4.2.2	Discussion	35
4.3	PROPHET	35
4.3.1	Results	36

4.3.2	Discussion	37
4.4	Comparison — UMass	37
4.5	Reality Mining Results	41
4.5.1	Epidemic Results	41
4.5.2	PROPHET Results	42
4.6	Comparison — Reality Mining	43
5	Changes to PROPHET	45
5.1	EQ-GRTR	45
5.1.1	Results	45
5.1.2	Discussion	46
5.2	Fast Start	47
5.2.1	Results	47
5.2.2	Discussion	48
5.3	Estimated Delivery Confirmation (EDC)	49
5.3.1	Results	50
5.3.2	Discussion	51
5.4	Conclusions	51
6	Hybrid Protocol	53
6.1	Decision limit	54
6.1.1	Results	54
6.1.2	Discussion	57
6.1.3	Limit decision	58
6.2	Behaviour analysis	60
6.3	Validation	61
6.3.1	Results	62
6.3.2	Discussion	63
6.4	Conclusions	63
7	Conclusion	65
7.1	Methodology Evaluation	65
7.2	Contribution Evaluation	66
7.2.1	Metrics	66
7.2.2	PROPHET changes	66
7.2.3	Hybrid Protocol	67
7.3	Future Work	67
	References	71

List of Figures

2.1	Number messages and delivery delay results from [2]	7
2.2	Queuing time results from [3]	8
2.3	Messages results from [4]	9
2.4	Delivery delay results from [4]	10
2.5	Message delivery delay results from [9]	12
2.6	Message delivery delay results from [10]	12
2.7	Pairs encounter results from [19]	14
2.8	Pairs encounters per hour from [19]	15
2.9	Average pairs encounters per day from [19]	15
2.10	Real and synthetic trace from [19]	16
2.11	Protocol results from [19]	17
2.12	Throughput in function of bundle size from [24]	18
3.1	DTN2 classes	22
4.1	Epidemic vs Flood, day 1 of UMass network	32
4.2	Epidemic, UMass network	33
4.3	Epidemic, UMass network	33
4.4	DTLSR, UMass network	34
4.5	DTLSR, UMass network	35
4.6	PROPHET, UMass network	36
4.7	PROPHET, UMass network	37
4.8	Comparison, day 1 of UMass network	38
4.9	Comparison, day 2 of UMass network	38
4.10	Comparison, day 3 of UMass network	38
4.11	Comparison, day 1 of UMass network: costs	39
4.12	Comparison, day 2 of UMass network: costs	39
4.13	Comparison, day 3 of UMass network: costs	40
4.14	Comparison, UMass network: DTLSR's costs relative to Epidemic	40
4.15	Comparison, UMass network: PROPHET's costs relative to Epidemic	40
4.16	Epidemic, Reality Mining network	41
4.17	Epidemic, Reality Mining network	41
4.18	PROPHET, Reality Mining network	42
4.19	PROPHET, Reality Mining network	42
4.20	Comparison, PROPHET relative to Epidemic	42
5.1	EQ-GRTR, UMass results	46
5.2	EQ-GRTR, Reality Mining results	46
5.3	Fast Start, UMass results	48

5.4	Fast Start, Reality Mining results	48
5.5	EDC, UMass results	50
5.6	EDC, Reality Mining results	50
6.1	Hybrid Router	54
6.2	Hybrid Router, UMass arrivals	55
6.3	Hybrid Router, UMass buffer cost	55
6.4	Hybrid Router, UMass xmit cost	56
6.5	Hybrid Router, Reality Mining arrivals	56
6.6	Hybrid Router, Reality Mining buffer cost	56
6.7	Hybrid Router, Reality Mining xmit cost	57
6.8	Hybrid Router, small limits tests	58
6.9	Hybrid Protocol, $\varepsilon = 0.5$	59
6.10	Hybrid Protocol, $\varepsilon = 0.5$	59
6.11	Hybrid Protocol, $\varepsilon = 0.5$	59
6.12	Reality Mining, week days and arrivals	60
6.13	Reality Mining, pairs	61
6.14	Hybrid validation, arrivals	62
6.15	Hybrid validation, buffer-cost	62
6.16	Hybrid validation, xmit-cost	62

List of Tables

3.1	Umass network contacts and nodes	24
3.2	Reality Mining network months	25
3.3	Reality Mining network contacts and nodes	26
3.4	Cost and its relation with arrivals	26
3.5	Cost and its relation with arrivals, simplified table	27
3.6	Umass network contacts and nodes, between 6 and 12 am	28
3.7	Reality Mining network contacts and nodes, between 6 and 12 am	28
5.1	EQ-GRTR's GRTR ratio	47
5.2	Fast Start's PROPHET ratio	49
5.3	EDC's PROPHET ratio	51
6.1	Reality Mining network week days, contacts and nodes	58
6.2	Hybrid protocol, Epidemic's ratio averages	60
6.3	Reality Mining pairs, average and deviation	61
6.4	Validation, average and deviation	62
6.5	Validation, Protocols Comparison	63
6.6	Validation, Epidemic's ratio averages	63

Chapter 1

Introduction

1.1 Context

Computational devices share data through wired or wireless links. Whenever a device tries to send data, it uses that link to determine if the destination is reachable. If it is not, the communication fails. Traditional network protocols do not solve this problem. For example, ad-hoc protocols are useful in mobility scenarios but do not work when network nodes disappear and partition the network. This happens because these protocols are designed with the assumption of permanent links and are useless to communicate in an environment where data is lost because connections are intermittent.

Being carried by people, personal wireless devices will see their location change with time, possibly moving out of range of any access point. Ideally, one would want to provide network connectivity without restricting users mobility. The obvious solution would be to place enough access points to cover the desired area, guaranteeing that there would always be a network connection. However, the area may be too large or not easily accessible, by which deploying the necessary access points would not make economic sense.

An approach to provide network connectivity in such partitioned networks is for the device which is sending the data to store that data until an available link is found. The node which receives the data should be able to wait for another connection to pass it on. This kind of network is called a Delay Tolerant Network (DTN), a kind of network that connects devices that have the ability to store information and forward it when a receiver is found. This way, even when no access point is reachable and there is not permanent links, communication is still possible.

Besides the simple description of a DTN, routing on it is very difficult. Consider an area of arbitrary size with people moving inside it. Each person carries a device capable of communicating with other devices. Consider also that each device has no knowledge

of its movements nor other devices movements. On this scenario, routing a message from one device \mathcal{A} to a device \mathcal{B} means that \mathcal{A} may choose to wait for a connection with \mathcal{B} or send the message to other devices, hoping that they will deliver it to \mathcal{B} . The second option seems to offer greater probability of message delivery. But, to which devices should the message be sent? Should it be sent to all devices met or a limited number? Should it be copied or not?

1.2 Goals

Any successful communication network has to be able to deliver data with low error rate which is essential for the messages to reach their destination and do that with its information untouched. The data should also be delivered as fast as possible and any required routing data should be kept in minimum quantities so the transmission medium is optimized for message delivery.

This work's objectives were to study approaches on DTN routing, improve them and build a protocol that combines the better protocols studied. These objectives were accomplished having in mind the requirements for general communication network applying them on the DTN context:

1. **Maximize arrivals** — this requirement is equivalent to the referred low error rate necessary for any network;
2. **Minimize buffer occupancy** — since a DTN node saves messages for future delivery, the space for message storage is essential and should be saved;
3. **Minimize transmissions** — the transmission medium should be saved so messages are transmitted as fast as possible between nodes.

It is also important to minimize the delivery delay, but that is more dependent on the network and frequency of node contacts itself than the routing protocol.

1.3 Structure

Approaches to answer the questions in section 1.1 were proposed on previous work and are presented in the next chapter. The related work provided information on successful routing protocols and the data used for realistic simulation. The details on the decisions done for tests and evaluation are presented in chapter 3.

Before optimize or create new protocols, a chosen number of them were tested (chapter 4). These tests allowed validation of implemented protocols and the simulator itself. After choosing and testing protocols, optimization approaches were experimented and the

results are presented in chapter 5. The combination of two protocols resulted into a new protocol (described and validated in chapter 6) which exhibits the best features of both base protocols.

Chapter 2

Related Work

Related work on this subject is rich, many articles may be found at the Delay Tolerant Research Group (DTNRG) web site [1]. This chapter is divided into four sections. The first presents related work on diverse routing protocols. The second refers some applications of DTN. Section 3 presents related work on user context and section 4 presents other papers that, while not directly related to this work, provide different views about DTNs that might be considered.

2.1 DTN protocols

A study about a scenario where nodes are free to move and do not have information about their neighbours is done in [2]. This work is about a successful albeit simple protocol called Epidemic Protocol. It consists of spreading the data like a disease for all nodes or a maximum number of nodes. Two assumptions are made in [2]:

1. Nodes are blind — meaning they do not have any external information about other nodes;
2. Nodes are independent — which means that nodes movement do not depend of other node's movement. Other works to be presented in this section observe that this is not a real situation but it was useful in the context of [2].

A metric called *willingness* was associated to each node. This metric reflects the willingness of a node to participate in messages forwarding. Four schemes for message spreading control were introduced as well:

1. Basic Probabilistic (BP) — uses of an uniform probabilistic function to determine the willingness of a node;
2. Time to Live (TTL) — the maximum number of hops for each message, when TTL reaches zero the message is discarded;

3. Kill Time — after this time the message is discarded;
4. Passive Cure — when the destination receives a message, sends an ACK which is copied to each node trying to send the already received message.

The evaluation metrics included the number of messages sent by each node and the message delivery delay. As can be seen in figure 2.1, BP generates the largest number of messages in the system. Other schemes added to BP allow less volume of messages, the better being BP + TTL + Passive Cure. It would be interesting to evaluate other schemes without BP.

Other approach to the same problem of creating message delivery confirmation is presented in [3]. This is done with four testing schemes:

1. Hop by Hop — each node sends an ACK to the node from which it received the message. This does not provide end to end reliability and was only used for comparison purposes;
2. Active Receipt — generates an ACK which behaves like any other message, being sprayed to all nodes;
3. Passive Receipt — generates an ACK sent only when a node tries to forward a message already delivered;
4. Network-Bridged Receipt — generates an ACK sent by cellular network. This requires all nodes to have a cellular network interface.

Simulation results are displayed in figure 2.2. The best scheme offering end-to-end reliability was Network-Bridged and the best scheme not requiring an external network was Active Receipt, considering the average queuing time as evaluation metric.

Despite the simplicity of the Epidemic Protocol, it uses excessive bandwidth. This problem is analysed in [4] which proposed a new protocol called PROPHET that does not require copying all messages to all nodes and can offer a similar performance to that of the Epidemic Protocol. This is a probabilistic type of routing, a way for the nodes to decide to which nodes they will forward messages based in the probability of successful message delivery. The routing metric, called *predictability*, was calculated on the history of nodes encounters and associated to each node and each destination. An ageing constant corrects predictability's value when nodes do not meet before a certain amount of time, and a transitivity effect which affects a node's predictability with other nodes predictabilities. The example given in [4] is: if a network node \mathcal{A} meets a node \mathcal{B} frequently and \mathcal{B} meets node \mathcal{C} frequently, then \mathcal{A} has a high predictability when delivering messages to \mathcal{C} even if \mathcal{A} has a low probability of seeing \mathcal{C} . The simulation introduced a scenario called *Communitary Model* in which network nodes movements are based in people's movements.

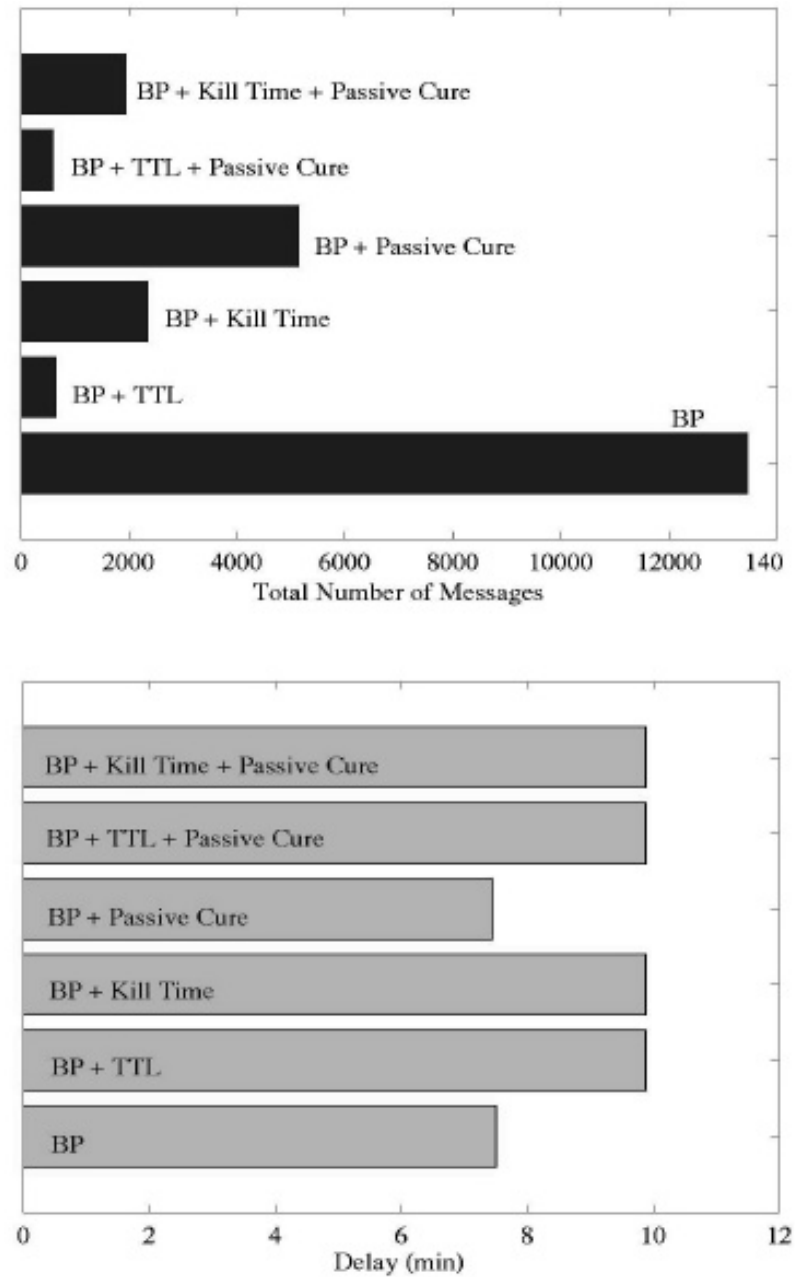


Figure 2.1: Number messages and delivery delay results from [2]

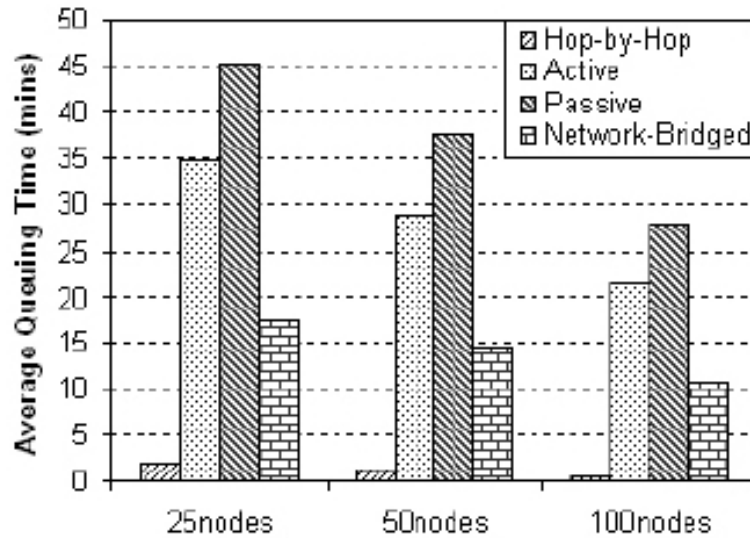


Figure 2.2: Queuing time results from [3]

In this model, each node has a place which it visits frequently and all nodes have a common *Gathering Place* where they all meet. This tries to simulate movements similar to people travelling from home to work. The PROPHEt protocol was simulated in the referred scenario and in a completely random scenario. The Epidemic protocol was simulated too, for comparison purposes. The evaluated metrics were the number of forwarded messages (figure 2.3) and delay (figure 2.4) against the queue size. PROPHEt shows better results in the communitary model and equivalent results in the random model.

Another approach to random mobility protocols called MoVe is analysed in [5] and compared against two simple protocols:

1. Broadcast — similar to Epidemic.
2. NoTalk — the carrier is only allowed to forward the message to the ultimate destination.

The MoVe protocol is a geographical form of routing that uses node position and velocity to forward messages to other nodes moving into the message's destination position. This protocol suffers from the lack of knowledge about the future movements of nodes and the lack of knowledge about the position of the message destination, when starting a communication. Results show performance between the other two protocols.

An integration of DTN with AODV is presented in [6]. The strategy is to use AODV, which is an ad-hoc distance vector protocol, to find routes. With this scheme, nodes use end-to-end connection when the found route allows that. The same mechanism used to find AODV routes also deliver information about DTN reachability. This means that even when no end-to-end route is found, it is possible for a node to reach a destination through

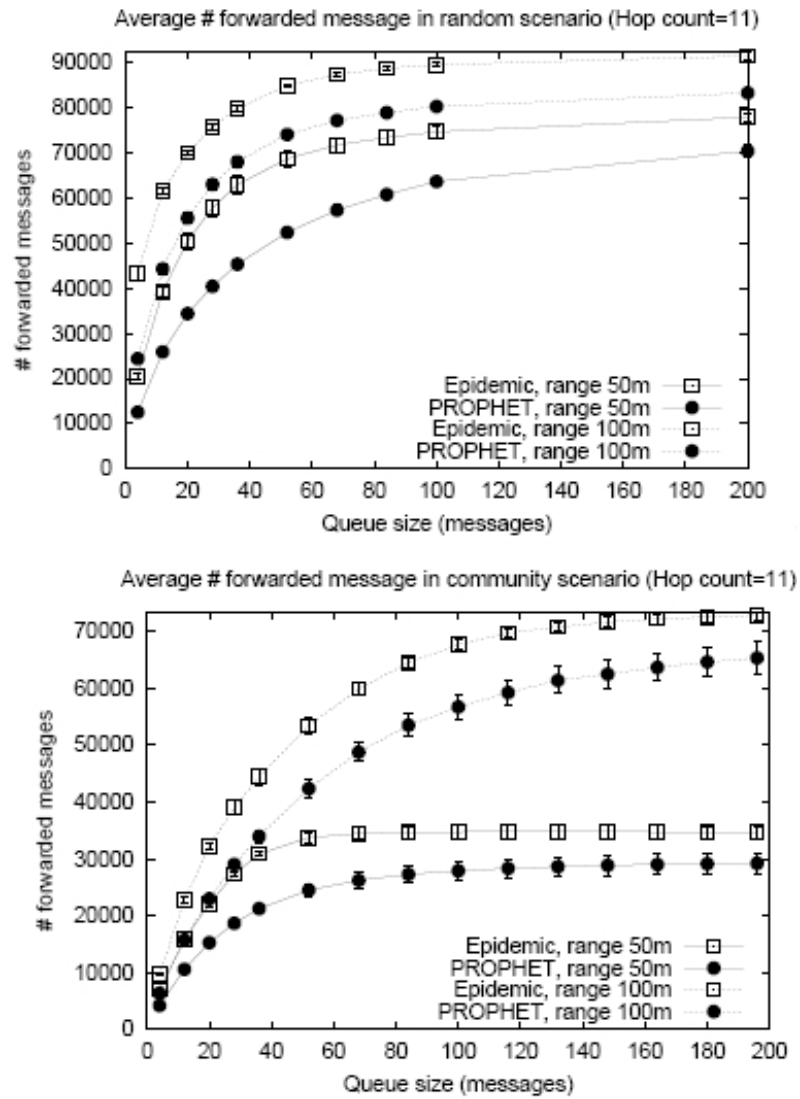


Figure 2.3: Messages results from [4]

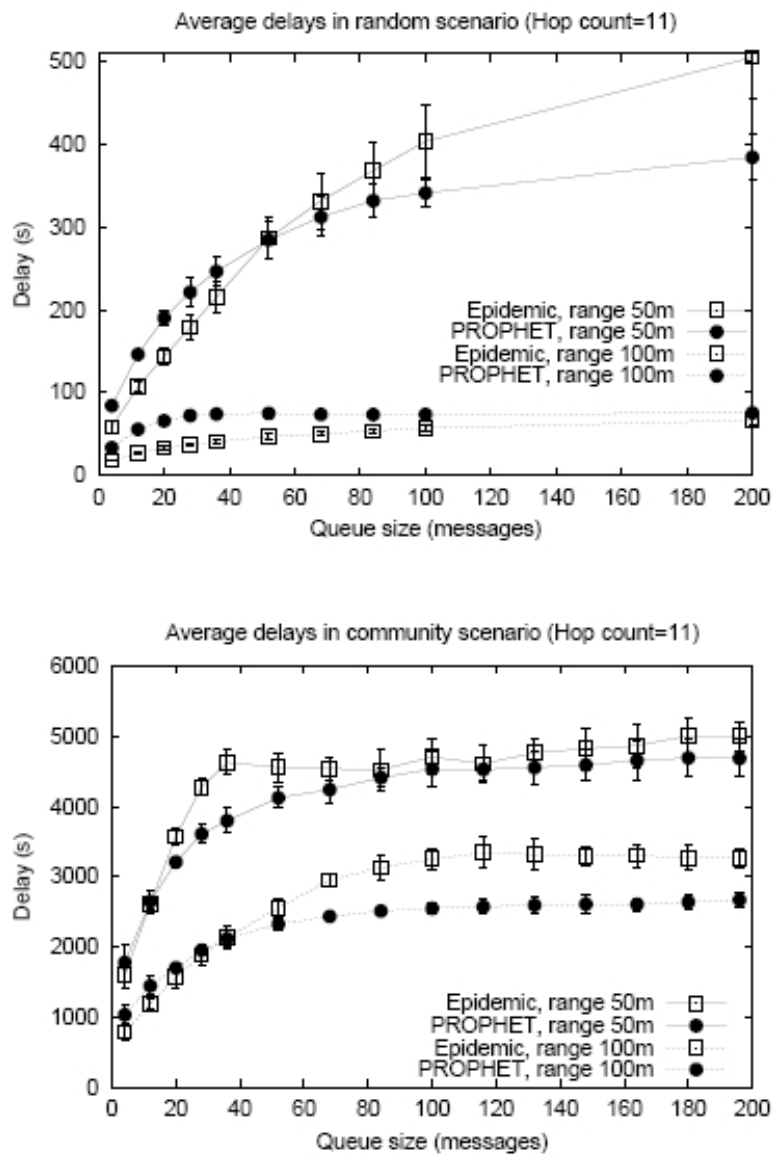


Figure 2.4: Delivery delay results from [4]

DTN routing. This network may be viewed as an ad-hoc network composed by a cloud of network nodes. When a node needs to send a message to other node that is not in the cloud it uses DTN routing. Simulations of this system included a scenario having network nodes with random movement without stops and speeds up to 20m/s. Results showed that AODV links using TCP dominated.

A probabilistic solution based in Kalman filters for network nodes encounter prediction is presented in [7]. The advantage of this prediction technique is that it does not require full knowledge of nodes encounters history to predict the next encounter. Instead, it uses a Kalman filter which is a filter from control theory that uses a system input to estimate the expected output value for that system. Then, the system's real output is used to correct the estimation. This filter uses only the system state to predict its output values and has good results even when dealing with unknown system model and non-linear systems [8]. The evaluated metrics were queue size, delivery ratio and delay. The results show that the proposed protocol (CAR) successfully delivered less messages than Epidemic Protocol but required less buffer bandwidth and has lower delay associated to message delivery.

[9] presents a protocol for predictable routes. In order to do that, a study on the effects of additional information over the network performance was done. A way to provide information to network nodes, or *oracles*, was created. The provided information was about nodes contacts connections, contacts statistics, queues occupancy and traffic demand. Three types of routing were defined:

1. Zero Knowledge — using no oracles;
2. Partial Knowledge — using one or more oracles except traffic oracle;
3. Full Knowledge — using all oracles.

Since the first two are the more realistic when considering the possibility of implementation, the attention was devoted to them. Some protocols were defined and the results shown that the best was EDLQ (Earliest Delivery with Local Queuing) (figure 2.5) which uses Contacts Oracle to calculate neighbours delay. This protocol also takes into account the local queue occupancy and uses a modified Dijkstra algorithm to calculate new paths every time a node is met, since optimal routes change with time. Dijkstra's algorithm is commonly used in link-state protocols to choose between a set of paths the one which has minimum cost. Calculations are done with data sent by routers but, in the case of a DTN that data is often out-of-date therefore, the modified version of Dijkstra's algorithm presented in [9] may provide benefits to network performance. Also, before the end of the paper is suggested that the best protocol choice for this scenario should have information about the contacts history, which is a challenge in partitioned systems.

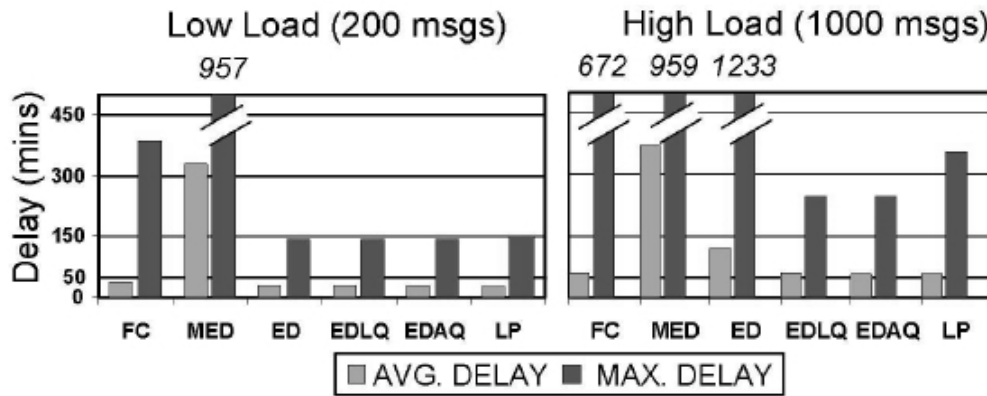


Figure 2.5: Message delivery delay results from [9]

Some link-state protocols are analysed in [10] with the objective of determining which is the best protocol for a slow, predictable mobility scheme. This was done with the conversion of a standard link-state protocol into a delay tolerant link-state protocol by introducing longer lifetime Link-State Advertisement (LSA) messages, longer lifetime of users messages and best path calculation that includes links which may not be reachable at the calculation time. The protocol, called DTLSR performed much better than common link-state protocol.(figure 2.6)

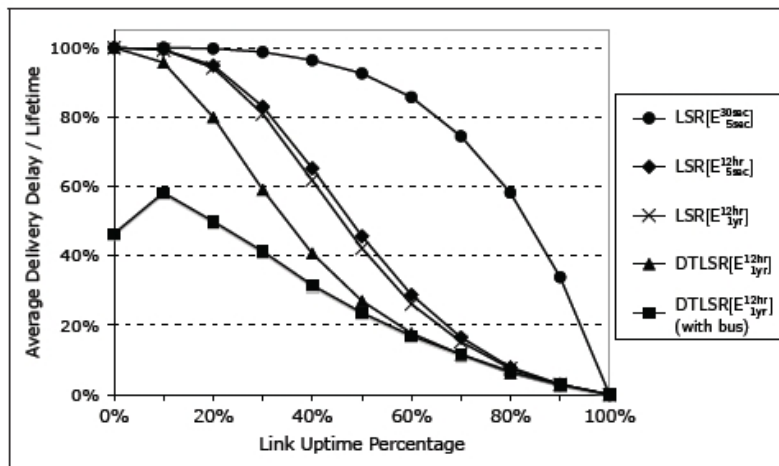


Figure 2.6: Message delivery delay results from [10]

The details of a predictable mobility scenario simulation are studied in [11]. This work collected data from 40 buses of DieselNet network that surrounds the UMass Amherst Campus. The results were collected applying Epidemic protocol with active cure for message spreading control and the metrics presented were the delay, number of copies made and number of hops each messages done. Models for cyclic movement were created and the one which shown results closer to the real scenario was the Route-Level model. This

and other traces may be found in [12] which may be useful to build a network model. This work also shown that it is important to have a detailed model to achieve trustworthy results.

2.2 DTN Applications

DTN is a kind of network that expects delay in messages delivery. Having that said, it seems clear that it won't perform very well with streaming communications such as video conference. This kind of network should have better performance with asynchronous applications instead of synchronous ones. Related work on this matter presents mostly distributed file systems for disaster scenarios [13] where redundancy is needed and a flexible network that is resistant to partitioning that often occurs in such situations. The proposed application is a knowledge repository about e.g. medical techniques to use in emergency situations. The results shown that random copies to several network nodes are more resistant to network partitioning and node loss.

The possibility to use queued messages in DTN routers as cached content for faster access is studied in [14]. This shown that some copies of messages sprayed through the network may also be used as cache and will improve network's performance. This only applies to answer messages which should have some usable content and it won't apply to private information which should be encrypted.

Tests on the performance of a DTN transporting HTTP was presented in [15]. HTTP was not modified, which allowed to use common web browsers (Firefox web browser was used), and web cache proxies with DTN routers. The paper ends with the observation that the HTTP over DTN system has similar performance of HTTP over TCP when no latency is present. This is important because it means that a mobile user may choose to use DTN all the time instead of switching to fixed mode of operation. A similar logic is exploited in [16]: the possibility of provide intermittent Internet access in highways. The discussion presented the situation of an highway having access points sprayed along it and travelling cars which would find the access point signal announcing network, connect to the network and request/download web pages. The main problems are car's velocity and network hardware range. This limits the amount of time each car have network connection and consequently the quantity of downloadable information.

[17] presents a system capable of increasing the network performance by introducing awareness of network context. This context awareness refers to the type of information being transferred and network performance. The objective was to build a system capable of adapting itself to the demanded traffic. The system should detect abnormal situations

degrading network performance such as nodes failures (one example gave in the referred article was a meteorite shower). After such situations are detected, the presented system should enter an evaluation phase in which it decides either to use other protocol capable of performing better in the new context or abort the transmission.

A DTN security problem is studied in [18]. In this kind of network, because connection time and link bandwidth are limited, it may be necessary to use message fragmentation to optimize link usage. [18] presents some schemes to guarantee that a fragment sent over a link is not a waste of bandwidth. However, this requires further work. In our work we will consider that fragmentation won't occur and will assume that the DTN will be deployed in a secure environment.

2.3 User's Context

As already noticed in previous works, network nodes do not move randomly. The real situation is that each is owned by a person having a home, work and friends. People have their own routines, for example they have a week-day routine of going to work but at night it is common for everyone to go home. It is reasonable to think that at weekends the behaviour changes. This situation is exploited in [19] which uses data from [20] to determine patterns in peoples behaviour and separate the studied environment into two classes — *friends* and *strangers* , based in encounter probability (figure 2.7). In this

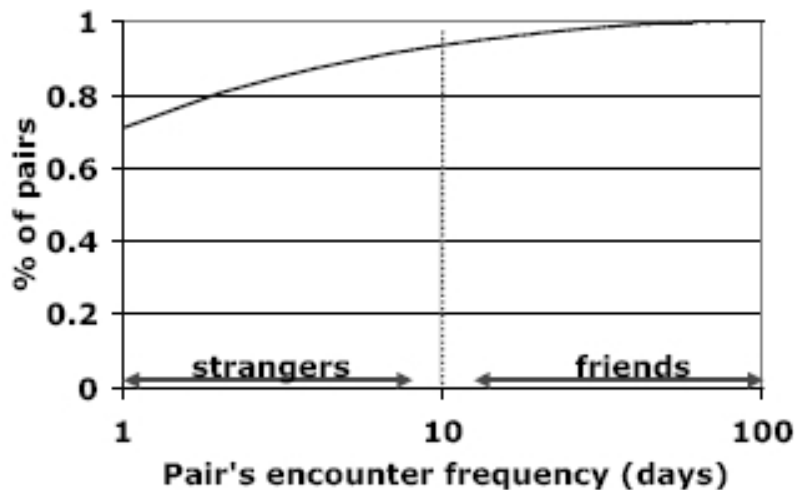


Figure 2.7: Pairs encounter results from [19]

analysis it was observed that during week days encounters occur more often at afternoons having a peak at 16:00 (figure 2.8) while weekends have lower activity (figure 2.9) being this greater at late afternoons and evenings.

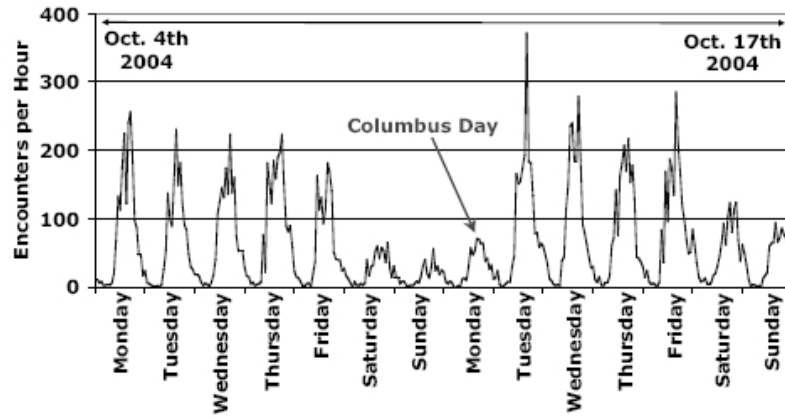


Figure 2.8: Pairs encounters per hour from [19]

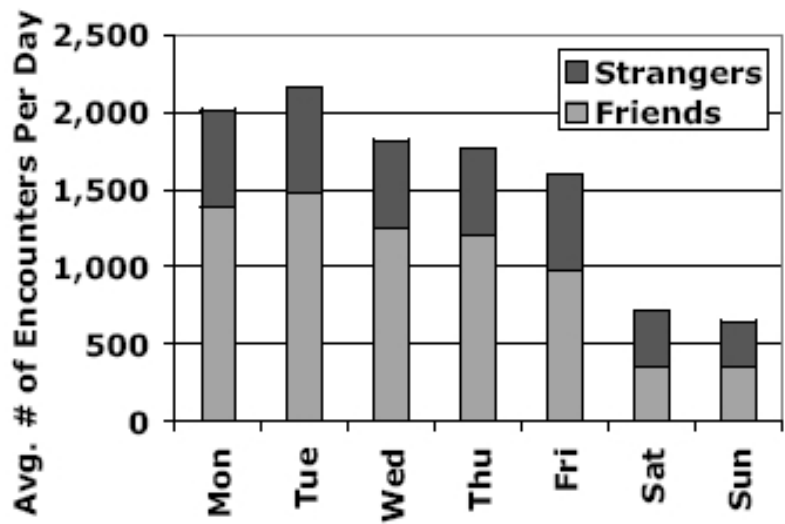


Figure 2.9: Average pairs encounters per day from [19]

With this and other analyses a simulator was created and validated showing similar trace generation (figure 2.10).

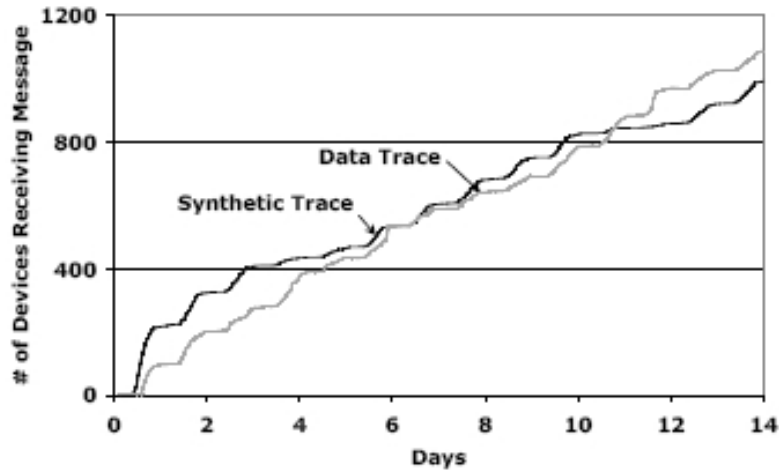


Figure 2.10: Real and synthetic trace from [19]

This simulator was used with four DTN protocols:

1. Direct contact — only forward to the final destination;
2. Forward to 1 person — the message is allowed to be forward only once;
3. Forward to 2 persons — the same as the previous but two forwards are allowed;
4. Forward to all — this scheme forward messages to all friends.

Epidemic protocol was also implemented for comparison purposes. Results show that in the presence of social information all routing protocols perform much better having "forward to all" the lowest delay, being only surpassed by Epidemic Protocol (figure 2.11).

Other analyses were done in [19] such as use social information to improve firewalls and file-sharing in P2P systems.

Another approach to the problem is absolute positioning of network nodes. This is what was tried in [21] where the encountered solution was to learn patterns in paths described by persons. Network nodes positions were determined assuming that all nodes had a global positioning system (GPS) and creating a simulation software to test the prediction technique. The learning techniques were used to build a new protocol (PGR) which was tested against other Epidemic-like protocols: Epidemic and three variations of Spray and Wait. PGR showed good efficiency results, not requiring all nodes to participate in messages delivery and delivery ratio success close but lower Epidemic's.

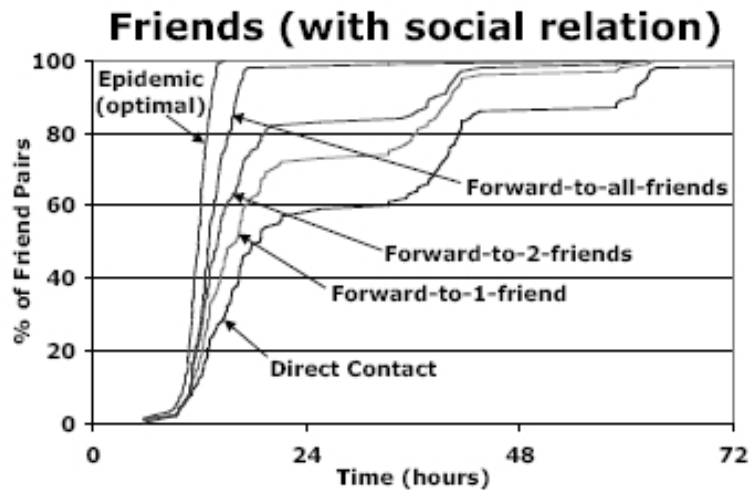


Figure 2.11: Protocol results from [19]

2.4 Other related work

A study on the influence of network nodes movements in the network performance is done in [22]. The objective was to study the possibility to use automated nodes that would travel to the place where traffic demand is higher and act as DTN router. The analogy used was taxicabs picking messages as it were passengers.

[23] presents the generic problems associated with Delay-Tolerant Networks. Items from this work to have in mind are:

1. reliability — this is intended to guarantee that all messages will reach the destination, although it is not always possible to do that being required some form of delivery confirmation;
2. delay — this is associated to asynchronous message delivery typical in DTN and should be minimized;
3. routing — this can be opportunistic, given that it is impossible for a network node to always know the optimal route for a message.

The hardware limits imposed by the technology required to create this kind of network are analysed in [24]. This interesting analysis is directed to the DTN Reference Implementation (DTNRI or DRI) by the IRTF Delay Tolerant Research Group (DTNRG) [1] on low powered devices. It ends with the conclusion that the most restrictive piece of hardware is the CPU. Also, it was found that bundle (a message or group of messages to be delivered when two nodes meet) size increases throughput but only until a size of 25KB, after that the increase in performance being minimal (figure 2.12).

The paper ends with the following recommendations:

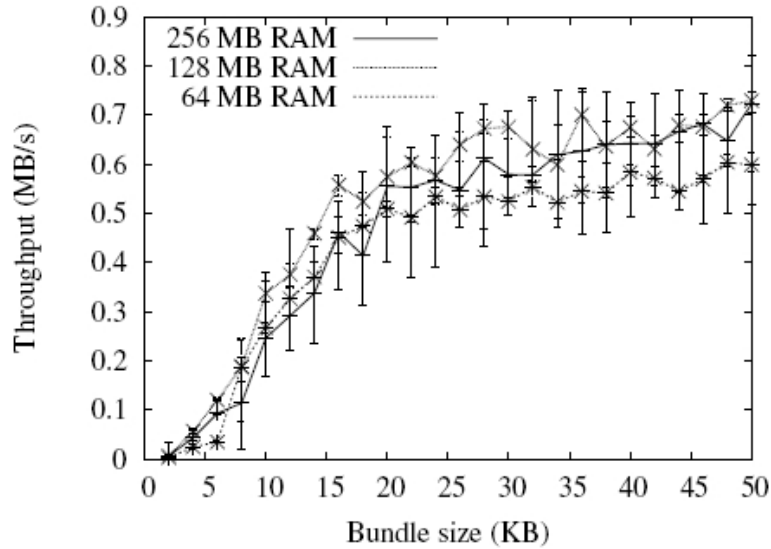


Figure 2.12: Throughput in function of bundle size from [24]

1. *DRI application developers should use the largest possible bundle size when using the DRI in-memory API;*
2. *The DRI should be restructured to increase parallelism. - use a multi-threaded router;*
3. *When deploying the DRI, invest more in the CPU than faster storage and increased memory.*

[25] is a paper about a naming architecture of a DTN. This is an important item, since we need to have some way to localize the network nodes. However this is out of the scope of this work, here we will consider that the naming problem is solved and concentrate in the routing problem.

The power management of network nodes is studied in [26]. This is closer to the physical level, although, it is still important to understand the communication delay caused by technological limits as seen in [24]. The work in [26] explores the idea of using two types of radio devices: one low powered with high range and high communication error rate for neighbourhood discovery and one with great power consumption with low range but with low communication error rate. This power management needs more information about the network behaviour so that nodes could use previous knowledge about low network activity to start power-saving modes. This hypothesis is also explored in [26] with the traffic-aware power management.

The concept of DTN Throwbox is presented in [27] and the problem of energy saving is studied in order to achieve high nodes detection while preserving battery charge. This

includes most ideas from the work presented in the last paragraph and is applied to dedicated fixed hardware — the DTN Throwboxes — built to act as additional forwarders to improve network performance.

Chapter 3

Methodology

In order to optimize routing protocols, it is necessary to study them. For that is required a simulator, adequate traffic models, networks and evaluation metrics.

On this work was considered that network nodes storage size and connection bandwidth are infinite. These are common simplifications done on previous work. Measures of used storage and bandwidth should indicate the hardware required for the DTN to work with the observed performance.

3.1 Simulator

The simulator used was DTNSim, which is part of the reference implementation DTN2 by the DTNRG [1]. The advantages include:

1. It already has protocols implemented;
2. Was built for DTN simulation, unlike other simulators targeted for general networks;
3. Implemented routers may be loaded by the DTN2 daemon to be used on real networks.

The simulations are defined in *.conf* files written in TCL scripting language [28]. A command *'sim'* is provided to create nodes and edit simulation options. Each node created has its own commands to set internal variables and execute other actions. The events written in the *.conf* file and other events launched by network nodes go into a stack of events each one with a time associated to it. Therefore, the simulator is *event-based*, having a timer that triggers the event when its time is reached.

The core of DTN2 daemon is the class *BundleDaemon*, which loads the necessary classes for the selected router to work. DTNSim creates a *Node* object that descends from the *BundleDaemon*. The router object integration within the *Node* object is similar to what is done by the *BundleDaemon*. This way, any router created for DTNSim can be

used on DTN2 daemon. However the opposite is not always true and that is why not all protocols provided by DTN2 did work with DTNSim.

Figure 3.1 shows classes hierarchy, the ‘my’ namespace was created for new protocols implemented during this work.

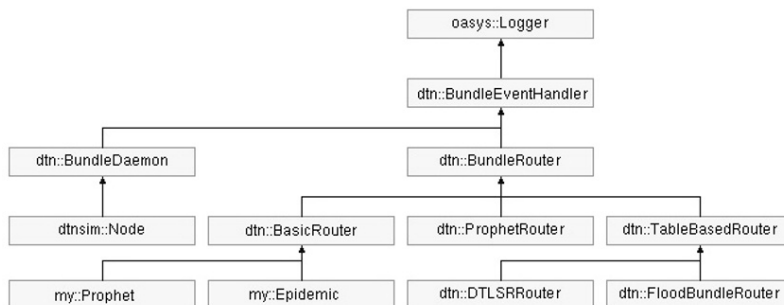


Figure 3.1: DTN2 classes

DTN2 uses *Singleton*, a class that can be instantiated only once, as superclass for various classes such as *BundleDaemon*, *BundleStore* which is used for bundle storage, and *RegistrationStore* for endpoints registration. Using Singleton objects ensures that critical resources will not be duplicated but, for simulation, such resources have to be duplicated which raises a problem.

When a new DTN2 daemon is created, a *BundleDaemon* object is created (among other objects) which owns a *BundleRouter* object. When DTNSim runs, several nodes are created and each is a *BundleDaemon* that behaves like a real DTN2 daemon running on a separated machine. Because the *BundleDaemon* class is a *Singleton* nodes cannot run in parallel, which prohibits the usage of threads, and each node have to *force* their *BundleDaemon* instance into the *BundleDaemon* class when its events are to be processed.

3.1.1 Addresses

A DTN node is identified by an ‘*eid*’ and may have several ‘*endpoints*’. The *eid* is an identifier, similar to the name given to nodes on other types of networks. An *eid* looks like a web address and starts with the protocol used (*dtn*), then ‘*://*’ and after that is placed the node name. An example of an *eid* is ‘*dtn://alice*’.

Other types of networks associate addresses to connections. Endpoints are like those addresses but, since on a DTN connections are intermittent, endpoints are associated to the data flow. The endpoint address is composed by the *eid*, followed by ‘*/*’, followed by the endpoint name. An example of an endpoint address is ‘*dtn://alice/work*’. If another node ‘*bob*’ tries to send a message to ‘*alice*’ it must define a source endpoint and a destination endpoint, for example: from ‘*dtn://bob/home*’, to ‘*dtn://alice/work*’.

3.1.2 Links and Connections

DTNSim nodes, like other types of network nodes, are required to have links in order to establish connections. Each link is associated to a pair of nodes, identified by their eids. Links are created with a command *'link'* that may also be used to set its availability and other parameters. A global command *'conn'* is provided by the simulator to set connections between two nodes up and down.

3.1.3 Messages

A network is useless if there is no messages traveling through it. DTNSim provides a command *'tragent'* to generate messages and set its size, source, destination and other options. Each generated message had a fixed size of one thousand of bytes (1KB).

3.2 Traffic models

On this work was considered that the simulated DTN will be used for Internet access. Therefore, on the simulation scenarios there are many movable nodes and one fixed access point. A real network may have multiple access points, but the delay to communicate through the Internet is much lower than that of a DTN so was considered that all access points behave as a single access point.

The frequency of messages and their size are more difficult to simulate [29]. Because the evaluation requires an environment with control over messages, a *'step'* traffic model was used instead a more realistic traffic model.

From systems control theory, an impulse is a burst of energy with infinite amplitude and zero duration. In practice this means a burst with maximum energy possible and the smallest duration possible. On a network this means all nodes sending one message at the same time. Again, from the control theory a step is an array of impulses.

Using the step traffic model means that from time to time all nodes will send a message to the access point. The network performance to this type of input is the system's step response. A correct selection of time between each impulse may allow more or less path exploring, however this was not tested and one impulse by hour was used.

3.3 Networks

This work used real data for network representation on DTNSim. A realistic network movement is important because, since we are not restricting buffer size, the protocol is influenced by nodes connections not their traffic demand. If each node has infinite buffer, the quantity of messages to deliver is not important. What matters are the paths available to deliver those messages. If the network node is not realistic those paths may not exist in

realistic quantities. Also, each society is different which leads to different results on other sets of traces. But the use of a realistic trace of connections should make results much closer to what is expected to obtain in reality than artificial networks would have. Two sources of network data were used.

3.3.1 UMass

Several sets of traces of this network are available at [12]. The data used was collected during the Spring of 2007 and includes two sets of traces: the one called ‘DieselNet - Spring 2007’ and ‘DieselNet - Access Point Connectivity’. The three days traces of access point connectivity were combined with the Spring 2007 traces and those days (26, 27 and 28 of March 2007) were used to create a representation of the UMass network with its buses and access points. The text files downloaded from the web site were parsed and the data was stored into a MySQL database. The total number of contacts is 2.594, table 3.1 shows the number of contacts by day and the number of buses on the same day.

Table 3.1: Umass network contacts and nodes

Day	Contacts	Nodes
1	682	25
2	1.140	29
3	772	23

3.3.2 Reality Mining

The Reality Mining database, from [20] has a total of 2.815.196 contacts. The contacts table was built with the data of connections between people’s devices, and the data of connections between people’s devices with cell towers. Cell towers were used as access points for this network. Table 3.2 shows the number of contacts and nodes for all months on the database. Table 3.3 shows the contacts and nodes by day for the first ten days of a randomly selected month (October of 2004).

The selected month was chosen among the months with greater contacts and nodes, months 09/04, 10/04, 11/04, 12/04, 01/05, 02/05 and 03/05. This is necessary for representation of a network on its activity peak, despite being a disadvantage when we consider the real simulation time required to simulate such high number of nodes and contacts.

3.4 Evaluation metrics

In order to correctly measure performance, evaluation metrics related to performance are also required. The metrics measured were the total number of arrivals (Arrivals), the average per hour storage occupancy (Buffer) and the average per hour transmissions (Xmit). Both buffer and xmit metrics are measured in thousands of bytes. Another metric was calculated from the three metrics referred: cost per arrival.

Table 3.2: Reality Mining network months

Month/Year	Contacts	Nodes
01/2004	1.035	83
03/2004	1.317	31
07/2004	15.586	211
08/2004	90.300	974
09/2004	331.078	2.351
10/2004	581.047	4.006
11/2004	585.709	5.496
12/2004	448.256	4.013
01/2005	304.046	4.082
02/2005	183.376	2.328
03/2005	173.441	3.434
04/2005	89.356	1.339
05/2005	10.649	275

3.4.1 Arrivals

This metric is directly measured from the simulation log as provided by DTNSim. It is obtained adding all arrivals that occurred each hour, registered as an event ‘ARR’ on the log. The number of arrivals by hour is then added to obtain the total amount of arrivals until each hour. The measured metric is the last (or bigger) value of that list.

3.4.2 Buffer

Values for this metric are obtained from the simulator log, but changes to DTNSim’s code were required to obtain these values. At each hour, each node outputs the number of bytes it has in its buffer. Adding all node’s buffers grouped by hour gives the buffer occupancy by hour. The average of those values is the measured metric.

3.4.3 Xmit

Values for this metric are obtained from the simulator log without any changes to the code. Each time a message is sent to another node, an event ‘XMIT’ is recorded. The log line includes the size of that message. The sum of those values grouped by hour gives the transmitted bytes by hour. The average of those values is the measured metric.

3.4.4 Cost

Different protocols may have very different values of arrivals, buffer and xmit. For example, protocol X may have 100 arrivals with 100 buffer and 100 xmit, or (100 100 100). If node Y has (10 10 10) which one is better? X should be better because it has bigger delivery. But Y also should be better because it has lower buffer and xmit. The solution to solve this problem is the cost per arrival. This metric is an attempt to define how much we

Table 3.3: Reality Mining network contacts and nodes

Day	Contacts	Nodes
1	17.747	349
2	14.629	259
3	15.712	273
4	17.836	272
5	19.948	301
6	18.234	309
7	18.479	329
8	19.817	387
9	16.506	311
10	19.043	249

have to pay with bytes to have a certain number of arrivals. The cost is defined by the expression:

$$cost = \frac{bytes}{arrivals} \quad (3.1)$$

where ‘bytes’ represents one of the two metrics measured in bytes.

On the given example, X has cost:

$$cost_{buffer} = cost_{xmit} = \frac{100}{100} = 1 \quad (3.2)$$

and Y has

$$cost_{buffer} = cost_{xmit} = \frac{10}{10} = 1 \quad (3.3)$$

So, cost of X is equal to cost of Y, which means that both use the same amount of bytes to deliver its messages. If X’s cost were bigger than Y’s cost we might conclude that X delivers more messages and use more resources to do that. Therefore, the meaning of this metric is related to the value of arrivals. The table 3.4 shows that relation when trying to see if X is better, worse or equal to Y.

Table 3.4: Cost and its relation with arrivals

Cost of X vs Y	Arrivals of X vs Y	Classification X vs Y
lower	bigger	better
lower	lower	better
lower	equal	better
bigger	bigger	worse
bigger	lower	worse
bigger	equal	worse
equal	bigger	better
equal	lower	worse
equal	equal	equal

Table 3.4 may be simplified. In the table 3.5 ‘-’ means ‘bigger, lower or equal’.

Table 3.5: Cost and its relation with arrivals, simplified table

Cost of X vs Y	Arrivals of X vs Y	Classification X vs Y
lower	-	better
bigger	-	worse
equal	bigger	better
equal	lower	worse
equal	equal	equal

3.4.5 Metrics discussion

Why to use these metrics? The first three are common metrics used on previous work and are logical choices. A network provider would like to know how much of its clients data is arriving at the destination. Buffer size control is also important because a delay tolerant network is directly dependent on buffer size for message delivery. This help us to know how much storage the node equipment will need and also to choose the protocol that better use that storage. The Xmit metric serves similar purposes, it allow us to determine how much data is required to travel through the transmission medium and how much data the network node should be able to send or receive when it encounters another node.

Another metric, also used on previous work, could be included here: the average delivery delay. Waiting for a second connection before sending messages means higher delay, but if the communication medium does not have a limit of bandwidth all messages will be sent to the next hop. On a real situation, due to the bandwidth limitation, some messages would not be sent which would mean higher delay than simulations predicted. Therefore the delivery delay measures would be affected by the described situation which we want to avoid.

The cost is useful to compare two protocol with very different values on the other three metrics. Previous work show that protocols that waste more resources also deliver more messages (e.g. flooding protocols). Different attempts to minimize the resources waste resulted into protocols that deliver less messages. But if a protocol delivers less messages and has lower buffer and xmit, how can we say if it is better? If a protocol delivers less messages, obviously less resources should be wasted. That is what Cost was created for. To measure how much bytes we saved with that messages that were not delivered. Therefore, Cost is useful when we want to objectively compare two protocols having different values of Arrivals, Buffer and Xmit.

3.5 Simulations

All protocols were simulated on all days of UMass network and October 2004 of Reality Mining Network. Since Reality Mining network has a greater number of nodes and contacts each day, the simulations were restricted to a fraction of the day, from 06:00 to 12:00 am. For consistency, the same reduction of time was applied on UMass network.

Because simulations included only a part of the day, the number of contacts occurring on the simulation are different. Tables 3.6 and 3.7 show the number of nodes on all day and the contacts that occurred between 6:00 and 12:00 am.

Table 3.6: UMass network contacts and nodes, between 6 and 12 am

Day	Contacts	Nodes
1	99	25
2	351	29
3	313	23

Table 3.7: Reality Mining network contacts and nodes, between 6 and 12 am

Day	Contacts	Nodes
1	4.530	349
2	2.663	259
3	3.516	273
4	5.280	272
5	5.676	301
6	4.747	309
7	5.026	329
8	4.867	387
9	3.213	311
10	4.423	249

3.6 Work Organization

This work's first step was to simulate all protocols on UMass network to study its behaviours and decide the best two protocols. Then, the best two protocols were simulated on Reality Mining network which reduced the real simulation time required.

After the protocols comparison, protocol's algorithm optimizations were tested and will be presented in chapter 5.

Finally, the chosen protocols were be combined and the resulting hybrid protocol, its simulations and performance will be presented in chapter 6.

The results are usually organized in two groups of charts: arrivals, buffer and xmit metrics into one chart; buffer and xmit costs in another chart. The term *ratio* is associated

to direct division of one metric by another. For example, consider protocols X and Y; then *X's Y arrivals ratio* would be: $\frac{arrivals_X}{arrivals_Y}$.

The term *relative* is associated to the division of one metric by another minus 1. For example, again with protocols X and Y; then *X's arrivals relative to Y* would be: $\frac{arrivals_X}{arrivals_Y} - 1$.

Ratio results are used to report values in tables, relative values are used to better visualize results differences.

Chapter 4

Comparison Protocols

The DTN2 framework, the official version (2.5.0) when this work started, includes three protocol implementations: Flood [2], DTLSR [10] and PROPHET [4]. However, some issues were detected:

1. **DTLSR accepts message duplicates** — sometimes it accepts more than one copy of the same message, causing duplicate events of arrivals;
2. **PROPHET does not work with DTNSim** — therefore it was impossible to work with the version implemented in DTN2;
3. **Flood is too simple** — it works as expected but has no intelligence. For example, the message destination keeps sending the message after receiving it.

To solve the enumerated issues, changes were necessary. DTLSR was fixed in order to stop accepting more than one copy of the same message. An interface, BasicRouter, was created to hide the details of low level operations from the implemented protocols and was used as basis for two protocol implementations that were created: another version of PROPHET protocol and the Epidemic protocol.

Another issue is the real simulation time. Each implementation has its own time, the Epidemic is the fastest and DTLSR the slowest. Since UMass network has less nodes and contacts, each protocol was first tested on it to decide which protocols should be used on the next phase of this work.

4.1 Epidemic

This protocol is an improvement over the simple Flood protocol included with DTN2. Both belong to a type of DTN protocols designated as ‘flooding’ or ‘aggressive’ routing. As previously said in this work (section 2.1) the idea is to copy all messages to all nodes met. However, the protocol that comes with DTN2 is too simple: it copies all messages in its buffer without knowing if the other node already have them, which wastes bandwidth.

Also, on Flood protocol the following situation was found: if a node \mathcal{A} has a message to \mathcal{B} , when \mathcal{B} receives the message it continues sending the message to other nodes.

The new protocol, called Epidemic, has features that allow it to decrease resources waste. The active receipt scheme [3] allows the network node, using Epidemic protocol, to send messages only when the node met does not have copies of those messages and stop copying them after being received. An example of the differences may be seen in figure 4.1. The flood protocol has higher xmit and buffer, while the Epidemic has lower values for those metrics.

From chapter 2, figure 2.1 shows the number copies of a flood protocol with several control schemes compared against the simpler version. As we may see on that figure, the addition of message confirmation reduces the number of messages by about a half. The same may be observed on figure's 4.1 buffer bars, adding message confirmation reduces the amount of messages in buffer by a half.

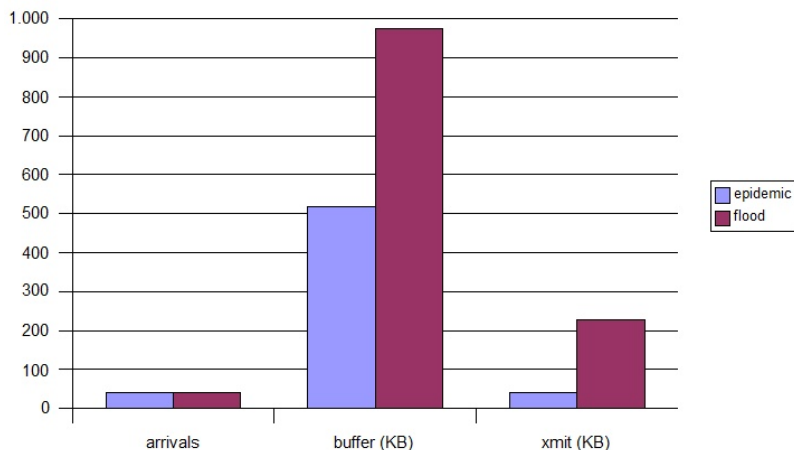


Figure 4.1: Epidemic vs Flood, day 1 of UMass network

4.1.1 Results

Here are presented the results on Epidemic simulations, figure 4.2 shows the arrivals, buffer and xmit metric for the first three days on the UMass network.

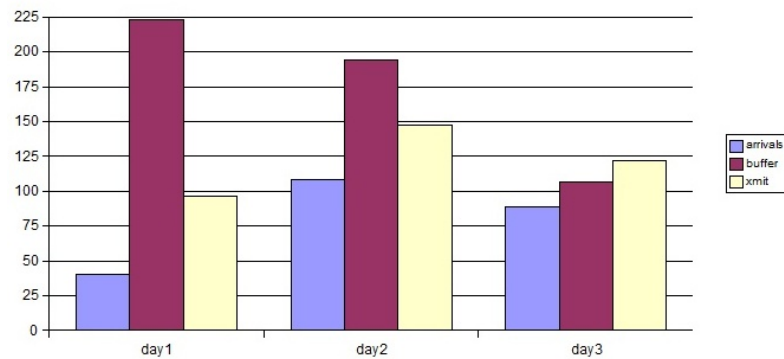


Figure 4.2: Epidemic, Umass network

Figure 4.3 shows the cost relative to buffer and xmit on the same network and days.

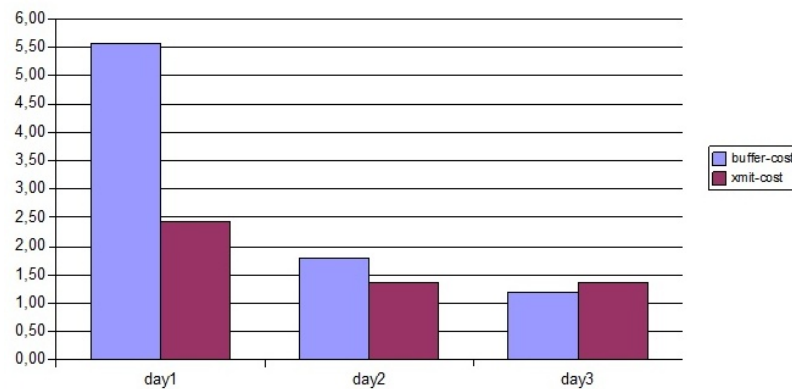


Figure 4.3: Epidemic, Umass network

4.1.2 Discussion

Day 1 has less contacts (table 3.6) than other days and the highest buffer and the lower xmit and arrivals. On day 2 the number of contacts and nodes increases and Epidemic's arrivals and transmissions increase as well while the buffer decreases.

This indicates that xmit and arrivals are directly related to nodes and contacts, and that buffer has an inverse relation to nodes and contacts. This makes sense, since a large number of nodes and contacts should increase network connectivity leading to more arrivals and xmit. The buffer is reduced by confirmation of delivered messages, so the increase of connectivity should decrease the buffer. However, buffer values should stabilize at some point so the relation should have an inverse component and a direct component, both related to the number of nodes and contacts.

On day 3 contacts and nodes are again lower than day 2. The protocol shows lower arrivals and xmit as expected. It is also the day with lower costs.

Based on the results obtained we conclude that Epidemic protocol has lower waste of resources on networks with high number of contacts and lower number of nodes.

4.2 DTLSR

Changes done to DTLSR were simple, it was only necessary to add a cache that keeps track of all messages delivered. When a message reaches its destination it can no longer be delivered to any node and is deleted from buffer. Unlike the other two protocols, DTLSR does not check what messages the contacted node has before forwarding its messages, which should increase xmit. DTLSR protocol uses link-state advertisements (LSA) to build its routing table. Two consequences result from the fact that the number of LSA is directly related to the number of network nodes: increased xmit and increased real simulation time.

4.2.1 Results

Figure 4.4 shows the arrivals, buffer and xmit metric for the first three days on the UMass network.

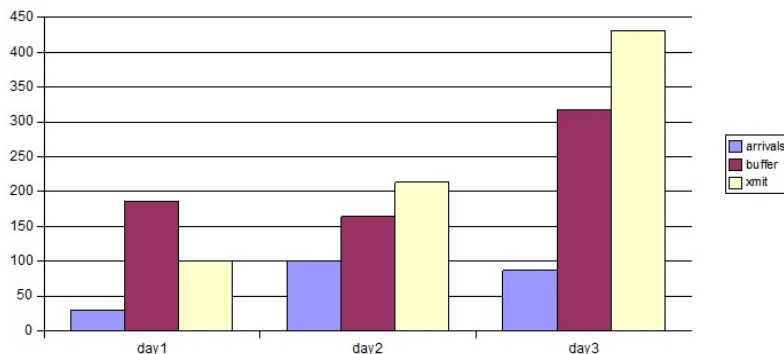


Figure 4.4: DTLSR, UMass network

Figure 4.5 shows the cost relative to buffer and xmit on the same network and days.

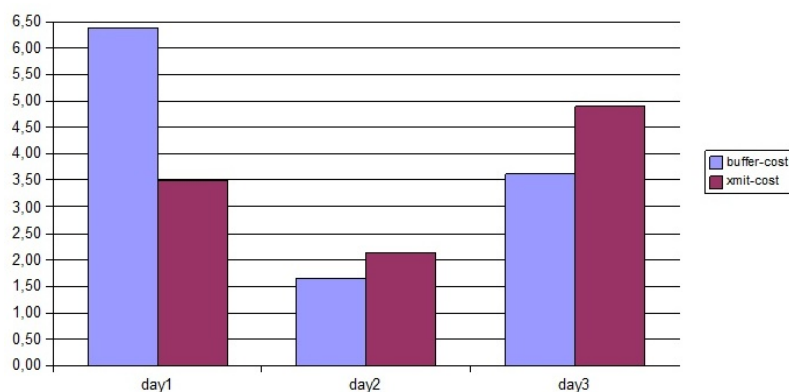


Figure 4.5: DTLSR, UMass network

4.2.2 Discussion

Day 1 have lower number of nodes and contacts than day 2 and its arrivals and transmissions are lower than day 2. But on day 3 arrivals are a little lower while buffer and transmissions are bigger. Since day 2 and day 3 have similar amount of contacts, this difference should be caused by the number of nodes.

Based on the results obtained we conclude that DTLSR has better performance (lower costs) on networks with high number of contacts and high number of nodes. This is surprising, since this protocol was presented as a solution for low mobility schemes [10].

4.3 PROPHET

The PROPHET version implemented is based on the Internet draft released by the protocol authors on February 11, 2008 [30]. The standard equations for predictability used are:

1. **update** — $P(A, B) = P(A, B)_{old} + (1 - P(A, B)_{old}) * P_{encounter}$
2. **aging** — $P(A, B) = P(A, B)_{old} * \gamma^K$
3. **transitivity** — $P(A, C) = P(A, C)_{old} + (1 - P(A, C)_{old}) * P(A, B) * P(B, C) * \beta$

Where $P_{encounter}$, γ and β are parameters which values are suggested on the draft. $P(A, B)$ is the predictability of delivery from A to B and K is the time since the predictability was aged.

The Internet draft includes several forwarding strategies:

- **GRTR** — Forward the message only if $P(A, C) > P(B, C)$;
- **GTMX** — Forward the message only if $P(A, C) > P(B, C)$ and $NF < NF_{max}$, where NF_{max} is a maximum number of nodes to which the message should be sent;

- **GRTR+** — Forward the message only if $P(A, C) > P(B, C)$ and $P(A, C) > P_{max}$, where P_{max} is the largest delivery predictability of any node it has forwarded the message to;
- **GTMX+** — Forward the message only if $P(A, C) > P(B, C)$ and $P(A, C) > P_{max}$ and $NF < NF_{max}$, which combines the last two;
- **GRTRSort** — Forward the message only if $P(A, C) > P(B, C)$, after sorting them by descending order of $P(A, C) - P(B, C)$;
- **GRTRMax** — Forward the message only if $P(A, C) > P(B, C)$, after sorting them by descending order of $P(A, C)$.

Also, various queuing policies were possible:

- **FIFO** — first message in, first message out;
- **MOFO** — most forwarded message first;
- **MOPR** — most favorably forwarded message first;
- **Linear MOPR** — most favorably forwarded first;
- **SHLI** — shortest life time first;
- **LEPR** — least probable first.

The forwarding strategy implemented was GRTR and the queuing policy implemented was SHLI.

The implemented version uses only one phase to deliver information about its routes, confirmed and buffered bundles, instead of the complex system described on the draft.

4.3.1 Results

Figure 4.6 shows the arrivals, buffer and xmit metric for the first three days on the UMass network.

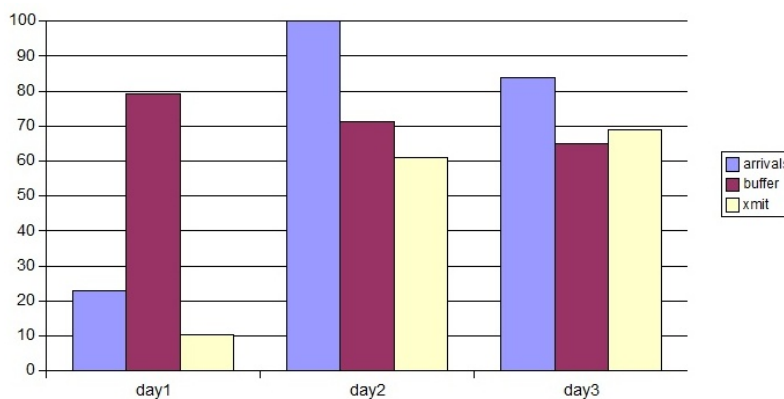


Figure 4.6: PROPHET, UMass network

Figure 4.7 shows the cost relative to buffer and xmit on the same network and days.

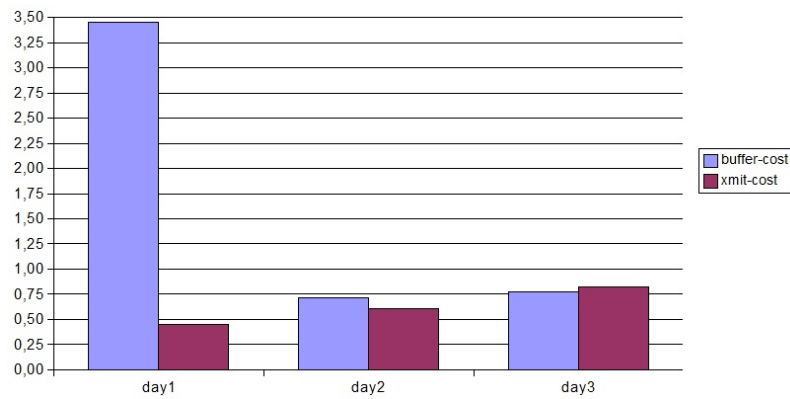


Figure 4.7: PROPHET, UMass network

4.3.2 Discussion

Day 1 indicates that nodes are encountering but not enough times to successfully deliver messages. Therefore messages can't find their destinations and are stored, which makes the buffer grow while arrivals and xmit are small. On day 2 the number of contacts are much bigger and this is observed on the arrivals. On day 3 the contacts are a little lower and we may see that arrivals are lower too. The differences should be mainly caused by the number of nodes and the encounter pattern.

Based on this results we conclude that PROPHET performs very well on networks with high number of contacts and nodes being the former more decisive to performance.

4.4 Comparison — UMass

Figures 4.8, 4.9 and 4.10 show the three days of UMass network. Each chart contains the three basic metrics (arrivals, buffer, xmit) for each protocol.

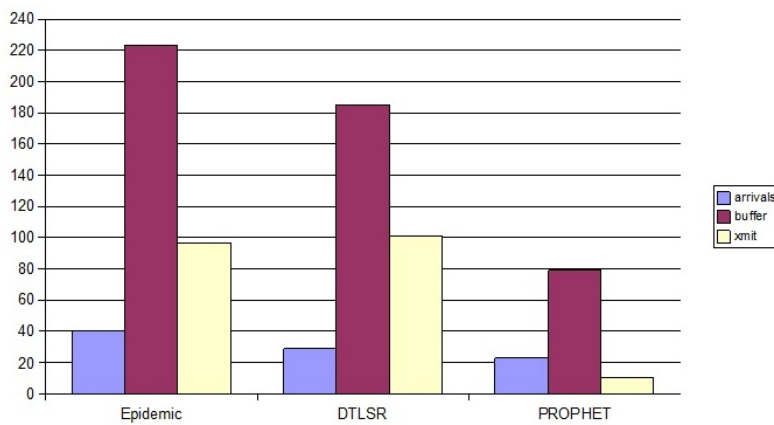


Figure 4.8: Comparison, day 1 of UMass network

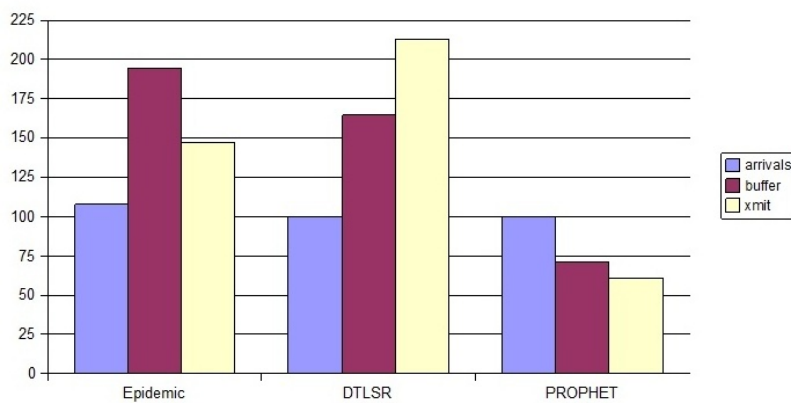


Figure 4.9: Comparison, day 2 of UMass network

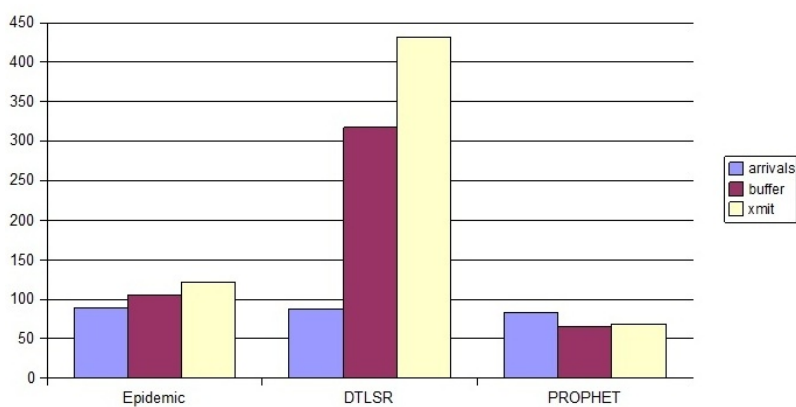


Figure 4.10: Comparison, day 3 of UMass network

Notice that on all figures DTLSR has higher xmit than Epidemic. This happens because with connections constantly changing, additional LSA messages are generated to report this to the network.

Epidemic protocol have higher arrivals than the others with PROPHET and DTLSR being very close to each other. On day 3 DTLSR has higher buffer than Epidemic; this is caused by the fact that Epidemic have messages confirmation and DTLSR does not.

Figures 4.11, 4.12 and 4.13 show all costs from each day. Notice that while day 2 with the highest amount of contacts and nodes is the best DTLSR day, it is still worse than Epidemic.

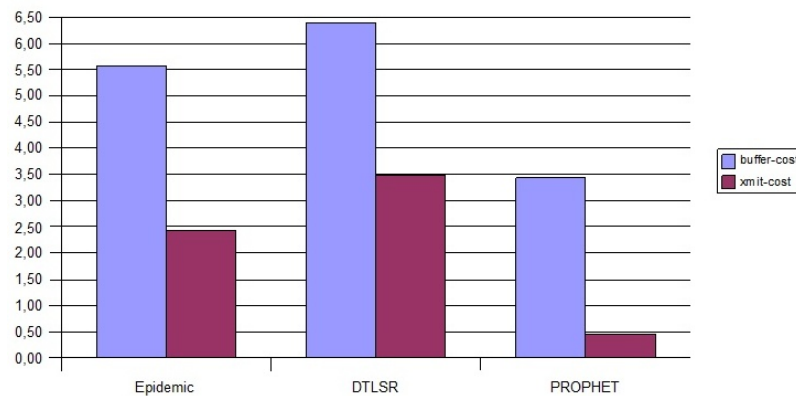


Figure 4.11: Comparison, day 1 of UMass network: costs

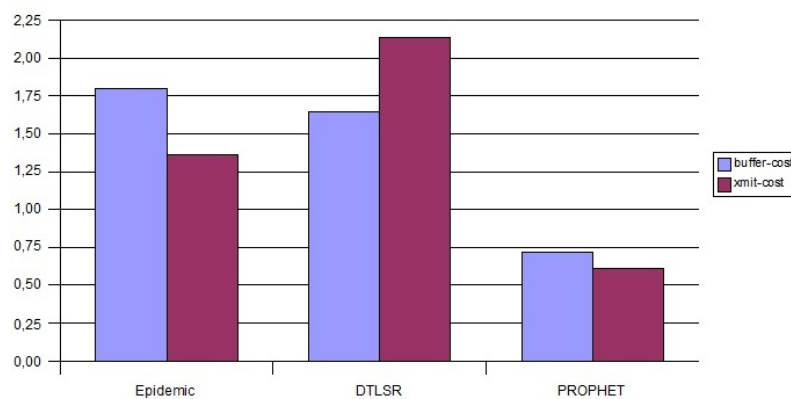


Figure 4.12: Comparison, day 2 of UMass network: costs

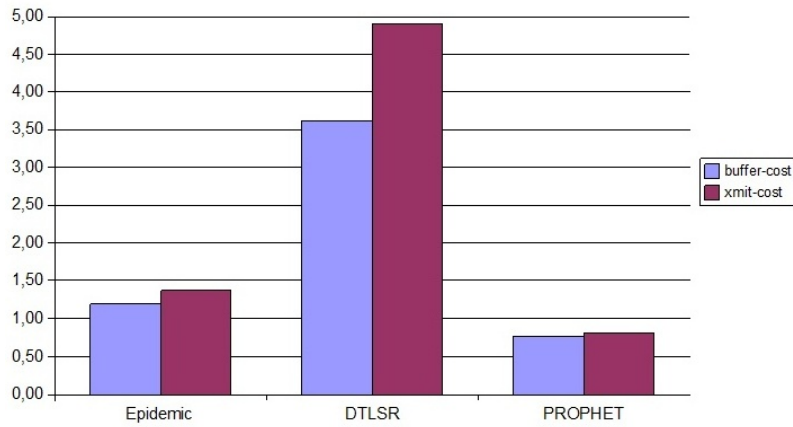


Figure 4.13: Comparison, day 3 of UMass network: costs

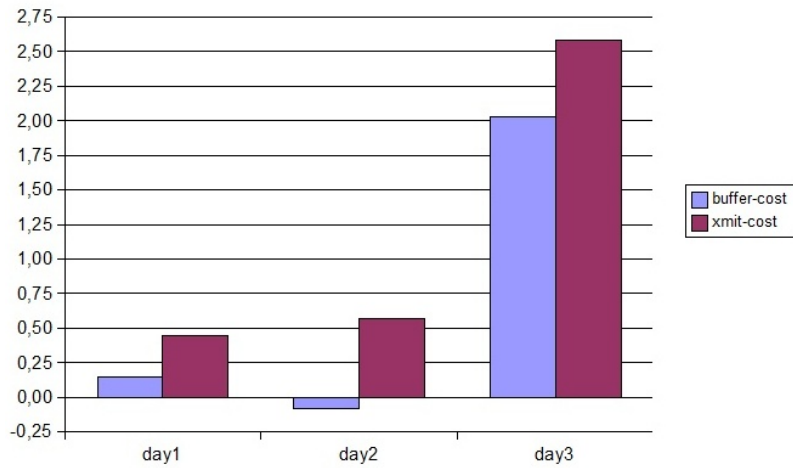


Figure 4.14: Comparison, UMass network: DTLSR's costs relative to Epidemic

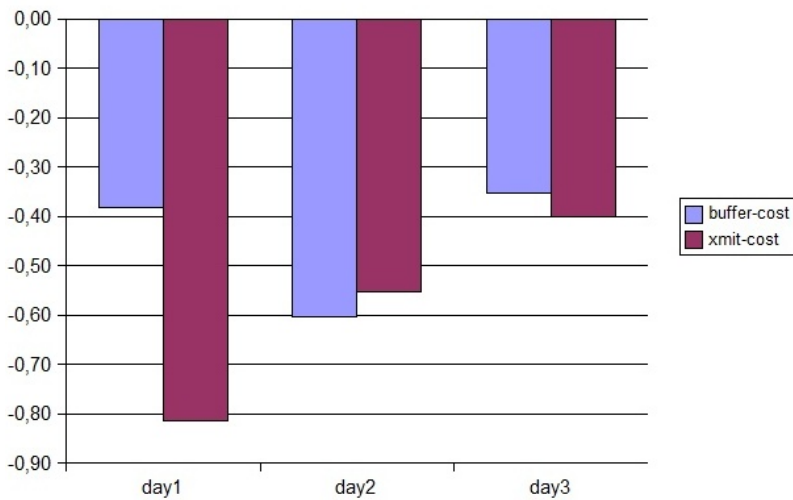


Figure 4.15: Comparison, UMass network: PROPHET's costs relative to Epidemic

As figures 4.14 and 4.15 show, DTLSR wastes more resources than Epidemic and PROPHET.

Based on this results Epidemic and PROPHET were chosen for the next step: simulate on the Reality Mining network. It was important to verify that DTLSR does not qualify as a good choice because otherwise the number of simulations to do with Reality Mining would be very limited, due to time restrictions for this project. The objective of these simulations was to finish characterizing the studied protocols before starting with optimization approaches.

4.5 Reality Mining Results

Simulations included the first ten days of this network. The other 21 days of the selected month were reserved for validation purposes.

4.5.1 Epidemic Results

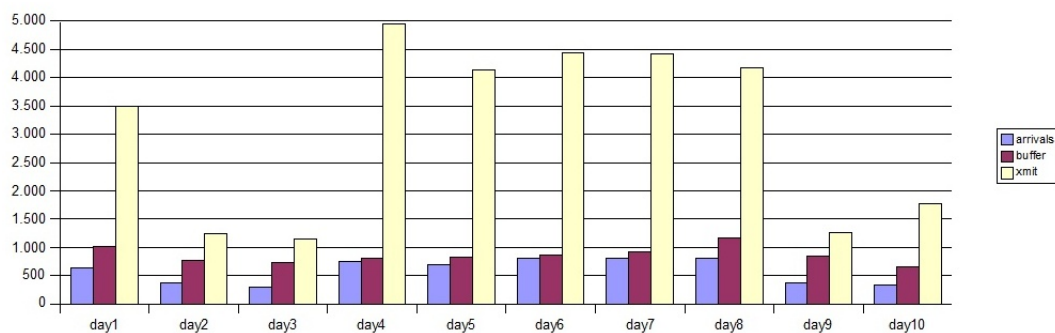


Figure 4.16: Epidemic, Reality Mining network

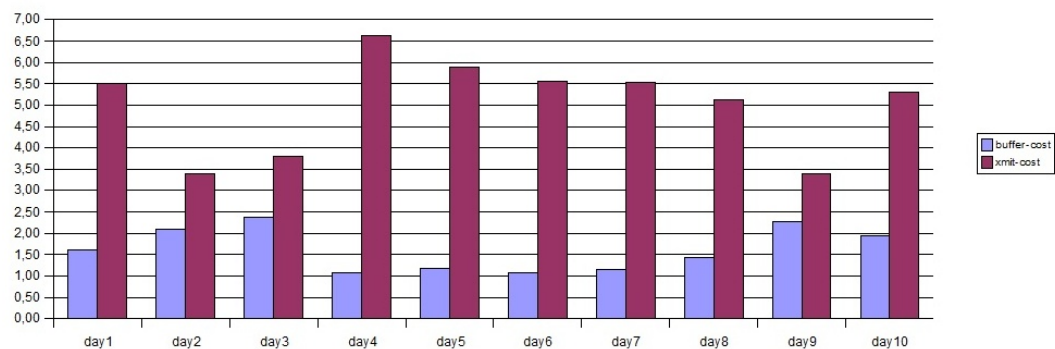


Figure 4.17: Epidemic, Reality Mining network

4.5.2 PROPHET Results

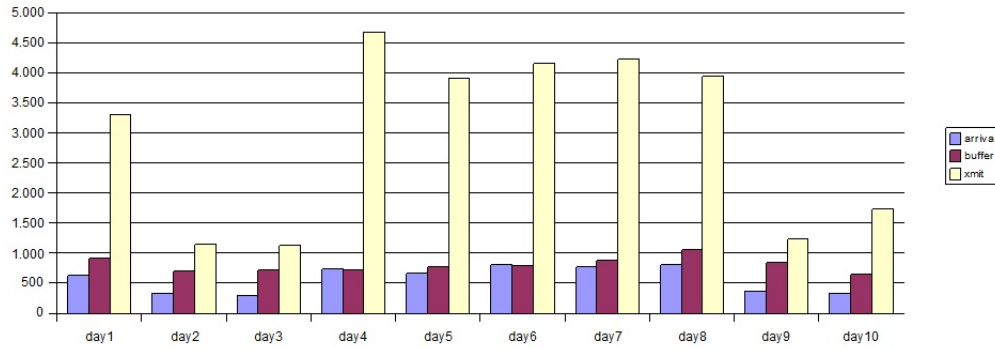


Figure 4.18: PROPHET, Reality Mining network

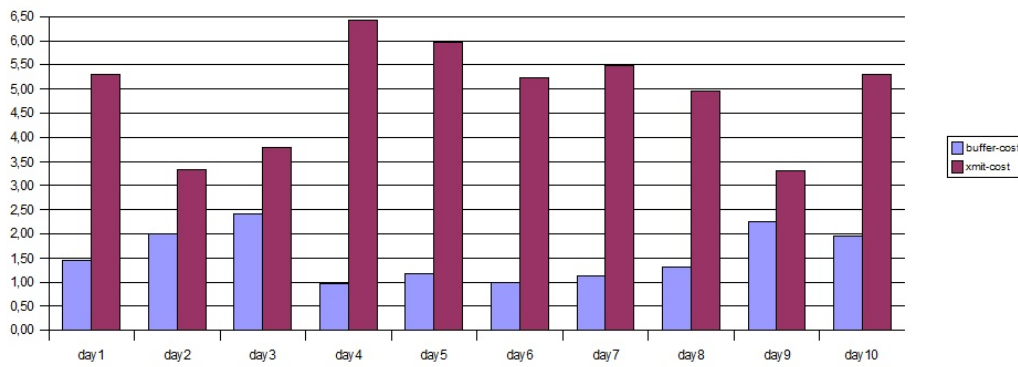


Figure 4.19: PROPHET, Reality Mining network

Figure 4.20 shows the PROPHET costs relative to Epidemic.

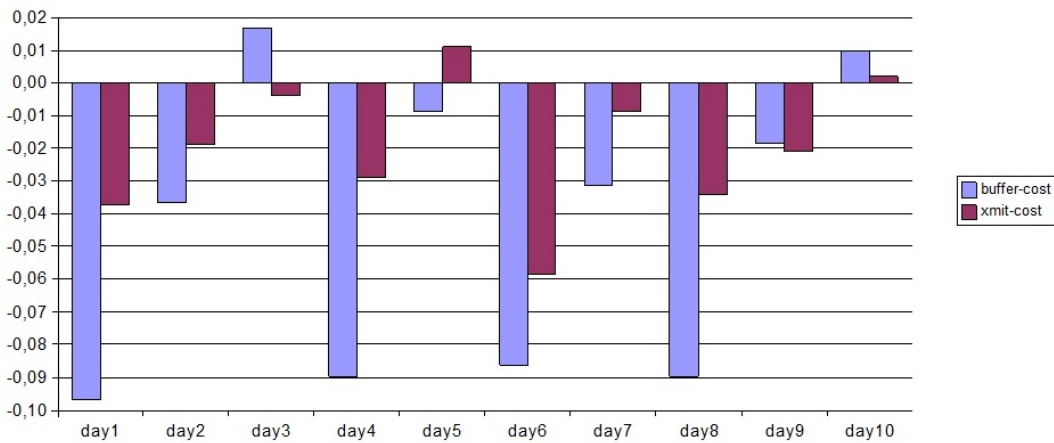


Figure 4.20: Comparison, PROPHET relative to Epidemic

4.6 Comparison — Reality Mining

Both protocols have similar results. On days with lower amount of contacts, e.g. days 2 and 3, xmit and arrivals are much lower. The buffer metric does not seem to be directly related to the number of contacts or nodes.

On figure 4.20 costs are usually negative. Since PROPHET costs are lower than Epidemic's, according to table 3.5, PROPHET is better than Epidemic. This is not verified on days 3, 5 and 10. Days 3 and 10 have in common low number of contacts and nodes which indicates that Epidemic will be better than PROPHET on this situation. Day 5 is an anomaly and the only explanation possible is the encounter pattern that for some reason did not provided the expected performance for PROPHET protocol.

Epidemic having better performance than PROPHET on situations with low number of nodes makes sense, and means that the Epidemic node will not have the same amount of opportunities to copy messages because the maximum number of messages it can copy is limited by the number of nodes it can meet. For example, imagine a node \mathcal{A} trying to deliver a message. If the total number of nodes in the network is x then, using Epidemic protocol, the message may be transmitted to $x - 2$ nodes before reaches the destination. So, for one arrival, there is $x - 2$ transmissions and $x - 2$ messages in buffer. Therefore, on this situation cost is $x - 2$. Consider now a new number of nodes y and that $y < x$, then the new cost is $y - 2$ which is lower than $x - 2$. Because lower cost is better, the conclusion is that reducing the number of nodes of a network while the number of contacts is maintained improves Epidemic's performance.

The same logic may be used to explain why PROPHET is worse with lower number of network nodes. What happens is that when we limit the number of network nodes we also limit the number of choices for the protocol to send messages. If $y < x$ then the number of nodes having bigger predictability than node \mathcal{A} does is also lower. Therefore PROPHET will perform worse on networks with low number of nodes.

Chapter 5

Changes to PROPHET

The first improvements to be presented are the changes to the PROPHET protocol. As said in chapter 4 PROPHET's Internet draft suggests several forwarding schemes. It was not required an extensive analysis of each scheme because the proposed improvements either are applied to, or may be combined with the forwarding schemes on the draft.

Each proposed improvement was simulated on UMass network and the first ten days of Reality Mining and compared with PROPHET's original version.

5.1 EQ-GRTR

This scheme is an improvement to GRTR. Recall that GRTR forwards messages if:

$$P(A, B) < P(B, C) \tag{5.1}$$

This assumption is correct, the message should be forwarded to nodes with higher predictability of delivery. This guarantees that the message will arrive at the destination with low waste of resources. The improvement proposed on this section is to change the GRTR expression to this:

$$P(A, B) \leq P(B, C) \tag{5.2}$$

Imagine that neither A nor B knows C . Applying 5.1 would mean that messages from A to C would not be forwarded. But, in the future, A or B may meet C or another node that knows C . Equation 5.2 covers that situation.

5.1.1 Results

Figures 5.1 and 5.2 show the results for this forwarding strategy. Each figure has two charts: the first shows each forwarding scheme arrivals on each day, the second shows the cost of EQ-GRTR relative to GRTR cost.

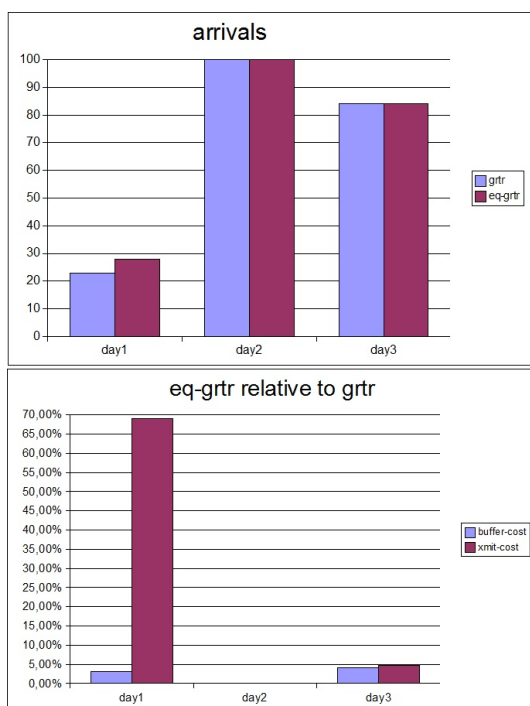


Figure 5.1: EQ-GRTR, UMass results

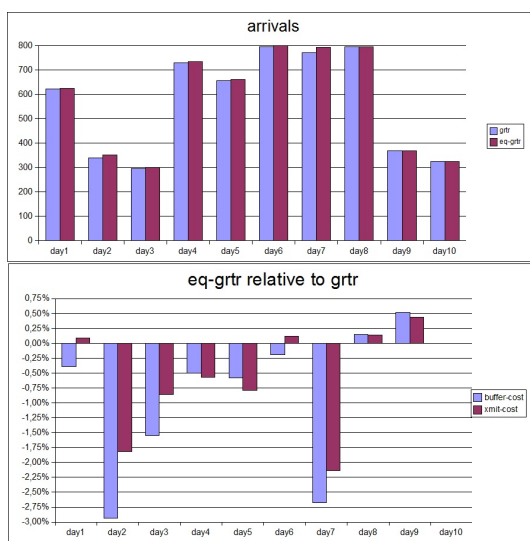


Figure 5.2: EQ-GRTR, Reality Mining results

5.1.2 Discussion

EQ-GRTR accomplishes more arrivals than GRTR on the first day of UMass network (figure 5.1). However, as we may see on the cost chart, those arrivals are too expensive. On figure 5.2 we may see that EQ-GRTR delivers more messages on 80% of the days and exhibits lower cost on 50% of the days. The days with lower costs are 2, 3 and 7 and those costs does not seem to be related to contacts nor number of nodes on those days because

days 2 and 3 are very different from day 7 (according to table 3.2).

Table 5.1: EQ-GRTR's GRTR ratio

Metric	UMass	Reality Mining
arrivals	107,25%	101,04%
buffer cost	102,45%	99,19%
xmit cost	124,59%	99,46%

From the results obtained we conclude that EQ-GRTR will have better results on Reality Mining days (table 5.1). The only reason would be the higher number of nodes and contacts but it was not possible to verify that relation.

5.2 Fast Start

The next improvement is based on the same logic of that presented on the previous section: maximize delivery. Using PROPHET's standard version, messages will stay in buffer until a node with higher predictability or the message destination is found. But, imagine a message that has an unknown destination and that destination is also unknown for the nodes met. It is more likely that such message would not be forwarded and is discarded after its life time. EQ-GRTR could be a good choice for such situations but imagine that the message is on node \mathcal{A} which in fact have route to messages's destination, node \mathcal{B} . Suppose also that \mathcal{A} has just met \mathcal{B} and therefore have higher predictability for that destination. On real situations is not natural for the probability of two meetings to be close in time, therefore waiting for that connection could be a waste of resources.

The Fast Start scheme tries to avoid this situation by creating one copy of each message that was not copied to another node yet. That is done even if the destination's predictability is not zero. This strategy guarantees that any message will have itself copied to a small number of nodes which should be enough to deliver it.

5.2.1 Results

Again we have one figure for network, with two charts per figure.

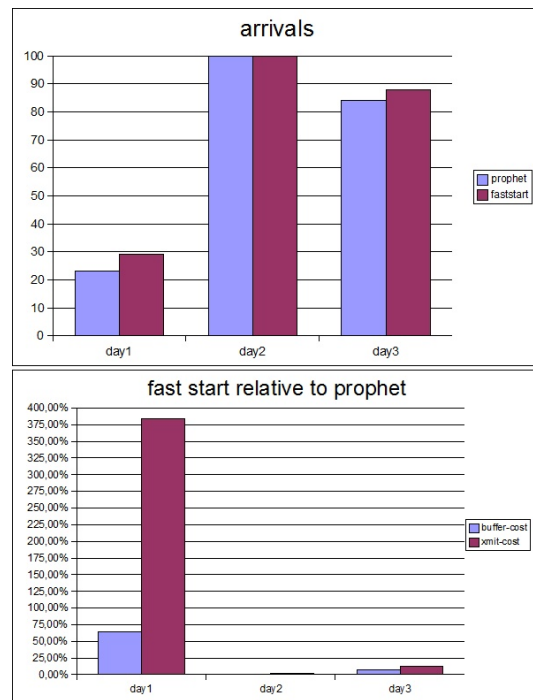


Figure 5.3: Fast Start, UMass results

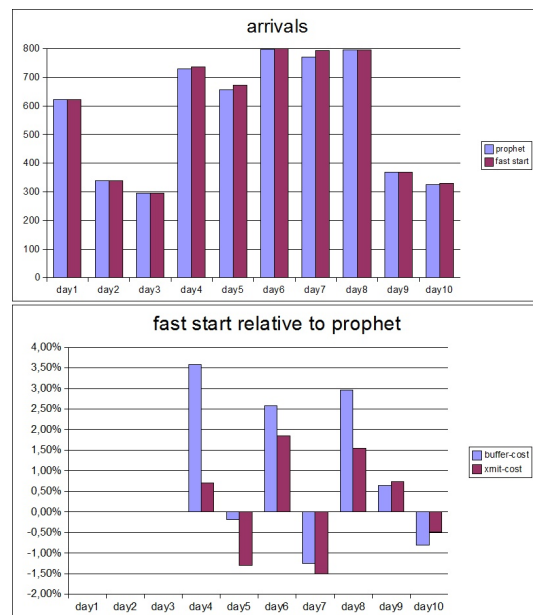


Figure 5.4: Fast Start, Reality Mining results

5.2.2 Discussion

Results are similar to EQ-GRTR, the first day of UMass network (figure 5.3) has more arrivals but also much bigger cost. The Reality Mining results (figure 5.4) are different. 30% of days have no difference of costs, 30% have lower cost and 40% have higher cost

than standard PROPHET. The table 5.2 shows the average arrivals on each network and the relative gain.

Table 5.2: Fast Start's PROPHET ratio

Metric	UMass	Reality Mining
arrivals	110,28%	100,75%
buffer cost	123,47%	100,75%
xmit cost	232,32%	100,16%

In average, Fast Start, allows more arrivals (more 0,75%) with additional xmit cost (more 0,16%) for Reality Mining network.

5.3 Estimated Delivery Confirmation (EDC)

PROPHET's dependency on probabilities opens the opportunity to estimate messages delivery. This is done on the assumption that if $P(A, B) = 1$ then forwarding the message to node B means message delivery. The problem is that it is very rare to encounter nodes with predictability equal to one. A message was considered to be delivered when the sum of all delivery predictabilities each node the message was copied to equals or exceeds 1 (equation 5.3). If that happens, the message should already been delivered and the router should stop copying it and store the message for any future encounter with the destination. If the message was not delivered, chances are that it will not be delivered due to connection number or pattern reasons.

$$\sum_i^N P(A, B_i) \geq 1 \Leftrightarrow \text{delivery}. \quad (5.3)$$

5.3.1 Results

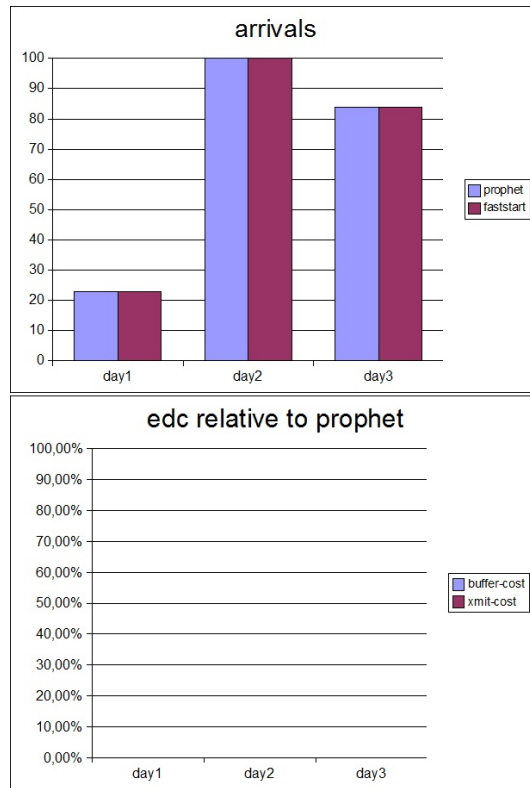


Figure 5.5: EDC, UMass results

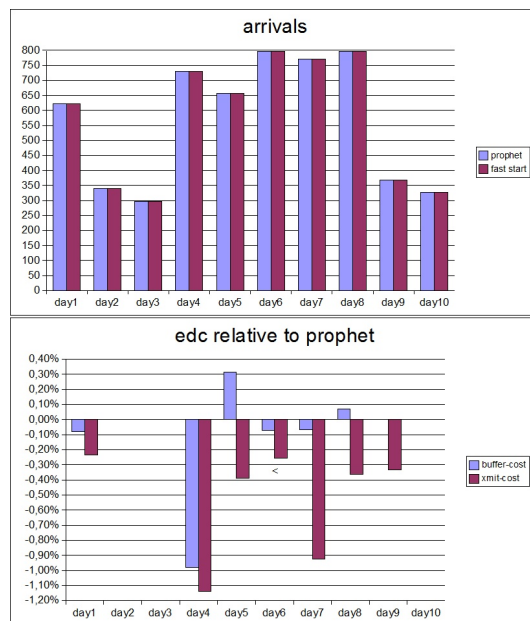


Figure 5.6: EDC, Reality Mining results

5.3.2 Discussion

UMass results (figure 5.5) are clear: the EDC scheme does not influence protocol's performance. On the Reality Mining network results are better: arrivals are not affected and costs are lower than standard PROPHEP's on most of days. Days 5 and 8 have higher buffer cost. That is an anomaly, there is nothing in the protocol's behaviour that justifies such result. Due to time restrictions, until now was not possible to determine the origin of this anomaly. Table 5.3 shows the arrivals and average costs on each network. From that table we see that EDC provides an average drop of 0,08% on PROPHEP's buffer cost and 0,36% on PROPHEP's xmit cost.

Metric	UMass	Reality Mining
arrivals	100,00%	100,00%
buffer cost	100,00%	99,92%
xmit cost	100,00%	99,64%

5.4 Conclusions

The three strategies proposed to improve PROPHEP's performance were presented. EQ-GRTR and EDC provide lower cost on Reality Mining network and Fast Start can only improve deliveries. EQ-GRTR's improvement is bi-dimensional: arrivals increases while cost decreases.

All schemes presented had better results on Reality Mining network. The reason why this happens is not completely explained. The number of contacts and nodes should influence such results. A close analysis of performance on Reality Mining days should confirm this: days with worse performance should be the days with lower number of contacts and/or nodes. However such relation was not found which indicates that performance will be much more influenced by the encounters pattern.

The improvements are small, all under or equal to 1%, but these three schemes may also be combined with each other. Also, they can be combined with the new approach presented in the next chapter.

Chapter 6

Hybrid Protocol

Using the Epidemic protocol guarantee maximum delivery possible for the network. However, we also know that we pay a high cost for those deliveries. PROPHET provide lower waste of buffer and bandwidth but also means delivery rate below the maximum.

What we want is a protocol that can use the network's full delivery potential and that also does that with the lower waste of resources possible. We want a combination of Epidemic and PROPHET.

Epidemic protocol and PROPHET have better results at opposite situations: when the number of nodes is lower Epidemic has less waste of resources and PROPHET delivers less messages. This is consistent with a real situation, e.g. at late hours of the night users wishing connectivity will have lower number of nodes to relay their messages therefore using a protocol like Epidemic that maximizes message delivery would be a better choice.

The Hybrid protocol combines Epidemic and PROPHET behaviours. When two nodes meet, the decision of forwarding all messages or a limited number of them is done based on which mode of operation the router is: Epidemic or PROPHET. Figure 6.1 shows the router architecture.

The BasicRouter interface hides the low level operations from the Hybrid router. Control messages are processed and the data to be used separated and sent to the Neighbour and Bundle list management which uses the information received to update routing tables, acknowledged bundles and neighbour lists. When all lists are updated the protocol decider chooses the type of protocol, based on information provided by network statistics module. Access to bundle lists is given to the chosen protocol which uses them to decide which bundles should be forwarded.

The number of encounters is stored by the network statistics component which then estimates the mobility factor (\mathcal{L}), measured in number of encounters per hour. If \mathcal{L} is lower than a limit ε the protocol used is Epidemic; otherwise PROPHET is used. In order

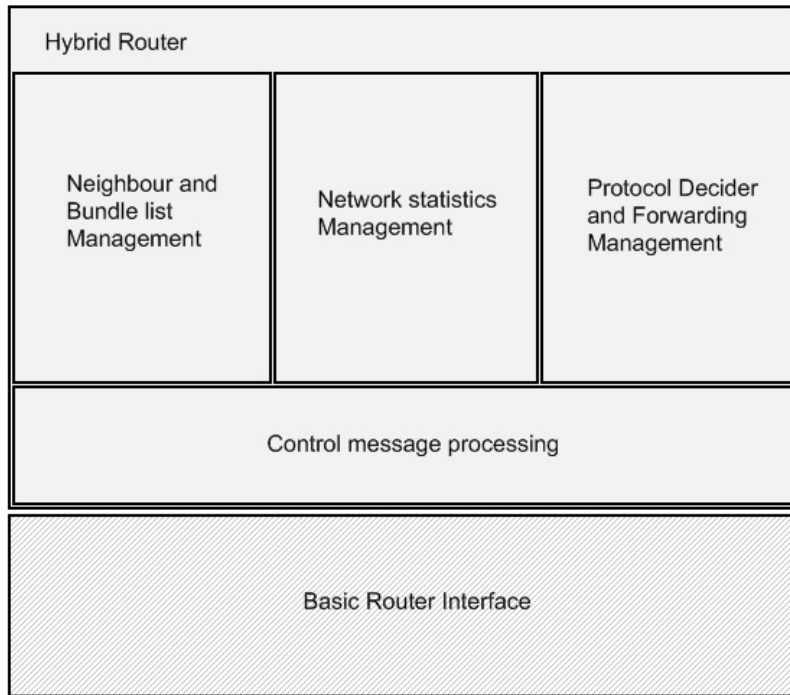


Figure 6.1: Hybrid Router

to minimize resources usage, each network node's initial state is PROPHET.

The network statistics component registers the time at which the router was started and has a counter that keeps track of the number of encounters. Every time a node is encountered, the counter is incremented and the mobility factor is calculated dividing the value in the counter by the total elapsed time .

$$\mathcal{L} < \varepsilon : \text{Epidemic} \quad (6.1)$$

$$\mathcal{L} \geq \varepsilon : \text{PROPHET} \quad (6.2)$$

6.1 Decision limit

The problem is to find each network's correct ε . Several values were experimented on both networks studied, figures in the next section show those results relative to Epidemic's results. The charts also include PROPHET's results.

6.1.1 Results

Because Epidemic protocol has higher delivery rate and cost than many protocols, results charts show almost only negative values which means that smaller arrivals bars are better and smaller cost bars are worse.

The value of ε is concatenated with the word 'hybrid', in charts legends. For example, the hybrid version for $\varepsilon = 0$ is 'hybrid0'.

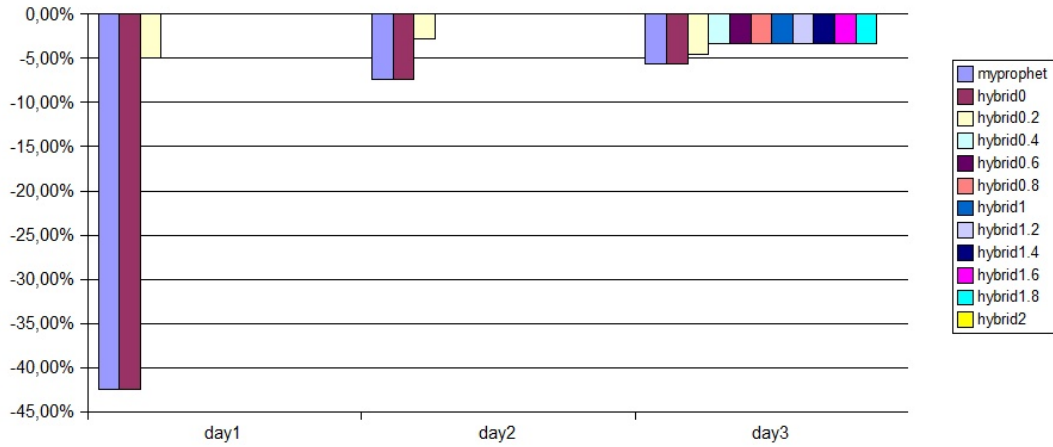


Figure 6.2: Hybrid Router, UMass arrivals

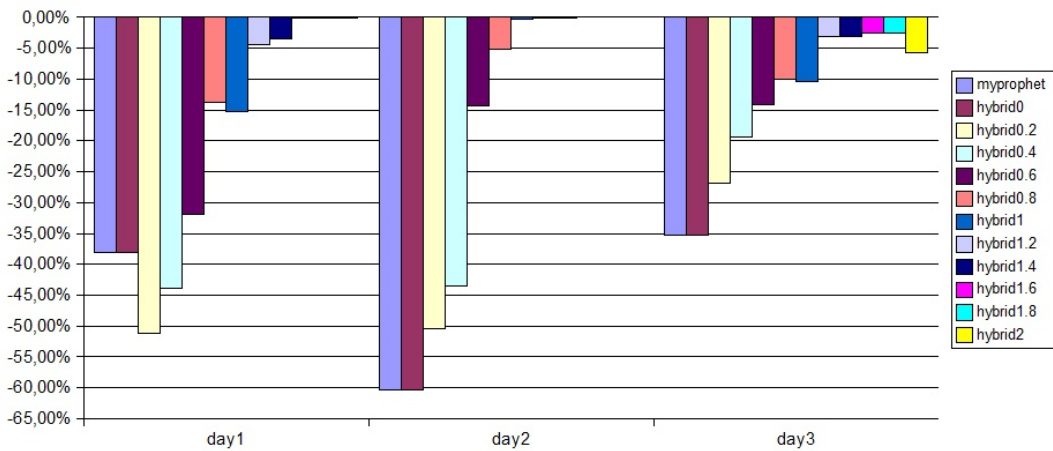


Figure 6.3: Hybrid Router, UMass buffer cost

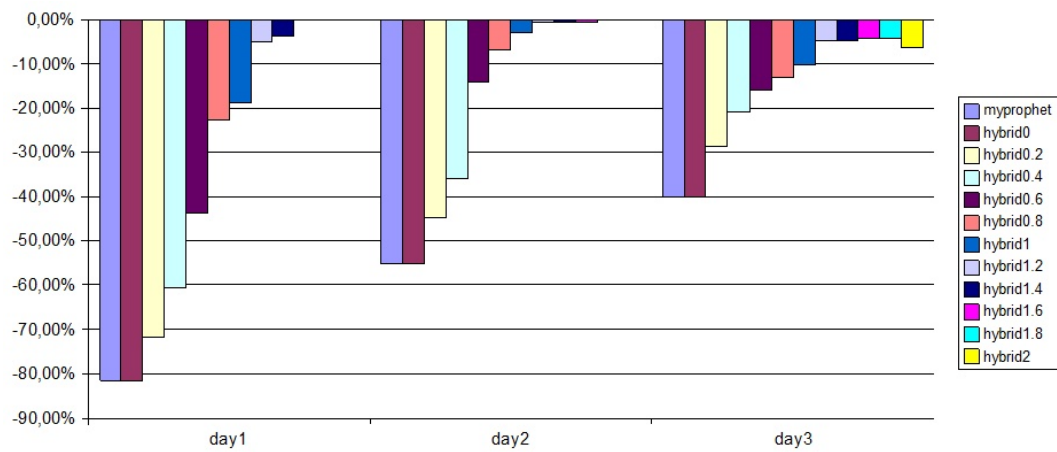


Figure 6.4: Hybrid Router, UMass xmit cost

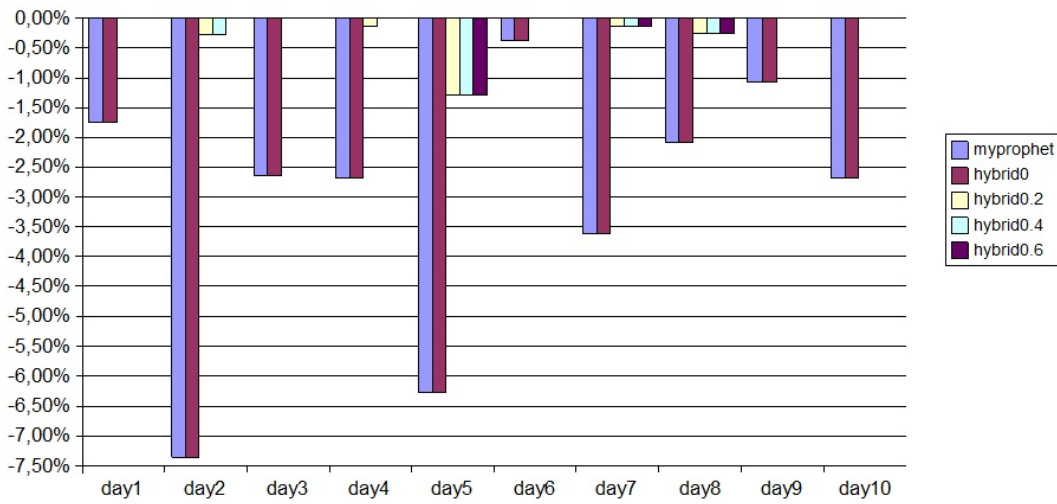


Figure 6.5: Hybrid Router, Reality Mining arrivals

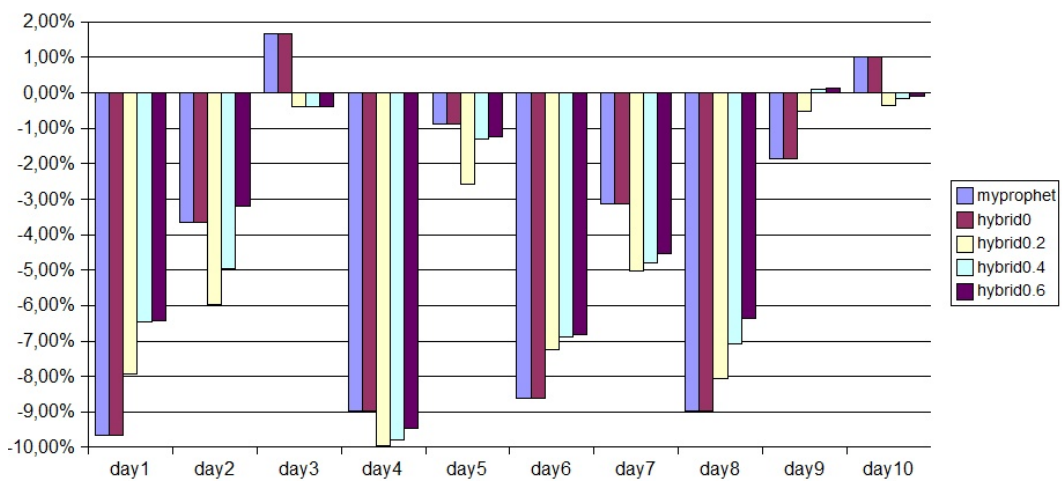


Figure 6.6: Hybrid Router, Reality Mining buffer cost

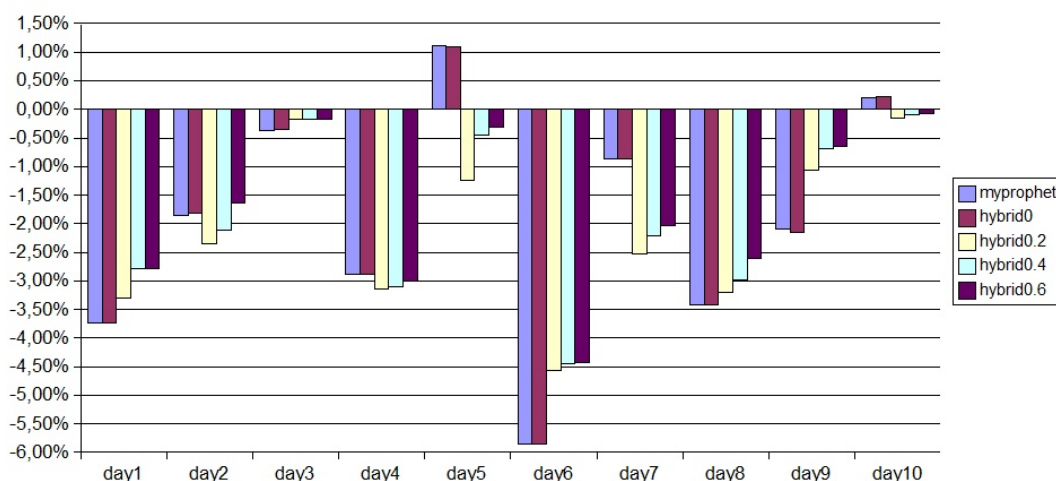


Figure 6.7: Hybrid Router, Reality Mining xmit cost

6.1.2 Discussion

On UMass results we see that hybrid protocol behaves like PROPHEt for $\varepsilon = 0$, as expected, because when ε is zero \mathcal{L} is always bigger or equal to ε and the router will use the PROPHEt protocol. As ε increases the relative arrivals decrease meaning that Hybrid’s arrivals are getting closer to Epidemic’s arrivals.

On day 1, for $\varepsilon = 0.4$ Hybrid’s arrivals are equal to Epidemic’s. However, on the same day and for the same ε both costs are lower than those of Epidemic’s. Similar results are obtained on the other days. This means that Hybrid protocol is capable of delivering the same amount of messages as Epidemic protocol, but does that with smaller resources waste. This is verified on all days of UMass network.

The same quantity of improvement is not observable on the Reality Mining network. Hybrid protocol had the same delivery as Epidemic but only on 70% of the simulated days (figure 6.5, only three bars of hybrid0.6 are different than zero). The cost is lower than Epidemic’s on 90% of the simulated days. There are two anomalies on the Reality Mining’s cost charts: on days 3 and 10 PROPHEt has higher cost than Epidemic. These days were Sundays, table 6.1 shows the days, its contacts and nodes.

The same anomaly does not happen on UMass network results. Coincidentally or not, none of the three simulated days belong to an weekend, the three UMass days are: Monday, Tuesday and Wednesday. We will take this into account in our further analysis of Hybrid’s behaviour and will concentrate our attention on Reality Mining results for which we have more data to support our conclusions.

Before continuing on our comprehension of this protocol’s behaviour, there are two questions to attend to: the ε limit value and granularity.

Table 6.1: Reality Mining network week days, contacts and nodes

Day	Week day	Contacts	Nodes
1	Friday	4.530	349
2	Saturday	2.663	259
3	Sunday	3.516	273
4	Monday	5.280	272
5	Tuesday	5.676	301
6	Wednesday	4.747	309
7	Thursday	5.026	329
8	Friday	4.867	387
9	Saturday	3.213	311
10	Sunday	4.423	249

Notice that while different ε values on UMass results provide a relatively soft transition on metrics values from PROPHET-like to Epidemic-like behaviour this does not happen on Reality Mining results. Figure 6.8 shows simulations done with small ε limits, done on day 10, in order to find a chart of soft transitions like that of UMass.

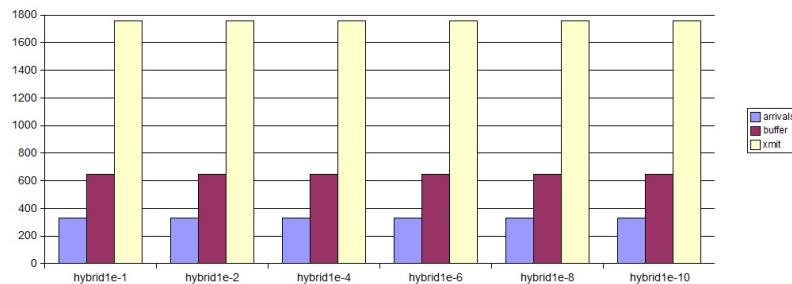


Figure 6.8: Hybrid Router, small limits tests

Despite reductions until $\varepsilon = 1e - 10$ we were unable to find the value to which the behaviour changes from PROPHET to Epidemic in a soft fashion, like that of UMass results.

6.1.3 Limit decision

The decision limit is difficult to choose. The decision would be made choosing the first limit that, for all days, provide the same arrivals as Epidemic protocol. However for some days this limit was not found and simulating more days with greater limits will result into choosing a limit that have cost too similar to Epidemic's. Therefore, we will present here the results for $\varepsilon = 0.5$ for the first ten days on figures 6.9, 6.10 and 6.11. This limit has results with the same delivery success and lower costs than $\varepsilon = 0.6$.

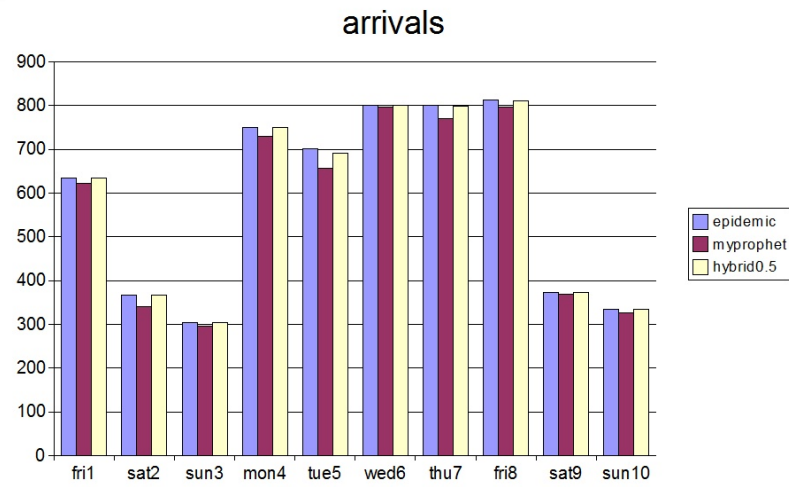


Figure 6.9: Hybrid Protocol, $\varepsilon = 0.5$

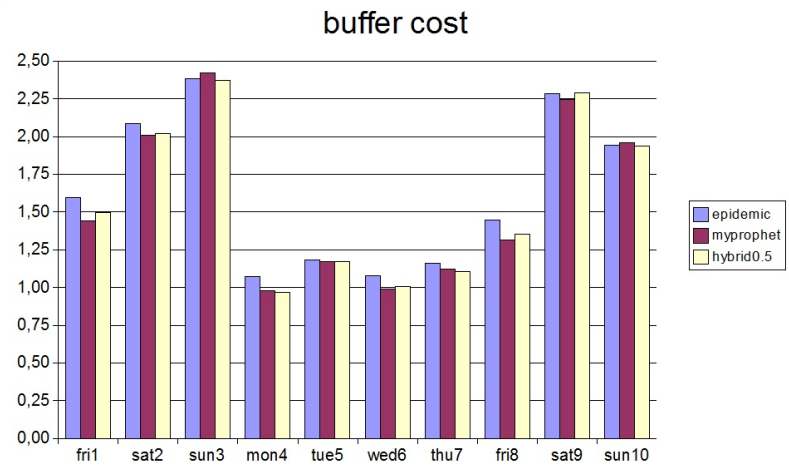


Figure 6.10: Hybrid Protocol, $\varepsilon = 0.5$

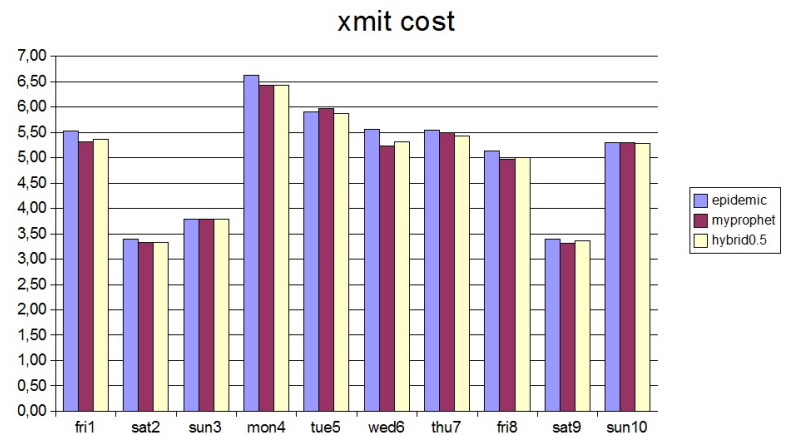


Figure 6.11: Hybrid Protocol, $\varepsilon = 0.5$

The decision was made by observations on the arrivals values that, for this limit, is equal to those of $\varepsilon = 0.6$ while costs are lower. Table 6.2 shows the average values on ten days for the other ε .

Table 6.2: Hybrid protocol, Epidemic’s ratio averages

metric	average (%)				
	hybrid0	hybrid0.2	hybrid0.4	hybrid0.5	hybrid0.6
arrivals	96,95	99,79	99,81	99,83	99,83
buffer-cost	95,70	95,20	95,83	96,12	96,16
xmit-cost	98,02	97,83	98,10	98,21	98,23

Table 6.2 indicates that, in average, we would have better results with $\varepsilon = 0.2$ for which the buffer cost is lower than that of PROPHET’s (which is equal to hybrid0) while the loss in arrivals are not significant. However, because the simulation of a whole month is a time consuming task that would not be possible on the time available, the results to be presented in the next sections will be that of $\varepsilon = 0.5$.

6.2 Behaviour analysis

Results differ from day to day, figure 6.12 shows these differences on each protocol’s results.

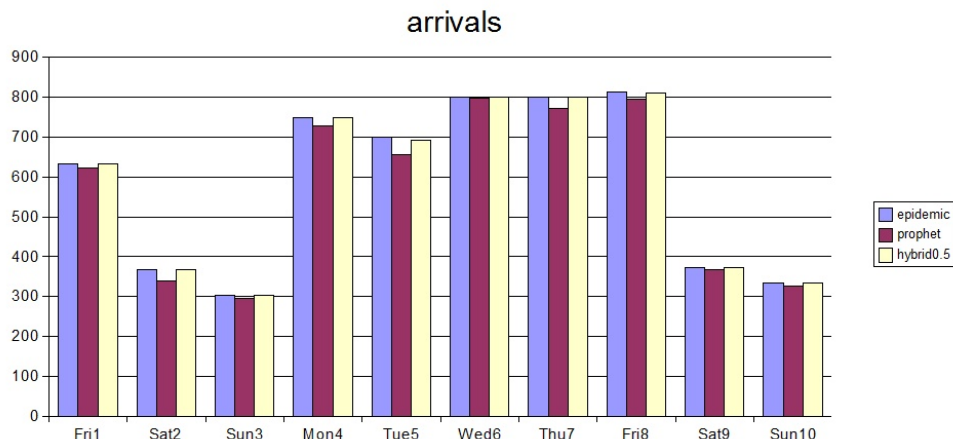


Figure 6.12: Reality Mining, week days and arrivals

Interestingly all protocols have lower number of arrivals on weekends. It makes sense when we consider the activity on week days and weekends in the time slot simulated. Figure 6.13 shows the number of pairs with more than three encounters and those with less than three encounters. The number of encounters was chosen based on the ε limit.

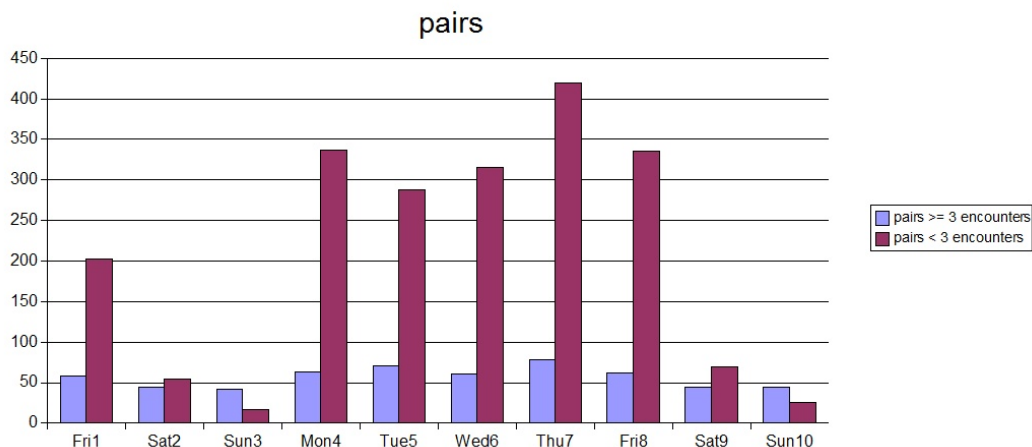


Figure 6.13: Reality Mining, pairs

We simulated six hours and, for being on PROPHET mode, a network node needs 0.5 encounters by hour or three encounters in six hours. The dark bars show the number of pairs with less than three encounters which indicates the amount of nodes in Epidemic mode. This is also similar to what was done in [19] (described in chapter 2) and we could associate the Epidemic mode to routing between ‘strangers’ and PROPEHT mode to routing between ‘friends’. Another interesting feature is that the number of pairs with more than three encounters has low standard deviation (table 6.3).

Table 6.3: Reality Mining pairs, average and deviation

pair	average	standard deviation
encounters ≥ 3 / ‘friends’	56,8	12,6
encounters < 3 / ‘strangers’	206,4	152,1

This indicates that the number of nodes in PROPHET mode have regular values while the number of nodes on Epidemic mode will differ from day to day, being higher on week days and lower on weekend days. Therefore, a different ε for week days and weekend days may improve this protocol’s performance. However on the simulation data used such differentiation was not available and required additional development which due to time restrictions was not possible to do.

6.3 Validation

After choosing the decision limit, the results on a full month were obtained. Each day was simulated as done until now: one day on each simulation, from 06:00 to 12:00. Epidemic and PROPHET protocols were also simulated for comparison.

6.3.1 Results

Figure 6.14, 6.15 and 6.16 shows the simulations on days 1 to 31 of Reality Mining network.

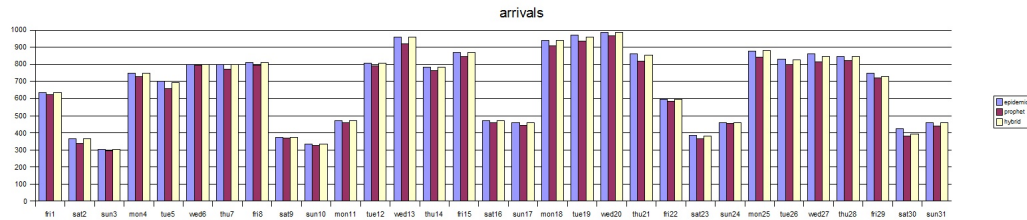


Figure 6.14: Hybrid validation, arrivals

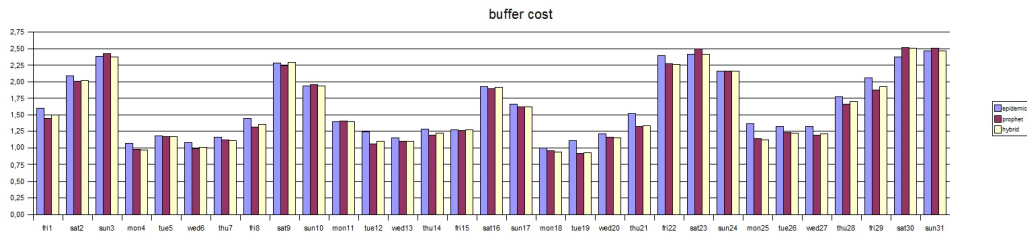


Figure 6.15: Hybrid validation, buffer-cost

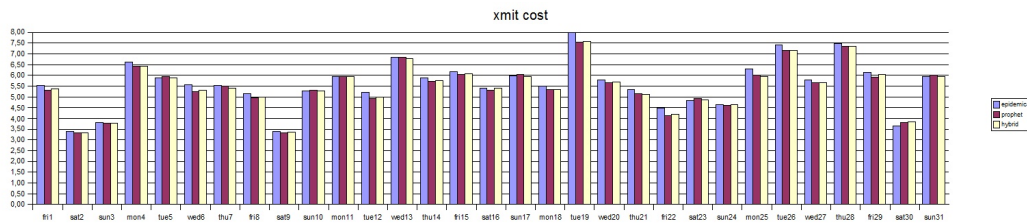


Figure 6.16: Hybrid validation, xmit-cost

Table 6.4: Validation, average and deviation

metric	protocol	average	standard deviation
arrivals	Epidemic	676	219,50
	PROPHEt	653	213,68
	Hybrid	672	219,48
buffer-cost	Epidemic	1,63	0,49
	PROPHEt	1,57	0,53
	Hybrid	1,57	0,53
xmit-cost	Epidemic	5,57	1,11
	PROPHEt	5,46	1,07
	Hybrid	5,47	1,05

Table 6.5: Validation, Protocols Comparison

metric	Hybrid vs Epidemic(%)		Hybrid vs PROPHET(%)	
	avg	stddev	avg	stddev
arrivals	99,48	1,37	103,01	1,48
buffer	94,75	4,88	103,35	1,80
xmit	97,69	1,93	103,17	1,11
buffer-cost	95,25	5,07	100,34	1,79
xmit-cost	98,21	2,26	100,16	1,04

Table 6.6: Validation, Epidemic's ratio averages

metric	protocol	average (%)		
		all days	week days	weekends
arrivals	PROPHET	96,59	96,85	96,05
	Hybrid	99,48	99,61	99,22
buffer-cost	PROPHET	94,99	92,39	100,44
	Hybrid	95,25	93,04	99,90
xmit-cost	PROPHET	98,07	97,01	100,30
	Hybrid	98,21	97,24	100,25

6.3.2 Discussion

The results show that, in average, Hybrid protocol delivers an amount of messages close to Epidemic's with costs close to those of PROPHET (table 6.6). The same table shows that both PROPHET and Hybrid have higher costs than Epidemic on weekends, being PROPHET the protocol with higher costs. This supports the idea that we should have two ε limits: one for week days and another, bigger, for weekends. A bigger ε would make Hybrid to behave more like Epidemic which would provide better results on weekends.

6.4 Conclusions

The results show that merging two protocols give a better outcome than either of those protocols separated. The evaluated router showed effective message delivery with low resources wasting on both networks studied.

Chapter 7

Conclusion

Routing on Delay Tolerant Networks is a complex task which would be easier if the router previously knew when and with whom it would have connections so it could decide the best route. But DTNs connectivity pattern is difficult to predict because it is directly related to human behaviour. Despite their complexity DTNs already have successfully routing approaches. However, each approach is usually for specific types of networks. An example is DTLSR, which for the studied cases did not show good results. That is not completely unexpected, since authors presented this routing protocol as a solution for slow mobility networks, which was not included on the studied cases.

However, generic solutions as Epidemic and PROPHET proved to be flexible enough to provide good connectivity — each one with its virtues and defects.

7.1 Methodology Evaluation

The network traces chosen were not perfect, the UMass network did not have enough days to allow complete comprehension of studied protocols behaviours. But this network was useful to quickly test protocols in order to see if they were working as expected.

The Reality Mining network provided a complete set of people movements and access points connectivity. However, the high number of contacts prohibited simulations including the entire day, more than one day and entire months. Even those simulations not being mandatory to evaluate the solutions proposed it would be better if such simulations were possible.

DTNSim proved to be an useful simulator. However, the internal API still requires some development. The simulator have a complex link system that, when the connection is lost, some times fails to pass the link name and other parameters to the router. This and other problems were solved in the BasicRouter interface so the interference on the developed routers was not significant.

The external router interface — a DTN2 functionality that allows developers to use external processes as routers, therefore develop routers without having to know the details of DTN2 implementation — did not work as expected. On DTNSim, due to the fact that each node behaves like a separated daemon, the data used to control routers is mixed making impossible to receive or send data to or from a specific router. An external router interface compatible with DTNSim would be extremely useful so developers could create their routers on higher level programming languages or script languages and test them on DTNSim.

The addition of a graphical user interface (GUI) could improve DTNSim's functionality. Such GUI could show network nodes movements and their internal properties, e.g. routing tables and buffered bundles.

However, such developments would be out of scope of this work's objectives.

7.2 Contribution Evaluation

7.2.1 Metrics

This work contributed with one metric for measure of DTN's performance. Its functionality is arguable because despite *cost* being an objective measure one may argue that there is no difference from using buffer-cost or xmit-cost instead of buffer and xmit directly.

But imagine that we use buffer and arrivals to measure a protocol's performance. And, when comparing it with another protocol we see that the first have lower buffer and arrivals. This only tell us that the first protocol uses less resources to deliver less messages. However, if we used the buffer-cost instead of buffer and if we observed that buffer-cost was lower we instantly knew that the first protocol was better on its resources usage. The importance of this fact would be still arguable because less messages were delivered.

Because of this, *cost's* usage have to be careful. The conclusions obtained from this metric are clearer on comparison between protocols with equal arrivals like some Hybrid protocol's situations and PROPHET's EDC improvement.

7.2.2 PROPHET changes

A set of three improvements to PROPHET protocol were proposed. The EQ-GRTR showed similar arrivals with lower cost than GRTR. This only proves that EQ-GRTR is better than GRTR, since other forwarding strategies were not tested.

Fast start have arguable results, it was not successful on all days. However it also provide some days with increased delivery over standard PROPHET.

EDC delivers the same messages as standard PROPHET but it is capable of saving buffer and xmit resources on most of tested days. This means that it is possible for the router to successfully estimate the delivery. However the method used for such estimation is arguable. For example, if a network node is meeting low predictability nodes it may be advisable to continue copying messages even when the sum of predictabilities exceeds 1. Or we may use another limit, lower than 1, that might allow the same delivery rate and lower resources than the actual limit presents.

Despite the possibilities yet to explore we are able to say that the current EDC version successfully allows resource saving for PROPHET protocol.

These improvements to PROPHET are only visible on the Reality Mining network.

7.2.3 Hybrid Protocol

The new protocol proposed is an attempt to unify the best of two protocols: a protocol that does the maximum effort to deliver its messages; and a protocol that carefully chooses its forwarders.

Results show that Hybrid protocol is capable of delivering an amount of messages close to those of Epidemic's, being equal on most of days. However, this success is not completely explained. The logic behind this protocol makes sense — network nodes with low number contacts should maximize its delivery — but a relation between network properties and Hybrid's performance was not found.

The ε granularity may also be discussed. On the UMass simulations we see the Hybrid performance moving from PROPHET to Epidemic but on Reality Mining that transition happens too fast. For some reason, on Reality Mining simulations, large number of nodes that are on PROPHET mode switches to Epidemic mode making the router to rapidly behave more like Epidemic.

The Hybrid protocol has good results, with similar delivery and lower costs on both UMass and Reality Mining network.

7.3 Future Work

The developed work answered questions and created new ones on DTN routing optimization. Further optimizations on PROPHET are not out of the table. As already said in this chapter, the optimizations proposed for PROPHET could use further simulations and different approaches. The Hybrid protocol is yet in its infancy, ideas about adaptive ε limit and differentiating ε with basis on the week day were not tested. Another untested possibility was to define one ε for each route table entry which could optimize resource

waste on specific difficult reaching destinations.

Another possible approach to routing improvement is the inclusion of a social model in PROPHET's equations. The social model would make the predictability to change differently with time which could be better. Also, the inclusion of information about user's context — e.g. calendar notes and GPS data — could allow the router to predict better routes than the techniques used allow.

References

- [1] Delay tolerant networking research group. <http://www.dtnrg.org/wiki/Home>, January 2008.
- [2] Khaled Harras Kevin Almeroth and Elizabeth Belding-Royer. Delay tolerant mobile networks (dtmns): Controlled flooding schemes in sparse mobile networks. *IFIP Networking*, May 2005.
- [3] K. Harras and K. Almeroth. Transport layer issues in delay tolerant mobile networks. *IFIP Networking*, May 2006.
- [4] A. Lindgren A. Doria O. Schelen. Probabilistic routing in intermittently connected networks. *Proceedings of the The First International Workshop on Service Assurance with Partial and Intermittent Resources (SAPIR 2004)*, August 2004.
- [5] Jason LeBrun Chen-Nee Chuah Dipak Ghosal. Knowledge based opportunistic forwarding in vehicular wireless ad hoc networks. *IEEE VTC*, Spring 2005.
- [6] Christoph Dwertmann Joerg Ott, Dirk Kutscher. Integrating dtn and manet routing. *Proceedings of the SIGCOMM CHANTS Workshop*, September 2006.
- [7] Mirco Musolesi Stephen Hailes and Cecilia Mascolo. Adaptive routing for intermittently connected mobile ad hoc networks. *Proceedings of IEEE 6th International Symposium on a World of Wireless*, June 2005.
- [8] Greg Welch and Gary Bishop. An introduction to the kalman filter. *University of North Carolina*, July 2006.
- [9] R. Patra S. Jain K. Fall. Routing in a delay tolerant networking. *SIGCOMM*, Aug/Sep 2004.
- [10] Michael Demmer Kevin Fall. Dtlsr: Delay tolerant routing for developing regions. *ACM SIGCOMM Workshop on Networked Systems for Developing Regions (NSDR)*, August 2007.
- [11] Xiaolan Zhang Jim Kurose Brian Neil Levine Don Towsley Honggang Zhang. Study of a bus-based disruption tolerant network: Mobility modeling and impact on routing. *ACM International Conference on Mobile Computing and Networking (Mobicom)*, September 2007.
- [12] Umass trace repository. <http://traces.cs.umass.edu/index.php/Network/Network>, January 2008.

- [13] Katrina M. Hanna Brian Neil Levine and R. Manmatha. Mobile distributed information retrieval for highly partitioned networks. *Proc. IEEE Intl. Conference on Network Protocols (ICNP)*, November 2003.
- [14] Jrg Ott Mikko Pitknen. Dtn-based content storage and retrieval. *The First IEEE WoWMoM Workshop on Atonomic and Opportunistic Communications(AOC)*, June 2007.
- [15] Jrg Ott Dirk Kutscher. Bundling the web: Http over dtn. *Proceedings of WNEPT 2006*, August 2006.
- [16] Jrg Ott Dirk Kutscher. Why seamless? towards exploiting wlan-based intermittent connectivity on the road. *TERENA Networking Conference (TNC) 2004*, June 2004.
- [17] Cathryn Peoples Gerard Parr Bryan Scotney Adrian Moore. A reconfigurable context-aware protocol stack for interplanetary communication. *International Workshop on Space and Satellite Communications (IWSSC '07)*, September 2007.
- [18] C. Partridge. Authentication for fragments. *Fourth Workshop on Hot Topics in Networks (HotNets-IV)*, November 2005.
- [19] Andrew G. Miklas Kiran K. Gollu Kelvin K. W. Chan Stefan Saroiu P. Krishna Gummadi Eyal de Lara. Exploiting social interactions in mobile systems. *UbiComp 2007: Ubiquitous Computing, 9th International Conference*, September 2007.
- [20] MIT Media Lab. Reality mining. <http://reality.media.mit.edu/>, January 2008.
- [21] Jani Kurhinen and Jukka Janatuinen. Geographical routing for delay tolerant encounter networks. *Computers and Communications, 2007. ISCC 2007. 12th IEEE Symposium*, July 2007.
- [22] Brendan Burns Oliver Brock and Brian Neil Levine. Mora routing and capacity building in disruption-tolerant networks. *Elsevier Ad hoc Networks Journal*, To appear 2008.
- [23] K. Fall. Messaging in difficult environments. *Intel Research Berkeley, IRB-TR-04-019*, December 2004.
- [24] Earl Oliver Hossein Falaki. Performance evaluation and analysis of delay tolerant networking. *1st International Workshop on System Evaluation for Mobile Platforms (MobiEval'07)*, June 2007.
- [25] M. Chuah L. Cheng B. Davison. Enhanced disruption and fault tolerant network architecture for bundle delivery. *IEEE Globecom*, November 2005.
- [26] Hyewon Jun Mostafa H. Ammar Mark D. Corner and Ellen Zegura. Hierarchical power management in disruption tolerant networks with traffic-aware optimization. *ACM SIGCOMM Workshop on Challenged Networks (CHANTS)*, September 2006.
- [27] Nilanjan Banerjee Mark D. Corner and Brian Neil Levine. An energy-efficient architecture for dtn throwboxes. *IEEE Infocom*, May 2007.
- [28] Tcl - wikipedia. <http://en.wikipedia.org/wiki/Tcl>, June 2008.

- [29] Carey Williamson. Internet traffic measurement. *IEEE Internet Computing* 5, 70-74,, November/December 2001.
- [30] A. Lindgren and A. Doria. Prophet internet draft. <http://www.ietf.org/internet-drafts/draft-irtf-dtnrg-prophet-00.txt>, February 2008.