# Probability Delegation Forwarding in Delay Tolerant Networks

Xiao Chen Dept. of Comp. Sci. Texas State University San Marcos, TX 78666 xc10@txstate.edu Jian Shen Dept. of Math. Texas State University San Marcos, TX 78666 js48@txstate.edu Taylor Groves
Dept. of Comp. Sci.
Texas State University
San Marcos, TX 78666
tg1106@txstate.edu

Jie Wu
Dept. of Comp. Sci. and Eng.
Florida Atlantic University
Boca Raton, FL 33431
jie@cse.fau.edu

Abstract—Delay tolerant networks are a type of wireless mobile networks that do not guarantee the existence of a path between a source and a destination at any time. In such a network, one of the most important issues is to reliably deliver data with a low latency. Naive forwarding approaches, such as flooding and its derivatives, make the routing cost very high. Many efforts have been made to reduce the cost while maintaining performance. Recently, an approach called delegation forwarding (DF) caught significant attention in the community because of its simplicity and good performance. In a network with N nodes, it reduces the cost to  $O(\sqrt{N})$  which is better than O(N) in other methods. In this paper, we put forward a scheme called probability delegation forwarding (PDF) that can further reduce the cost to  $O(N^{\log_{2+2p}(1+p)}), p \in (0,1)$ . In addition, we propose the threshold probability delegation forwarding (TPDF) scheme to close the latency gap between the DF and PDF schemes. Simulation results show that our schemes can reduce the cost while maintaining the routing performance.

Index Terms—delay tolerant networks, forwarding algorithms, routing, traces

#### I. INTRODUCTION

Delay tolerant network (DTN) is a type of wireless mobile network that does not guarantee the existence of a path between a source and a destination at any time. When two nodes move within each other's transmission range during a period of time, they *contact* or *meet* each other. When they are out of each other's transmission range, the connection is lost. The message to be delivered needs to be stored in the local buffer. Examples include people carrying mobile devices moving in conferences, university campuses and in social settings. The message delivery in this kind of network is multihop and the connection between nodes is *non-predictable*. Furthermore, there is limited knowledge of each node in the network.

In such a DTN, the most important metric is the *delivery ratio*, because the network must be able to reliably deliver data. The second metric is the *delivery latency* [10]. The third one that attempts to minimize resource consumption such as buffer space or power is the number of *copies* duplicated.

The rudimental routing approach in a non-predictable DTN is flooding [21], which incurs a high cost. Many algorithms have been put forward to reduce the cost of flooding [2], [5], [6], [11], [15], [18] by forwarding messages to a higher quality

node that has a better chance to deliver the message to the destination. One approach called *delegation forwarding* (DF) [7] caught significant attention in the community because of its simple approach and good performance. Its main idea is to assign each node a quality. In each hop of the routing, a message holder not only will forward the message to a node with a higher quality than itself but also raise the quality of itself to the same level. In this way, the cost of routing is reduced while achieving similar performance. Analysis shows that in an N-node network, delegation forwarding has an expected cost of  $O(\sqrt{N})$  while a naive scheme of forwarding to any higher quality node has an expected cost of O(N).

In this paper, we show that there is still room to improve DF. We put forward a new scheme called *probability delegation forwarding* (PDF) which can further reduce the cost. Based on DF, our main idea is to insert a probability p into the algorithm. That is, when node  $u_i$  meets node  $u_j$  with a higher quality than itself, there is a p ( $p \in (0,1)$ ) chance that  $u_i$  will forward the message to  $u_j$ . Analysis shows that using our scheme, the cost will be brought down to  $O(N^{\log_2+2p}(1+p))$ . Simulation results show that PDF can achieve a similar delivery ratio as the DF scheme. In addition, we propose another scheme called *threshold-base probability delegation forwarding* (TPDF) to close the latency gap between the DF and PDF schemes. Simulation results show that our schemes can reduce the cost and maintain the performance.

The rest of the paper is organized as follows: Section II mentions the related work; Section III puts forward the probability delegation forwarding algorithm; Section IV presents analysis of PDF; Section V shows the simulation results of PDF; Section VI proposes the threshold-based probability delegation forwarding scheme; Section VII presents the simulation results of TPDF and the conclusion is drawn in Section VIII.

# II. RELATED WORK

Due to the uncertainty and time-varying nature of DTNs, routing poses unique challenges. In the literature, some routing approaches are based on deterministic mobility [8], [9], [12]–[14], [16], [19], [20] while some others are based on non-predictable mobility [2], [5], [6], [11], [15], [18], [21]. Here,

#### Algorithm DF: Delegation Forwarding

```
1: Let u_1, \dots, u_N be nodes
 2: Let m_1, \dots, m_M be messages
 3: Node u_i has quality x_{ik} and level \tau_{ik} for m_k.
 4: INITIALIZE \forall i, k : \tau_{ik} \leftarrow x_{ik}
 5: On contact between u_i and node u_i:
 6: for k in 1, \dots, M do
 7:
      if m_k is currently held by u_i and \tau_{ik} < x_{jk} then
 8:
 9:
          if u_i does not have m_k then
            forward m_k from u_i to u_j
10:
11:
12:
      end if
13: end for
```

we discuss the situation of non-predictable mobility: nodes move dynamically in different directions with different speeds.

If the non-predictable mobility model is used, one rudimental approach for routing is to perform a flooding-based route discovery as in [21] where whenever a host receives a message, it will pass it to all those nodes it can reach directly at that time so that the spread of the message is like the epidemic of a disease. Epidemic routing has the highest performance. However, its cost is too high. Many algorithms have been put forward to reduce the cost [2], [5], [6], [11], [15], [18] by forwarding message only to a higher quality node that is more likely to meet the destination.

Recently, a strategy called delegation forwarding [7] has been proposed. Its main idea is that each node has an associated quality metric. A node will forward a message only if it encounters another node whose quality metric is greater than any seen by the message so far. The authors show that despite the simplicity of the strategy, it works surprisingly well. Analysis shows that in an N-node network, delegation forwarding has an expected cost  $O(\sqrt{N})$  while the naive scheme of forwarding to any higher quality node has an expected cost O(N). Simulations on real traces show performance as good as other schemes at a much lower cost. Delegation forwarding is presented in Algorithm DF here. All the algorithms presented in this paper consider multiple messages. In our comparisons, for convenience's sake, only a single message is considered.

In this paper, we strive to extend the DF algorithm to further bring down the cost while maintaining similar performance.

#### III. PROBABILITY DELEGATION FORWARDING (PDF)

We believe that we can reduce cost even more by simply involving probability p in the DF algorithm. Our approach seeks to forward the message to the highest quality nodes in the system with a probability. That is, if the probability is set as p ( $p \in (0,1)$ ), and if a node  $u_i$  meets a node  $u_j$  with a higher quality than itself,  $u_i$  will forward the message to  $u_j$  with a probability of p (see Algorithm PDF). In other words, it is not 100% as in the DF algorithm.

This approach does not need global knowledge. Each node decides whether to forward the message or not by itself.

### Algorithm PDF: Probability Delegation Forwarding

```
1: Let u_1, \dots, u_N be nodes
 2: Let m_1, \dots, m_M be messages
 3: Node u_i has quality x_{ik} and level \tau_{ik} for m_k.
   INITIALIZE \forall i, k : \tau_{ik} \leftarrow x_{ik}
    On contact between u_i and node u_i:
    for k in 1, \dots, M do
7:
      if m_k is currently held by u_i and \tau_{ik} < x_{jk} then
 8:
          if u_i does not have m_k and u_i is chosen by p then
 9:
              forward m_k from u_i to u_j
10:
11:
          end if
12:
      end if
13: end for
```

## IV. ANALYSIS

In this section, we compare the costs of the DF and PDF algorithms mathematically. We consider a single message and calculate the number of copies created for each message.

# A. Cost of DF

The cost of DF is given in [7]. To make the paper inclusive, we include the idea here. For any node  $u_i$  maintaining a quality metric  $x_i$  and a level value  $\tau_i$ , we focus on the gap  $g_i = 1 - \tau_i$  between the current level and 1. The node that generates the message has an initial level  $\tau_i = x_i$ . The initial gap  $g = 1 - x_i$ .

Consider a node that updated its gap value n times. The node's current gap is denoted as the random variable  $G_n$ . Since nodes meet according to rates that are independent of node quality, the node is equally likely to meet a node with any particular quality value. The next update of the gap occurs when it meets a node with a quality greater than  $G_n$ , and all values above this level are equally likely.

Hence, we can write

$$G_{n+1} = G_n \times U,\tag{1}$$

where U is independent of  $G_n$  and follows a uniform distribution on (0,1]. By induction we then find:

$$E[G_{n+1}|G_n] = \frac{G_n}{2}$$
, hence,  $E[G_n] = \frac{g}{2^n}$ .

Moreover, from Eq.(1), we see that  $G_n$  approximately follows a lognormal distribution (see [3]), with median  $\frac{g}{e^n}$ . Hence the distribution is highly skewed with most of the probability mass below the mean, and so with large probability we have  $G_n \leq \frac{g}{2^n}$ .

The replication process can be described by a dynamic binary tree T, which contains all the nodes that have a copy of the message. Initially T contains a single node with associated gap g. Each time a node with a copy of the message meets another node having higher quality than any node seen so far, two child nodes are created for the node. Both have an updated gap value. Some branch of the tree will grow faster than others. The total size of the tree represents the upperbound on the number of copies created. We wish to bound the total size of the tree.

We define the set  $B=\{i|x_i\geq 1-\frac{g}{\sqrt{N}}\}$ , which we call the *target set*. We will also identify a subtree of the tree T in which children are excluded for nodes having a level above  $1-\frac{g}{\sqrt{N}}$ . In other words, all the nodes in the subtree have a gap  $<\frac{g}{\sqrt{N}}$ . This subtree is called the *target-stopped tree*.

The essential observation is the following: if n is close to  $\log_2(\sqrt{N})$ , then except with a small probability, a node at generation n in the tree has a gap at most  $\frac{g}{2^n} \leq \frac{g}{\sqrt{N}}$ . This is because of the highly skewed nature of the distribution of  $G_n$ , as described above. Hence, we can safely assume that the target-stopped tree has a depth of at most n. Note that the total number of nodes appearing at generations  $0, 1, \dots, n-1$  is at most  $2^n = \sqrt{N}$ .

Now we can calculate the total number of copies generated in this process:

$$C_{DF}(n) = 2^n + \frac{Ng}{2^n}.$$

In the worst case, g is 1. So,

$$C_{DF}(n) \le C_{WDF}(n) = 2^n + \frac{N}{2^n}.$$

The minimum value  $\min\_C_{WDF}$  of  $2^n+\frac{N}{2^n}$  is obtained by making the two items  $2^n$  and  $\frac{N}{2^n}$  equal. That is,  $2^n=\frac{N}{2^n}$ . Thus,  $n=\frac{1}{2}\log_2 N$ . So,

$$\min C_{WDF} = 2\sqrt{N} = O(\sqrt{N}).$$

## B. Cost of PDF

In the PDF algorithm, node i has a p  $(p \in (0,1))$  probability to forward the message. For example, if  $p=\frac{3}{4}$ , then the node has 75% of the chance to forward the message. If the node is not chosen by p, it is equivalent to truncating the subtree from this node in the binary tree. Since the nodes are randomly chosen by the probability p,  $E[G_n]=\frac{g}{2^n}$  still holds.

We define the set  $B=\{i|x_i\geq 1-\frac{2^n}{2^n}\}$  as the target set, and the subtree with all the nodes whose gap  $<\frac{g}{2^n}$  as the target-stopped tree.

Now we calculate the total number of copies generated as:

$$C_{PDF}(n) = (1+p)^n + \frac{Ng}{2^n}.$$

In the worst case, g is 1. Therefore,

$$C_{PDF}(n) \le C_{WPDF}(n) = (1+p)^n + \frac{N}{2^n}.$$
 (2)

Now the minimum value  $\min_{CWPDF}$  of  $C_{WPDF}(n)$  can be obtained by making its derivative equal to 0.

$$C'_{WPDF}(n) = (1+p)^n \ln(1+p) - N \cdot 2^{-n} \ln 2 = 0$$

So, 
$$(2+2p)^n = \frac{N \ln 2}{\ln(1+p)}$$

Then,

$$n = \log_{2+2p} \frac{N \ln 2}{\ln(1+p)}$$
  
=  $\log_{2+2p} N + \log_{2+2p} \ln 2 - \log_{2+2p} \ln(1+p)$ 

So,

$$\begin{aligned} \min \_C_{WPDF} &= \\ C_{WPDF} (\log_{2+2p} N + \log_{2+2p} \ln 2 - \log_{2+2p} \ln (1+p)) \\ &< C_{WPDF} (\log_{2+2p} N) \end{aligned}$$

If  $n = \log_{2+2n} N$ , according to Eq. (2),

$$C_{WPDF}(n) = (1+p)^n + \frac{N}{2^n} = 2 \cdot (1+p)^n$$
  
=  $2 \cdot (1+p)^{\log_{2+2p} N} = 2 \cdot N^{\log_{2+2p}(1+p)}$ 

So.

$$C_{WPDF}(n) = 2 \cdot N^{\log_{2+2p}(1+p)} = O(N^{\log_{2+2p}(1+p)}).$$

Since  $p \in (0,1)$ ,  $1+p < \sqrt{2+2p}$ . So  $2 \cdot N^{\log_{2+2p}(1+p)} < 2\sqrt{N} = \min\_C_{WDF}$ . Therefore,  $\min\_C_{WPDF} < \min\_C_{WDF}$ . Hence we see that if  $p \in (0,1)$ , probability delegation forwarding can further reduce the number of copies.

# V. SIMULATIONS OF PDF

We conduct simulations to compare DF and PDF. Actually DF can be treated as a special case of PDF with a probability of 100%. So in the simulations, the results for probability 100% are actually for algorithm DF and the results for probabilities less than 100% are for PDF algorithm with different probabilities.

In our simulations, we use real traces posted on [1]. The data sets consist of contact traces between short-range Bluetooth enabled devices (iMotes [4]) carried by individuals in conference environments, namely Content 2006 and Infocom 2006. In short, we call them Content trace and Info trace.

In the simulations, we use three metrics as follows.

- Delivery Ratio: it is the most important network performance metric in DTNs. It is defined as the fraction of generated messages that are correctly delivered to the final destination within a given time period.
- Latency: it is the time between when a message is generated and when it is received. This metric is important because minimizing latency lowers the time messages spend in the network, reducing contention for resources.
   So lowering latency indirectly improves delivery ratio.
- Copies: it is the number of copies of a message that
  a protocol generates in routing. It is an approximate
  measure of the computational resources required, as there
  is some processing required for each message. Also it is
  also an approximate measure of power consumption, and
  bandwidth and buffer usages as more copies will use more
  of these resources. This is what we call cost in the paper.

The quality of each node in DF and PDF can be decided using different criteria in the forwarding algorithms as follows:

- Frequency (**Freq**) [6]: Node  $u_i$  forwards  $m_k$  to node  $u_j$  if  $u_j$  has more total contacts with all other nodes than does  $u_i$ . This algorithm is destination independent.
- Last Contact (**LastContact**) [7]: Node  $u_i$  forwards  $m_k$  to node  $u_j$  if  $u_j$  has contacted any node more recently than has  $u_i$ . This algorithm is destination independent.

- Destination Frequency (**DestFreq**) [7]: Node  $u_i$  forwards  $m_k$  to node  $u_j$  if  $u_j$  has contacted  $m_k$ 's destination more often than has  $u_i$ .
- Destination Last Contact (**DestLastContact**) [5]: Node  $u_i$  forwards  $m_k$  to node  $u_j$  if  $u_j$  has contacted  $m_k$ 's destination more recently than has  $u_i$ .

We randomly generate a source and a destination. We try different probabilities starting from 80% to 100% with an increase step of 5%. For each source and destination pair, under a certain probability, we use all the forwarding algorithms above on both traces. We record delivery ratio, latency and the number of copies used for each set of data. The process is repeated for 10,000 pairs of randomly generated source and destination pairs. The results are averaged and shown in Figs. 1(a), 1(b), 1(c), 1(d), 1(e), and 1(f).

From the results in both traces, we can see that if we use a probability above 80%, the curves in the delivery ratio are almost flat. There is a slight increase in the delivery latency. That means, the latency will increase with the decrease of probability. For the number of copies, we know that DF (probability 100%) uses the most number of copies. Suppose the number of copies used by DF is  $C_{DF}$  and the number of copies used by PDF with probability p is  $C_{PDF}$ , we calculate ratio  $\frac{C_{PDF}}{C_{DF}}$ . Since DF is the baseline, its ratio is 100%. As the results in both traces show, more and more copies can be saved with the decrease of probability.

# VI. THRESHOLD-BASED PROBABILITY DELEGATION FORWARDING (TPDF)

As we can see from the above simulations, with the decrease of probability, the delivery latency increases. If we use Freq algorithm as an example and look at Fig. 2(c), there is a latency gap between DF and PDF. Our next task is to close the gap between the two. Our main idea is: if node  $u_i$  meets node  $u_j$  with a much higher quality, that is, if  $\frac{x_{jk}-\tau_{ik}}{\tau_{ik}}$  is higher than a certain *threshold* (TH), then without hesitation, node  $u_i$  will forward the message to node  $u_j$  if  $u_j$  does not have the message (see Algorithm TPDF). Otherwise, forward or not will be decided by the probability as in the PDF algorithm. TH is a value which can be set as 0.05 (5%), 0.10 (10%), 0.25 (25%), or 0.50 (50%).

The intuition of this algorithm is that if a node meets a node with a much higher quality, then forwarding message to this node without the decision by the probability will help the message to get higher chance to reach the destination sooner.

# VII. SIMULATIONS OF TPDF

In this section, we conduct simulations to compare TPDF, PDF and DF. DF is PDF with a probability of 100% and PDF is TPDF without the threshold. In our simulations, we set TH to be 0.05, 0.1, 0.25, and 0.5, and the probability to be 80% for the Content trace and 85% for the Info trace. We still look at the three metrics: delivery ratio, latency and number of copies.

For the delivery ratio, we try Freq, LastContact, DestFreq and DestLastContact algorithms using both traces. The results are shown in Figs. 2(a) and 2(b). From the figures, the delivery

# Algorithm TPDF: Threshold-based Probability Delegation Forwarding

```
1: Let u_1, \dots, u_N be nodes
 2: Let m_1, \dots, m_M be messages
 3: Node u_i has quality x_{ik} and threshold \tau_{ik} for m_k.
 4: INITIALIZE \forall i, k : \tau_{ik} \leftarrow x_{ik}
 5: On contact between u_i and node u_i:
 6: for m in 1, \dots, M do
       if m_k is currently held by u_i then
 7:
         if \frac{x_{jk} - \tau_{ik}}{\tau_{ik}} > TH then \tau_{ik} \leftarrow x_{jk}
 8:
 9:
10:
             if u_j does not have m_k then
11:
               forward m_k from u_i to u_j
12:
13:
         else
          if \tau_{ik} < x_{jk} then
14:
15:
            \tau_{ik} \leftarrow x_{jk}
16:
           if u_i does not have m_k and u_i is chosen by p then
               forward m_k from u_i to u_j
17:
            end if
18:
          end if
19:
         end if else
20:
       end if
22: end for
```

ratios of DF, TPDF with TH= 0.05, 0.1, 0.25, and 0.5, and PDF are almost the same, but we can still see that the delivery ratio can get closer to DF's if a threshold is set.

For the delivery latency, to take a closer look, we just use Freq and DestLastContact algorithms as examples. In the Freq algorithm, we set the probability to be 80% and use the Content trace while in the DestLastContact algorithm, we set the probability to be 85% and use the Info trace. The results are shown in Figs. 2(c) and 2(d). From the figures, we can see that setting some threshold can bring down latency.

For the number of copies, again we use Freq and Dest-LastContact algorithms with the same setting. We use DF's copy number  $C_{DF}$  as the baseline and calculate ratio  $\frac{C_{other}\ algorithm}{C_{DF}}$ . The results are in Figs. 2(e) and 2(f). DF has the highest number of copies and PDF has the least. TPDF with some threshold has a copy number between the two.

From the results we know that DF, PDF and TPDF have similar delivery ratio which is the most important metric in DTNs. The selection of a good threshold TH is important to saving more copies at a cost of slight increase in latency. For example, in the Freq algorithm, setting TH=0.10 can decrease the number of copies by 5.9% from PDF at an expense of increasing latency by 1.7% from DF. And in the DestLastContact algorithm, setting TH=0.05 can bring down the number of copies by 3.45% from PDF at a cost of increasing latency by only 0.28% from DF.

#### VIII. CONCLUSION

In this paper, we put forwarding a PDF scheme to further reduce the cost in the DF scheme. PDF can achieve a similar delivery ratio, which is the most important metric in DTNs,

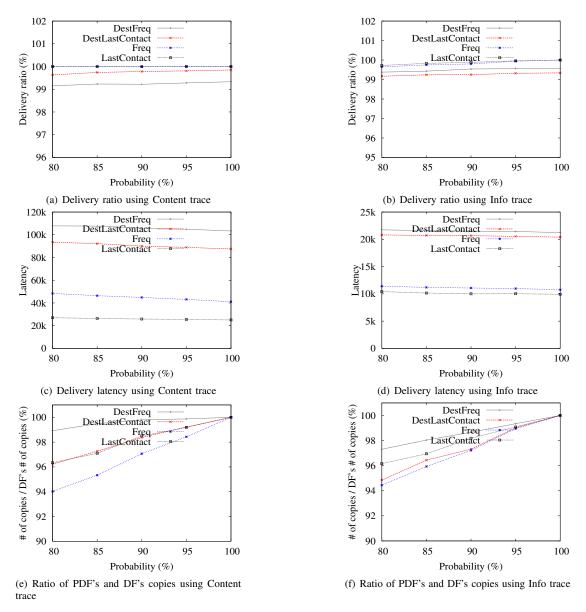


Fig. 1. Comparison of DF and PDF using Content and Info traces

as in the DF scheme. The delivery latency in PDF increases a little compared with DF. That can be mutualized by the TPDF scheme. If a threshold is set properly, TPDF can achieve the similar latency as the DF scheme at a low cost.

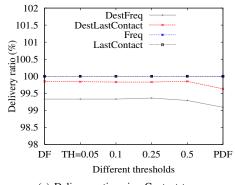
In this paper, we set the probability p to be a value of at least 80%. What happens if p is less than 80%? What is a good choice for the TPDF threshold? What if multiple messages are considered? These questions will be addressed in our future work.

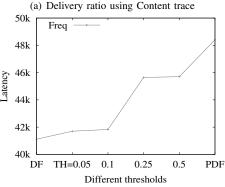
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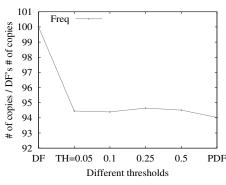
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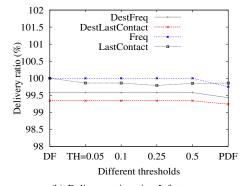


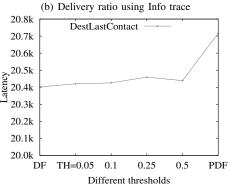


(c) Latency of Freq algorithm using Content trace

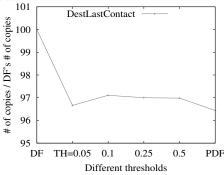


(e) Ratio of PDF's, TPDF's and DF's copies using Content trace





(d) Latency of DestLastContact algorithm using Info trace



(f) Ratio of PDF's, TPDF's and DF's copies using Info trace

Fig. 2. Comparison of TPDF, PDF and DF using Content and Info traces

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