Solving Consensus in Opportunistic Networks
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ABSTRACT

Opportunistic networks are partially connected wireless ad hoc networks, in which pairwise unpredicted transient contacts between mobile devices are the only opportunities for these devices to exchange information or services. Ensuring the coordination of multiple parts of a distributed application in such conditions is a challenge. This paper presents a system that can solve consensus problems in an opportunistic network. This system combines an implementation of the One-Third Rule (OTR) algorithm with a communication layer that supports network-wide, content-driven message dissemination based on controlled epidemic routing. Experimental results obtained with a small flotilla of smartphones are also presented, that validate the system and demonstrate that consensus can be solved effectively in an opportunistic network.

Categories and Subject Descriptors


General Terms

Design, Experimentation, Measurement, Performance

Keywords

Consensus, opportunistic networking, opportunistic computing

1. INTRODUCTION

Opportunistic networks constitute a category of mobile ad hoc networks (MANETs) in which the sparse or irregular distribution of mobile devices (or nodes) yield frequent link disruptions and network partitions [6]. While a MANET is often represented as a time-varying graph (TVG [9]) that remains connected at any time, an opportunistic network must rather be depicted as a non-connected TVG, as shown in Figure 1, mobile nodes only come into contact with each other every now and then, depending on the mobility patterns of their respective carriers (which can be human beings, but also vehicles, animals, etc.). When a transient contact occurs between two nodes, this contact has not been planned in advance so it must be exploited opportunistically by these two nodes to exchange information or services [32].

Relying exclusively on unpredicted pairwise contacts between mobile nodes does not imply that long-distance communication is not feasible in an opportunistic network. Indeed, each mobile node can serve as a “data mule” for messages that propagate in the network, storing messages in a local cache, and carrying them for a while before they can be forwarded to other nodes. This “store, carry and forward” principle is the foundation principle of Delay/Disruption-Tolerant Networking (DTN [17]): messages can indeed propagate in a partially or intermittently connected environment, but their delivery is not guaranteed and can be delayed by minutes, or even hours or days, as it depends on the wanderings of benevolent mobile carriers.

Opportunistic computing has recently been proposed as a new computing paradigm to build upon opportunistic networking. The general idea is to develop a framework that enables collaborative computing in environments where long disconnections and network partitions are the rule [14]. Several middleware systems have already been proposed to ease the development of distributed applications that are meant to run in opportunistic networks: a middleware system that supports asynchronous messaging based on opportunistic communication is proposed in [3], one that supports the tuple-space model in opportunistic networks is described in [4], and systems that support service provisioning and delivery in opportunistic networks are presented in [15] and [28].

Such middleware systems constitute a first step toward application development targeting opportunistic networks. Yet, in complex distributed applications, multiple processes running on different mobile nodes must be able to agree on a common course of actions. Consensus problems have been studied extensively in the literature, with various system models. Each system model makes specific assumptions
about the underlying communication model (synchronous or asynchronous, based on either reliable or faulty links) and about the processes themselves (reliable, susceptible to crash or to exhibit Byzantine failures). A large majority of papers addressing consensus adopt system models that suit traditional wired networks, such as the Internet. To date only a few papers have considered the consensus problem in wireless ad hoc networks and, to the best of our knowledge, no paper has considered the consensus problem in opportunistic networks yet.

As observed in [7] the models for wired networks are often strongly biased towards node failures to the detriment of link failures. Yet, mobile ad hoc networks—including opportunistic networks—require a model that admits both transient process and link faults, and that considers such faults as benign failures. The Heard-Of (HO) model [13] meets these requirements. This model does not distinguish faulty processes from faulty links, as it focuses on transmission faults (effects) without accounting for the faulty components (causes) [13].

Indeed, in an opportunistic network such as that shown in Figure 1, a wireless link between two neighbor nodes is inherently transient, so when this link disappears (for example because both nodes have moved away from each other) this should not be considered as a “fault”: this is the expected consequence of both mobility and limited transmission range in an opportunistic network. Similarly, mobile nodes in an opportunistic network often run on batteries, so a common strategy to preserve their power budget is to turn them off—or put them in suspend mode, or disable their wireless transceiver—frequently. When a node suddenly “disappears” from the network, this should not always be considered as a “fault” (not even a benign one), because this is a perfectly legitimate behavior for this kind of node in this kind of environment.

As mentioned above, delivering messages to remote nodes in an opportunistic network is not guaranteed, for it depends on the wanderings of carriers whose mobility patterns are neither planned nor controlled. Messages can therefore get lost before reaching their destination(s). Two of the consensus algorithms defined in [13] for the HO model can easily tolerate message loss: the Paxos/Last Voting (P/LV) algorithm, and the One-Third Rule (OTR) algorithm. An implementation of the P/LV algorithm for mobile ad hoc networks has been proposed in [7]. This implementation could not run in an opportunistic network, though, as it does not allow mobile nodes to store and carry messages while moving in the network.

In the remainder of this paper we present a middleware system that implements the OTR algorithm, and that can run effectively in an opportunistic network. This system has been fully implemented in Java, and it has been tested in real conditions using a small flotilla of smartphones as mobile nodes. The OTR algorithm is implemented on top of a communication layer that supports network-wide, content-driven message dissemination based on controlled epidemic routing. This combination of the OTR algorithm and epidemic routing is consistent, because epidemic routing is an effective way to disseminate messages in an opportunistic network, where mobility patterns are neither planned nor controlled, and where mobile nodes (with the messages they carry) can disappear for a while from the network. Besides, the OTR algorithm requires n-n communication: in every step each process (or mobile node) must send a message to all other processes (or nodes). With epidemic routing, sending a message to all nodes is not significantly different from sending a message to a single node, so combining the OTR algorithm with epidemic routing makes sense when targeting opportunistic networks.

The remainder of this paper is organized as follows. Related work is presented and discussed in Section 2. The system model we consider in this work is detailed in Section 3. The Heard-Of (HO) model and the One-Third Rule (OTR) algorithm are presented in Section 4. Section 5 presents our implementation of the OTR algorithm, and Section 6 presents experimental results obtained with this implementation. Section 7 concludes the paper.

2. RELATED WORK

Consensus problems have been studied extensively during the last decades, most often with system models that fit the characteristics of traditional wired networks. As a general rule, the papers assume that the network is static and connected, that is, any node can send a message at any time to any other node. Moreover, they also assume that the system can eventually become synchronous, or that it can be augmented with failure detectors, so as to go round the so-called “FLP impossibility result” [18].

Most system models focus on node failures and tend to neglect link failures, though. According to Borran et al. [7] this bias may have its root in the FLP paper [18] (which assumes process crashes and reliable links), but solutions designed for environments where this bias is acceptable should not be used in environments where it is not acceptable.

Mobile ad hoc networks are such environments where reliable links should never be assumed, and for which system models must admit transient link failures. Yet several consensus protocols and algorithms have been proposed for such networks, based on overly optimistic assumptions.

Crash-tolerant broadcast protocols that can be used to solve consensus problems in mobile ad hoc networks are presented in [38]. These protocols assume the existence of oracles that can predict contacts and transmission times between nodes at any point of time. Such an assumption can only be satisfied in very specific mobile networks such as those where the mobility patterns of nodes are planned or controlled explicitly.

A hierarchical consensus protocol involving failure detectors is proposed in [59]. Mobile hosts are distributed into clusters, each cluster being controlled by a clusterhead. The protocol can tolerate faulty nodes, but links are assumed to be reliable.

An algorithm to solve consensus without knowing which nodes—and how many of them—are participating in the consensus is presented in [19]. But the system model for this algorithm assumes reliable links and nodes that cannot crash. Variants of this algorithm have later been proposed in [11] and [19]. Both variants can admit faulty nodes, but links are still assumed to be reliable.

An implementation of the Paxos/Last Voting (P/LV) algorithm is proposed in [7]. This round-based algorithm requires the election of a coordinator, which once elected collects the contributions of all other nodes until consensus is reached. Interestingly, unlike the abovementioned solutions the P/LV algorithm can admit both link and node failures, as it relies on the Heard-Of (HO) model that makes no dis-
tinction between both kinds of failures. It assumes end-to-end connectivity in the network, though, which is a reason-
able assumption for traditional mobile ad hoc networks, but an unfit one for opportunistic networks.

Byzantine agreement protocols for highly dynamic syn-
chronous networks have been proposed in [2] and [20]. In [2]
two randomized round-based protocols are presented, that
achieve high probability, even under a large number of Byzantine
nodes and continuous adversarial churn. The network is rep-
resented as a sparse bounded degree expander graph that
is assumed to remain connected at any time, although its
topology can change arbitrarily from round to round. The
protocol presented in [20] creates and maintains an expander
overlay of clusters. Each cluster is used to inhibit the behav-
ior of Byzantine nodes, and the overlay ensures communica-
tion among clusters. Although these protocols can support
Byzantine failures, they can only run in connected networks
and could hardly be used in opportunistic networks.

As explained in Section 1, an opportunistic network is a
mobile ad hoc network that is at best only partially con-
ected. Such a network can be continuously partitioned,
so end-to-end transmission between remote nodes cannot
rely exclusively on multiple rounds of short-range wireless
transmissions. In fact, a temporaneous multi-hop path be-
tween two nodes may never exist during the network’s whole
lifetime. In such conditions the “store, carry and forward”
principle, which is the key principle of Delay/Disruption-
Tolerant Networking (DTN), must be used to allow mes-
sages to cover long distances by being carried physically by
mobile nodes [30] [36]. With this approach mobility is an
asset, as it helps bridge the gap between nodes that would
otherwise be unable to communicate.

Message routing in opportunistic networks has already
justified a fair amount of research. Most of the algorithms
proposed rely on more or less constrained variants of the
epidemic dissemination scheme, as defined in [37] and [16]:
whenever a message is sent, several copies of the message
are actually produced so these copies can propagate sepa-
rately in the network, each copy being carried by a distinct
mobile node. This approach is clearly a costly one but it
helps preventing message loss, which can occur because of
transmission failures (when a message is transmitted wire-
lessly between two neighbor nodes), or because the carrier
of a copy may unexpectedly disappear from the network.

Many solutions have been proposed to keep the cost of epi-
demic routing at a reasonable level, using heuristics that
basically aim at reducing the number of carriers for each
message, and also sometimes at selecting the “best” carri-
ers for each message. Some solutions rely on probabilistic or
semi-probabilistic heuristics [27] [31] [40], or take into account
the context of each mobile node [3] [5] [33]. Others assume
that mobile nodes are carried by human beings, whose social
interactions can be captured and used to drive message for-
warding by predicting how people move or meet [29] [34], or
by identifying what communities each person belongs to [12]
[23] [24].

In this work we make no assumption about the carriers of
mobile nodes. These carriers could thus be human beings,
but they could as well be vehicles, robots, animals, or any
combination of these. The system we define relies on a se-
lective epidemic dissemination model, which in essence can
be considered as an effective implementation of the abstract
model described in [16].

3. SYSTEM MODEL

3.1 General architecture

The system model we consider consists of a finite set of
mobile nodes \( V \). Each node features a short-range wireless
interface that allows it to exchange messages in ad hoc mode
—that is, without relying on any fixed infrastructure— with

nodes in its radio range.

At any time a node is either up or down: a node can
disable its wireless transceiver or enter standby mode spon-
taneously to save battery, or it can be switched off and on
alternatively by a user. This behavior is normal and is not
considered as a failure. When a node is down, it cannot
communicate with its neighbors. When it is switched on
again, its previous state is restored, and it initiates a neigh-
bor discovery phase in order to adjust rapidly to its current
surroundings.

No assumption is made about the mobility of nodes, or
about their spatial distribution. A node can thus be some-
times isolated from the other nodes, when the distance to
its closest neighbor exceeds its radio range.

The relations between nodes take place over a time span
\( T \). At some time in \( T \) the system model is represented by
the static graph \( G = (V, E) \), where \( V \) denotes the set of nodes and
\( E \) the set of edges. There exists an edge between two
nodes \( u \) and \( v \) if \( u \) and \( v \) are within mutual radio range, and
can thus exchange messages over the wireless medium.

Over time, the system model can be represented by a time-varying
graph (TVG) \( \mathcal{G} = (V, E, T, \rho, \psi) \), where \( \rho : E \times T \to \{0,1\} \)
is the edge presence function that indicates whether a given
message is available at a given time; and \( \psi : V \times T \to \{0,1\} \)
is the node presence function that indicates whether a given
node is available at a given time. At some time in \( T \), the
underlying graph \( G \) of \( \mathcal{G} \) is not necessarily connected. In
fact, the underlying graphs of \( \mathcal{G} \) may all be disconnected
over \( T \).

3.2 Communication model

In the TVG formalism, a journey in \( \mathcal{G} \) is defined by a
sequence of couples \( \{(e_1,t_1),(e_2,t_2),\ldots,(e_n,t_n)\} \), such that
\( \{e_1,e_2,\ldots,e_n\} \) is a walk in \( G \) and \( \rho(e_i,t_i) = 1 \) and \( t_{i+1} > t_i \)
for all \( i < n \) [9]. At time \( t \), a message sent by a node \( u \)
may eventually be received by a node \( v \) if a journey \( u \rightsquigarrow v \)
exists in the TVG \( \mathcal{G} \), with a starting date \( t_1 \) such that
\( t_1 > t \). In other words, a message sent by a node \( u \) may reach \( v \)
after being stored, carried, and forwarded successively by
several nodes, even if an end-to-end path between \( u \) and
\( v \) never exists in any underlying graph \( G \) of \( \mathcal{G} \). The time
elapsed between two consecutive hops (\( t_{i+1} - t_i \)) may range
between a few milliseconds (the message being forwarded
rapidly between successive neighbor nodes) and minutes or
hours (the message being carried for a while before being
forwarded to another node).

In our system model, any message sent by a node dissemi-
nates in the network according to the epidemic model: when
two nodes meet, they can seize this opportunity to create
new copies of the messages they are carrying, thus increas-
ing the number of carriers for these messages. This basic
interaction scheme takes inspiration from the Autonomous
Gossiping (A/G) algorithm [16], which itself defines a selec-
tive version of the epidemic routing model proposed in [37].
In our model, each node defines an interest profile that determines the kinds of messages it is willing to collect from other nodes, and for which it is therefore willing to serve as a mobile carrier.

The epidemic model increases the probability of message delivery, as it simultaneously exploits all possible journeys that involve nodes whose interest profiles match the kind of message considered. In practice, messages that match an interest profile \( p \) can be carried only by a subset of nodes \( V_p \subseteq V \), which consists of all nodes in \( V \) whose interest profile is \( p \) (or larger than \( p \)). Possible journeys for copies of a message that matches \( p \) are defined in \( G_p = (V_p, E, T, \rho, \psi) \).

### 3.3 Fault model

The nodes in our system model can be switched off and on at any time. When a node is switched off this is not considered as a “failure”, for this is consistent with its normal behavior. Nodes may sometimes crash spontaneously and never recover, though, but this is assumed to be an exceptional event, concerning only a very small number of nodes. Using the TVG formalism, the notation \( \psi(t) = 0 \) is used to indicate that the node \( v \) is switched off at time \( t \).

In an opportunistic network, any wireless link between two neighbor nodes is inherently transient. However, journeys between any pair of nodes do not require any temporal changes to end-to-end connectivity between these nodes, and can thus rely on successions of transient links. As a consequence, disruptions of wireless links are not considered as “link failures” in our model.

The epidemic routing model cannot guarantee that each message sent in the network eventually gets delivered to all possible recipients. Our system model therefore admits receive omissions, and these are considered as benign faults.

We do not consider Byzantine faults in this system model: an active node is assumed to behave properly, and to recover appropriately after being switched off and on again. Furthermore, messages transferred wirelessly between neighbor nodes are assumed to be unaltered. More specifically, whenever a message is received from a neighbor node, the integrity of the received copy is checked, and this copy is simply discarded if it has been altered during the transmission. The epidemic model ensures that the receiver will find other opportunities to obtain the message anyway (possibly again from the same neighbor).

### 3.4 Assumptions

In an opportunistic network, no guarantee can be provided about message delivery ratios, transmission delays, node availability, etc. A few assumptions can however be made to limit uncertainties in our system model.

In the remainder of this paper we call a session the process that consists in starting a consensus agreement procedure, and pursuing this procedure until a decision is made. Several sessions may of course progress simultaneously in the network, and some nodes may participate in several consensus sessions simultaneously.

We assume that a consensus session \( S \) involves a subset \( V_S \) such that \( V_S \subseteq V \) (i.e., all nodes in the network are not necessarily involved in \( S \)). Nodes in \( V_S \) are called participants for session \( S \). Each of these nodes is expected to provide an initial value for \( S \), and to help run the consensus algorithm until a decision is made. We assume that any node \( u \in V_S \) knows that it belongs to \( V_S \), and knows \( |V_S| \) (i.e., how many nodes participate in \( S \)). The definition and creation of the subset \( V_S \) is application-dependent, and is therefore out of the scope of this paper.

A consensus session \( S \) spans over a time period \( T_S \). During some periods in \( T_S \), some nodes in \( V_S \) may disappear temporarily from the network. We assume that these nodes will eventually reappear during \( T_S \), and resume their activity regarding \( S \). Some nodes in \( V_S \) may also crash and disappear definitively from the network. We assume that the number of crash failures for nodes in \( V_S \) is unknown but can be bounded, and that the bound is low with respect to \( V_S \).

Nodes in \( V_S \) are all expected to serve as mobile carriers for messages pertaining to \( S \). Additionally, any node \( u \in V \setminus V_S \) can also serve as a benevolent carrier for messages pertaining to \( S \). Nodes that can carry messages pertaining to \( S \) thus belong to a subset \( V \setminus V_S \) (carriers for session \( S \)) such that \( V_S \subseteq V \setminus V_S \subseteq V \). These nodes all exhibit an interest profile \( p \) that matches messages pertaining to \( S \).

Any message pertaining to \( S \), whatever its sender \( u \in V_S \), is assumed to disseminate thanks to nodes in \( V \setminus V_S \), and eventually reach all or some of the nodes in \( V_S \). The reason why only some of the nodes in \( V_S \) are considered here is a consequence of the epidemic routing model, which cannot guarantee that each message eventually reaches all its possible recipients. The failure ratio is assumed to be low and bounded, though.

Finally, we assume that, for any given node \( u \in V_S \), there are some periods in \( T_S \), called “good periods for \( u \)” during which all the messages sent by \( u \) eventually reach all other nodes in \( V_S \) (except nodes that crashed during \( T_S \)). This assumption is not required to ensure the termination of the OTR algorithm, but it makes it possible to terminate faster in some cases: as soon as node \( u \) decides, then if \( u \) is in a “good period” its decision can be transmitted to all other nodes in \( V_S \). As observed in [7], assuming good periods in an asynchronous system is often more realistic than assuming that the system is partially synchronous.

### 4. SOLVING CONSENSUS WITH THE OTR ALGORITHM

#### 4.1 Consensus

The consensus problem over a set \( \Pi = \{p_1, p_2, \ldots, p_n\} \) of processes (which in our system model are called participants and are assumed to run on distinct nodes) is defined by the following properties:

- **Validity**: Any decision is the initial value of some participant.
- **Agreement**: No two participants decide differently.
- **Termination**: All correct participants eventually decide.

#### 4.2 Overview of the Heard-Of model

A computation in the Heard-Of (HO) model evolves in asynchronous communication-closed rounds, without any need for a failure detector. In each round, each process sends a message to all the other processes and then waits...
to receive similar messages sent in the same round. Late messages pertaining to former rounds are discarded. The features of a specific system are captured by a communication predicate, which is expressed in terms of Heard-Of sets: \( HO(p, r) \) represents the set of processes from which process \( p \) “hears of” (i.e., receives some messages) at round \( r \). A consensus problem is solved in the HO model by a Heard-Of machine defined by a pair \( M = (A, P) \) where \( A \) is an algorithm and \( P \) is a communication predicate.

Several consensus algorithms have been expressed in the HO model \[13\]. These algorithms can hardly tolerate message loss, except for the Paxos/LastVoting (P/LV) algorithm and the One-Third Rule (OTR) algorithm. An implementation of the P/LV algorithm for mobile ad hoc networks has been proposed in \[7\]. This implementation could not run in an opportunistic network, though, as it is coordinated-based which requires temporal one-to-one connectivity between the coordinator and all the other processes.

### 4.3 Overview of the OTR algorithm

The One-Third Rule (OTR) algorithm is a perfect candidate for opportunistic computing. In \[13\] it is defined with the formalism of the HO model, but similar structure and decision conditions can be observed in other algorithms (e.g., first round in \[8\]). Each round \( r \) in the OTR algorithm consists of two steps (see Algorithm 1): a sending step \( S_p^r \) in which process \( p \) sends its current contribution (for round \( r \)) to the other processes (line 3), followed by a transition step \( T_p^r \), in which, provided it has received enough contributions from the other processes (line 5), process \( p \) either takes a decision (line 8) or determines its contribution for the next round (line 6) and proceeds to that round (line 7).

#### Algorithm 1: The One-Third Rule algorithm

**Initialization:**

1. \( x_p \leftarrow v_p \) \( \{v_p \text{ is the initial value of process } p\}\)

**Round \( r \):**

2. \( S_p^r \)
3. send \( <x_p> \) to all other processes
4. \( T_p^r \)
5. if \( |HO(p, r)| > 2n/3 \) values then
6. \( x_p \leftarrow \text{the smallest most often received value} \)
7. if more than 2n/3 values received are equal to \( x \) then
8. DECIDE(\( x \))

A node \( p \) can tolerate not to receive messages from up to one-third of the other participants in round \( r \), while still being able to decide or proceed to the next round. In practice, in an opportunistic network this may occur because some of the other participants have not reached round \( r \) yet (for example because these participants are currently in suspend mode), or because some participants have indeed sent their contributions for round \( r \) but these messages have not reached node \( p \) yet (and possibly never will).

Moreover, with the OTR algorithm a consensus computation involving \( n \) participants can progress from round \( r \) to the next if at least one participant can receive contributions (pertaining to round \( r \)) from more than \( 2n/3 \) other participants. In an opportunistic network where message delivery can sometimes be delayed significantly (at least for some receivers), this property of the OTR algorithm is an asset, for the consensus computation can proceed from one round to the next as soon as one node has received enough contributions to do so.

Conversely, at least \( \frac{2n}{3} \) participants must send contributions in each round, since this is a requirement for the algorithm to proceed to the next round or to the final decision. Based on these observations the assumptions made in our system model (Section 5.1) can be refined so as to account for the specific requirements of the OTR algorithm:

- At each round \( r \), the number of participants sending their contribution in the network should not be smaller than \( \frac{2n}{3} \) (which means that about \( \frac{n}{3} \) participants can actually “skip” a round without preventing the computation to progress). By extension, the number of crash failures among the participants must be smaller than \( \frac{n}{3} \).
- At each round \( r \), message loss should be such that at least one node can receive more than \( \frac{2n}{3} \) contributions (which means that receive omissions are admitted for most of the participants but one, which should be able to receive enough contributions to proceed to the next round). Using the HO formalism, this assumption can be expressed as:

\[
\forall r, \exists p \in \Pi, s.t. |HO(p, r)| > 2n/3
\]

Note that these requirements fit perfectly with the characteristics of an opportunistic network, in which node availability and message delivery cannot be guaranteed. Moreover, the n-to-n communication pattern used in the OTR algorithm is satisfied by the epidemic routing model, since with this model sending a message to many or all nodes is not really different from sending a message to a single node.

According to \[13\] the communication predicate \( P_{(C_0)\infty} \) associated with the OTR algorithm ensures that the consensus can be solved if there exists a round \( r_0 \) where \( C_0 \) holds:

\[
\exists \Pi_0, s.t. |\Pi_0| > 2n/3, \forall p \in \Pi: HO(p, r_0) = \Pi_0
\]

where \( \Pi \) stands for the set of participants involved in the consensus. In other words, during \( r_0 \) all participants must be able to receive contributions from the very same subset of more than \( \frac{2n}{3} \) participants (with \( n = |\Pi| \)), so they can make the same decision. This predicate can be expressed nicely in terms of HO sets, and it is sufficient to ensure that consensus is solved. Yet it does not define a necessary condition. If there exists a round \( r_0 \) where the contributions collected by participants are such that the smallest most frequent value is the same for all participants, then the consensus can be solved as well, even though all participants may not have received these values from the same contributors. This condition can hardly be expressed in terms of HO sets, yet it is weaker than predicate \( P_{(C_0)\infty} \) and allows more flexibility in the system model.

### 5. Opportunistic Implementation of the OTR Algorithm

Our system is architected in two layers: the lower layer can support network-wide, content-driven message dissemination based on controlled epidemic routing, and the upper layer is an implementation of the OTR algorithm that interfaces with the communication layer.
5.1 Opportunistic communication layer

The communication layer implements a content-driven message dissemination model. It can actually be perceived as an effective implementation of the abstract model described in [10]. An overview of this communication layer is provided below. Further details can be found in [21].

Each node periodically broadcasts an announce in order to inform its neighbors (if any) about its presence. When two nodes meet they first exchange their interest profiles, that characterize the types of messages each node is willing to receive, and for which it is therefore willing to serve as a mobile carrier. Based on this exchange of profiles, each node can determine accurately which messages could be of interest to its neighbor, and make an offer accordingly. Messages are then effectively exchanged through a succession of query-and-reply cycles.

Each node maintains a local cache to store the messages it carries, so they can be proposed to any new neighbor. In order to prevent network congestion, each message is allowed a specific lifetime. When this lifetime is over, all copies of this message are removed from the caches of mobile nodes, so this message actually stops disseminating in the network. The dissemination of a message can also be canceled explicitly on a mobile node. Once a message is canceled, the node does not propose it to any neighbor anymore, and if conversely a neighbor actually offers to provide this message, this neighbor is notified that it too should cancel the message whose dissemination is not required anymore. This approach is referred to as network healing in the literature. It is an effective way to limit the cost of epidemic routing in an opportunistic network, using either Passive Cure [22] or Active Cure techniques [55].

The communication layer provides a publish/subscribe application programming interface (API), presented in Algorithm 2. With this API, the content-driven nature of message selection is based on the notion of group. A group, identified by a group identifier (grpId), is a set of nodes that cooperate in a common task and are thus potentially interested in the same kind of messages. The interest profile of a node is a compilation of the ids of all the groups it belongs to. Function subscribe (line 2) allows a process to specify that it is interested in receiving the messages for a given group. This function adds the specified group id to the node’s interest profile, and consequently the node will try to collect messages addressed to this group from any neighbor it will meet thereafter. Whenever a message is received that matches a group the node belongs to, a receive event is triggered (line 3) accordingly.

Algorithm 2 The communication layer API

1. Function publish (msgId, grpId, sndId, BODY, [din])
2. Function subscribe (grpId)
3. Event receive (msgId, grpId, sndId, BODY, [din])
4. Function cancel (msgId)
5. Function relay (grpId)

The communication layer assigns a unique identifier to each node. This identifier can for example be the IMEI on a smartphone, or an auto-configured link-local IPv6 address. Each message must likewise be assigned a unique identifier, which will notably be used by the system to detect duplicates, and thus prevent useless transfers of message copies between neighbor nodes. Sending a message in the network is done with the publish function (line 1), that takes as parameters identifiers for the message itself (msgId), for its sender (sndId), and for the group of nodes it is addressed to (grpId). The body of the message is just perceived as a payload by the system.

When a message is published on a node, or received from a neighbor, this message is deposited in the local cache. Afterwards every contact with a new interested neighbor is an opportunity to transfer a copy of the message to that neighbor.

As explained above, each message can optionally be assigned a set lifetime when it is published (last parameter in line 1), and the dissemination of a message can also be canceled explicitly (line 4). Both mechanisms, if used wisely, can help reduce the cost of epidemic routing.

By default a message disseminates by being stored, carried, and forwarded by nodes that have subscribed to the group this message is addressed to. A node can however be configured so as to serve as a benevolent carrier for messages addressed to a group it does not belong to. With the API this is obtained through the relay function (line 5), which basically has the same effect as function subscribe, except that messages received by a benevolent carrier will not trigger any receive event on that node.

5.2 Opportunistic OTR algorithm

Our implementation of the OTR algorithm based on the opportunistic communication layer is shown in Algorithm 3. A consensus session is initiated by calling function startSession, taking the group identifier, the number of participants in this group, and the initial value for the local participant as parameters. A subscription is then set for the group (line 8) before starting the first round of the OTR algorithm (line 9).

At each round, the current contribution is published (line 12), and the identifier of the message hence published is recorded in contribIds so it can be canceled later (line 13).

When a contribution is received from another participant, the behavior of the receiver depends on whether this contribution pertains to a new round (lines 18-22), to the current round (lines 24-30), or to a former round (line 32). In the first case the receiver cancels messages pertaining to the current round before moving to the new round. In the second case it checks if enough contributions have been received to either make a decision (line 29) or start the next round (line 30). In the third case it simply discards the contribution it has just received, but cancels the corresponding message so as not to take part in its dissemination (line 32).

Note that message cancellation is here used systematically as a means to prevent messages that pertain to former rounds to keep propagating in the network. This form of cross-layering between the OTR algorithm and the opportunistic communication layer helps reduce the cost of epidemic message dissemination.

Function decide is called when a decision is made locally (line 29), or when a decision message is received from another participant (line 41). In any case this function is run only once on each node (line 33): all pending messages are canceled (lines 36-37) and the decision is published (line 40) with a unique message identifier that is produced using grpId as a seed (line 39). Thus, if several nodes make a decision in the same session, all messages carrying this decision will have the same identifier and will thus be considered as duplicates of the same message by the communication layer.
As explained in Section 3.2, publishing the decision is not required by the OTR algorithm, but it can help terminate a consensus session faster.

### Algorithm 3: Opportunistic version of the OTR algorithm

**Initialization:**

1. \( id_{gen} \leftarrow \text{id of local node} \) \{ must be unique in the network \}
2. \( contrib_{p} \leftarrow \{ \} \) \{ contributions received for \( r_{p} \) (multi-set) \}
3. \( contribIds_{p} \leftarrow \{ \} \) \{ ids of messages received during \( r_{p} \) (set) \}

**Function** \( \text{startSession}(grpId, nbNodes, v) \):

4. \( x_{p} \leftarrow v \) \{ initial value for node \( p \) \}
5. \( grpId_{p} \leftarrow grpId \)
6. \( nbNodes_{p} \leftarrow nbNodes \)
7. \( solved \leftarrow \text{false} \)
8. \( subscribe(grpId_{p}) \)
9. \( startRound(1) \)

**Function** \( \text{startRound}(r) \):

10. \( msgId \leftarrow genId() \) \{ call id generator \}
11. \( publish(msgId, grpId_{p}, id_{p}, \text{CONTRIB}(r_{p}, x_{p})) \)
12. \( contribIds_{p} \leftarrow contribIds_{p} \cup \{ msgId \} \)
13. \( solved \leftarrow \text{false} \)
14. \( cancel(msgId) \)
15. \( switch(r) \)
16. \( case(r > r_{p}); \)
17. \( \quad \text{for } id \in contribIds \text{ do} \)
18. \( \quad \quad cancel(id) \)
19. \( \quad contrib_{p} \leftarrow \{ x_{p} \} \)
20. \( \quad contribIds_{p} \leftarrow contribIds_{p} \cup \{ msgId \} \)
21. \( \quad startRound(r) \)
22. \( case(r = r_{p}); \)
23. \( \quad contrib_{p} \leftarrow contrib_{p} \cup \{ x_{p} \} \)
24. \( \quad contribIds_{p} \leftarrow contribIds_{p} \cup \{ msgId \} \)
25. \( case(r < r_{p}); \)
26. \( \quad if \left( \left| contrib_{p} \right| > 2/3 \ast nbNodes_{p} \right) \text{ then} \)
27. \( \quad \quad x_{p} \leftarrow \text{smallest most often received value in } contrib_{p} \)
28. \( \quad \quad if \text{ all values are equal to } x \text{ in } contrib_{p} \text{ then} \)
29. \( \quad \quad \quad \text{decide}(\top) \)
30. \( \quad \quad startRound(r_{p} + 1) \)
31. \( \quad cancel(msgId) \)
32. \( \text{Function} \ \text{decide}(v); \)
33. \( \quad solved \leftarrow \text{true} \)
34. \( \quad x_{p} \leftarrow v \)
35. \( \quad \text{for } id \in contribIds \text{ do} \)
36. \( \quad \quad cancel(id) \)
37. \( \quad contribIds_{p} \leftarrow \{ \} \)
38. \( \quad msgId \leftarrow genId(grpId) \)
39. \( \quad publish(msgId, grpId_{p}, id_{p}, \text{DECISION}(x_{p})) \)
40. \( \text{Upon receive} \ (msgId, grpId_{p}, sndId, \text{CONTRIB}(r_{p}, x)) \text{ do} \)
41. \( \quad \text{decide}(v) \)

### 6. EXPERIMENTAL EVALUATION

Evaluating the performance of a system capable of running in opportunistic networks is a challenge. In the literature, protocols and systems designed for such networks, and more generally for Delay/Disruption-Tolerant Networks, are often evaluated using simulators, and little or no effort is devoted to producing code that can be used in a real setting. Yet, as observed in [26] “rare are the [DTN] protocols that were implemented, tested in real-life and proven to be free of lethal stealthy assumptions”. A salient feature of our system is that it has been fully implemented, and validated in real conditions using a small flotilla of smartphones as mobile nodes.

#### 6.1 Experimentation conditions

In order to demonstrate that our system can indeed solve consensus in an opportunistic network, volunteers have been equipped with HTC smartphones, whose Wi-Fi chipsets were configured to operate in ad hoc mode. Each smartphone ran a small Android application (named \( iAgree \)) based on our system. This simple application allows a user to initiate new consensus sessions, and to join, participate, and display the status of ongoing and past sessions. The volunteers were asked to carry their smartphone while roaming the laboratory building or its surroundings, and to use application \( iAgree \) every now and then. Trace logs were collected and analyzed after the end of the experiment.

The experiment spanned over 8 hours and involved a small population of 7 volunteers (and as many smartphones). It was mostly meant to serve as a proof of concept. A comprehensive evaluation of the system’s effectiveness and efficiency would require a far larger population of mobile nodes, and should ideally span over several days or weeks.

During this experiment the system was configured to give each message a lifetime of 12 hours. No message was therefore removed because of an exhausted lifetime. Moreover, because only 7 smartphones were available for this experiment, none of them was configured to serve as a benevolent carrier: every smartphone was systematically enrolled as a consensus participant.

#### 6.2 Results

3424 radio contacts occurred between smartphones during this 8 hour experiment, with an average contact duration of 162 seconds. The cumulative distribution of contact durations is presented in Figure 2. It can be observed that almost 60% of radio contacts lasted for less than a minute, which confirms the transient nature of the radio contacts established between smartphones.

A timeline of the evolution of the average number of neighbors is presented in Figure 3 and the cumulative distribu-
It can be observed that each smartphone had at most one neighbor during about 40% of the experiment’s duration, but was actually alone (i.e., with no neighbor) during more than 20% of that time. Moreover, there was no period of time during which all smartphones were connected all together (each smartphone would otherwise have detected 6 neighbors simultaneously). These results confirm the dynamic and disconnected nature of the network, whose topology changed continuously and rapidly during the whole experiment.

151 consensus sessions were initiated by users (either sequentially or concurrently) during the experiment, and running these sessions led to the exchange of 4530 contributions among the smartphones.

Figure 5 shows the distribution of the number of hops required for contributions to reach each smartphone. For example, 50% of the contributions received by smartphone SP6 were received directly from the sender (1-hop), 33% contributions were received by SP6 after 2 hops, etc. The smartphones actually received most of the contributions after multi-hop trips during the experiment, a few of these trips requiring up to 6 hops between sender and receiver.

This observation confirms the interest of multi-hop relaying between smartphones, but it is not sufficient to demonstrate the interest of the store, carry and forward principle in an opportunistic network such as that formed by the smartphones during the experiment. In order to clarify this point, Figure 6 illustrates how one particular consensus contribution sent by smartphone SP2 actually disseminated during the experiment. A few minutes after this particular contribution was published (i.e. sent to all other nodes) by our system on SP2, a radio contact was established with SP6, which thus got a copy of the contribution and became a new carrier for this contribution. SP2 later managed to forward the contribution to SP7, while SP6 forwarded it to SP1 first, and later to SP3. The contribution thus kept disseminating, until it reached the last smartphone SP5, almost half an hour after it was initially published on SP2.

These results confirm the need to rely on delay/disruption-tolerant multi-hop forwarding to ensure information dissemination in an opportunistic network. Any consensus algorithm requiring temporal and/or end-to-end connectivity would be completely ineffective in such a network. Since consensus contributions had to propagate opportunistically between the smartphones, each consensus session could take a while before a decision was made. Actually, 9% of the sessions could not reach a decision during the experiment, but these sessions were initiated shortly before the end of the experimentation period. Figure 7 presents the cumulative distribution of the execution times for all completed sessions. The execution time of a session is here defined as the time elapsed between the sending of the first contribution by one of the participants, and the last decision made or received by any of the participants. It can be observed that a few sessions were completed in only a few seconds, but most of them required several minutes, and some even took a couple of hours to complete. These results clearly show that consensus solving in an opportunistic network can indeed take a while, but is perfectly feasible provided an appropriate communication model is used.

Figure 8 shows the cumulative distribution of the average number of rounds required to solve consensus in all completed sessions. It can be observed that consensus was obtained in only one round for about 5% of the sessions. Most sessions required a couple of rounds to complete, though, and it is likely that more rounds would have been required with a larger population of consensus participants.
Figure 8: Cumulative distribution of the number of rounds required to solve consensus

All these results globally demonstrate that our system, which combines the OTR algorithm with an opportunistic communication layer, is effective at solving consensus in an opportunistic network. As mentioned above this short experiment was only meant to serve as a proof of concept. Further experiments would be required to observe how this system can perform in far larger networks involving hundreds of mobile nodes, roaming large areas over long time spans. Larger experiments would notably make it possible to observe phenomena that could not be highlighted with a flotilla of only 7 smartphones, such as the effect of relying on benevolent carriers (i.e., mobile nodes that carry messages they are not directly interested in) to carry consensus contributions, or the effect of receive omissions when consensus sessions involve many participants.

7. CONCLUSION

The One-Third Rule (OTR) algorithm is an elegant solution to solve consensus in networks where message loss can occur. Being based on the Heard-Of (HO) model, it is well suited to support transient process and link faults, which makes it an ideal solution to solve consensus in opportunistic networks. In such networks, messages propagate by being carried physically by mobile carriers whose mobility is usually neither planned nor controlled, so there is no guarantee that messages finally get delivered to their destinations.

In this paper we have presented a system that combines an implementation of the One-Third Rule (OTR) algorithm with a communication layer that supports network-wide, content-driven message dissemination based on controlled epidemic routing. Experimental results obtained with a small flotilla of smartphones confirm that this system is effective at solving consensus problems in an opportunistic network. The source code of this system is now distributed under the terms of the GNU General Public License.

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9. REFERENCES

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