

# Adaptive Random Re-Routing in Sensor Networks

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*Abstract*—Much of the traffic carried by Sensor Networks will originate from routine measurements or observations by sensors which monitor a particular situation, such as the temperature and humidity in a room or the infrared observation of the perimeter of a house, so that the volume of routine traffic resulting from such observations may be quite high. When important and unusual events occur, such as a sudden fire breaking out or the arrival of an intruder, it will be necessary to convey this new information very urgently through the network to a designated set of sink nodes where this information can be processed and dealt with. This paper addresses the important challenge of avoiding that the volume of routine background traffic creates delays or bottlenecks that impede the rapid delivery of high priority traffic resulting from the unusual events. Specifically we propose a novel technique, the “*Randomized Re-Routing Algorithm (RRR)*”, which detects the presence of novel events in a distributed manner, and dynamically disperses the background traffic towards secondary paths in the network, while creating a “fast track path” which provides better delay and better QoS for the high priority traffic which is carrying the new information. When the surge of new information has subsided, this is again detected by the nodes and the nodes progressively revert to best QoS or shortest path routing for all the ongoing traffic. The proposed technique is evaluated using a mathematical model as well as simulations.

## I. INTRODUCTION

Wireless sensor networks (WSNs), and sensor networks (SNs) in general, must forward significant data promptly and efficiently from the different sensors where the data originates to one or more sinks in the network at which data is collected and where it may also be interpreted. In such networks: (1) routine measurements and sensing take place constantly resulting in a steady volume of data being transmitted towards the sink(s), and (2) unusual events of particular interest will occur unexpectedly, and the information related to such events will require fast transmission to the sink(s). While routine data are essential for reporting on the conditions that the SN is monitoring, unusual events are more critical and need a faster or “better QoS” treatment by the network, such as short delay, very low loss, possibly high bandwidth, better security, etc..

In this paper we propose an adaptive technique which we call **Randomized Re-Routing** (RRR) for addressing these needs of SNs. During network operation, the network nodes send out packets of sensor readings and maintain the running average value of the measurements. If the packet’s contents are very similar to those of the running average for the same connection; the node then inserts a ZERO bit header in that packet. Otherwise, the packet whose content differs significantly from the running average value will be classified as being “unusual” and be inserted a ONE bit in its header. Each node also

monitors the rate at which it receives “unusual” packets, and if this rate does not exceed a threshold  $\tau_0$ , then the node forwards all packets it receives along their preferred (e.g. shortest or best QoS) path towards their destinations. Obviously the preferred path may be determined by criteria such as the minimum delay, greatest security, lowest power consumption, smallest loss etc. On the other hand, if a node (source or transit) senses that the rate at which it forwards unusual packets exceeds the threshold  $\tau_0$ , then it will forward all ONE-bit packets along the best QoS path to their destination, while all ZERO-bit carrying packets will be directed along a randomized route which spreads the lower priority traffic across the network away from the high priority paths, reserving the better paths to the high priority traffic.

We will present this algorithm in detail and provide an evaluation of its effectiveness using both an analytical model and simulations. In particular, we will use a mathematical approach based on diffusion approximations [1], [2] to estimate the resulting packet travel delays from source to destination.

The remainder of the paper is organized as follow. In Section II, we present some related work in the area. In Section III, we describe the mathematical model for analyzing the packet travel delay. In Section IV, we present the RRR algorithm for providing better quality of service to unusual events in SNs. Sections V and VI summarize the numerical and simulation results that we have obtained, and we conclude the paper in Section VII.

## II. RELATED WORK

QoS becomes a crucial element as networks move to an environment inherently more susceptible to delay, interference, etc. [3], [4], [5]. With the popularity of wireless and sensing technologies, several real-time communication protocols have been studied for sensor networks. He et al. [6] propose SPEED, a protocol which combines feedback control and non-deterministic QoS aware geographic forwarding. Lu et al. [7] describe a packet scheduling policy, called Velocity Monotonic Scheduling, which inherently accounts for both time and distance constraints. Felemban et al. [8] propose Multi-path and Multi-Speed Routing Protocol (MMSPEED) for probabilistic QoS guarantee in WSNs. Multiple QoS levels are provided in the timeliness domain by using different delivery speeds, while various requirements are supported by probabilistic multipath forwarding in the reliability domain; our approach has some similarity to this work. Huang et al. [9] consider a spatiotemporal multicast protocol, called

“mobicast”, which provides reliable and just-in-time message delivery to mobile delivery zones. Ergen et al. [10] present a routing algorithm that maximizes the sensor network lifetime, and further incorporates delay guarantees into energy efficient routing by limiting the length of paths from each sensor to the collection node. Sankarasubramaniam et al. [11] propose a Event-to-Sink Reliable Transport (ESRT) protocol in Wireless Sensor Networks, which can served as a congestion control protocol.

In our work we focus on the quality of service in forwarding routine data and unusual events in SNs, and consider how to manage routing so that network capacity within the SN is created so as to offer uncongested paths to traffic emanating from unexpected events, in addition to routing the routine parts of the traffic.

Some congestion control algorithms have also been proposed for wireless sensor networks. Most of the existing work provides congestion control and avoidance in wireless SNs by backpressure and limiting the transmission rate, or by checking the congestion states of neighboring nodes [12], [13], [14]. These mechanisms require feedback from sensor nodes which result in extra overhead in the network. The RRR approach we propose is quite simple and easy to implement in a distributed manner since all decisions are locally taken by the nodes, and it requires no feedback messages from the congested nodes.

### III. MODELING THE PACKET TRAVEL DELAY

In this section we present a model of a network with sensors distributed over some open or built area. In order to evaluate the travel delay of packets in the SN, we will use an approach based on diffusion approximations developed in [15], [16], [17], which is also detailed in [18]. This approach was first used to compute travel delays in wireless sensor networks in [1], [2] and we briefly recall it in this section.

We model the location of nodes and the relays of packets on the path from source to destination as follows. At time  $t$  we assume that a packet is located at some node which is positioned at a distance measured in minimum number of hops  $X_t$  from the destination. Time changes from hop to hop are represented by an increment of  $\Delta t > 0$ . The value  $X_0 = D$  indicates that the packet is at the source with an initial distance  $D$  to the sink at time 0, while  $X_t = 0$  indicates that the packet is at the sink at time  $t$  so that any further transmission of that particular packet ceases. The model allows for the loss of packets, and this is represented by a loss rate (loss per unit time)  $f$ . Furthermore, we assume that there is some acknowledgement mechanism that informs the source nodes about the correct reception of packets at the sinks; in a wireless sensor network this can be achieved, for instance, via a reserved wireless frequency that the sinks use to broadcast a message saying “packet so-and-so has been received”, but it could also be achieved by packets going upstream in a multi-hop manner from the sinks towards the destinations.

Let  $x$  be the instantaneous value of  $X_t$ , and let  $b(x)$  be the average speed at which the packet moves towards the destination at distance  $x$  from the sink. Clearly if  $b(x) < 0$

then the motion is towards the sink, and if  $b(x) > 0$  then it is drifting away from the sink. Similarly, we have a non-negative second moment parameter  $c(x) \geq 0$  which represents the variance of the motion per unit time and is strictly positive if the motion has a random component. If  $c(x) = 0$  means that the packet is moving in a deterministic manner.

In a well designed system, we would expect that packets carrying data from unusual events would have a negative  $b(x)$  and a small  $c(x)$ , while lower priority routine packets could possibly have a larger value of both  $b(x)$  and  $c(x)$ . Apart from that,  $b(x)$  and  $c(x)$  also depend on the total traffic arrival rate at the node which is given by  $\tau \geq 0$ , which is the total rate of arriving packets, except for the packet that is being considered.

Based on our definitions of  $b(x)$  and  $c(x)$ , the delay  $E[T]$  for a packet to travel from its source to destination has been derived in [1], [2] as:

$$E[T] = 2D \frac{1 + \frac{f+r}{m} + \frac{f}{r}}{|b| + \sqrt{b^2 + 2c(f+r)}}, \quad (1)$$

where  $D$  is the distance from the source to the destination, and  $f\Delta t$  is the probability that the packet is lost in a small time interval  $[t, t + \Delta t]$ . Since losses can occur, we also assume that a source node use a time-out of average value  $1/r$  to decide when to retransmit a packet, and that after the time-out it actually waits another  $1/m$  time units on the average before actually retransmitting the packet that is assumed to have been lost. Note that the time-out will operate both when the packet is actually lost, and when the packet’s travel time has exceeded the time-out delay. The mathematical model assumes that both the time-out and the additional “safety time” are exponentially distributed.

### IV. RANDOMIZED RE-ROUTING (RRR)

In order to provide preferential treatment to ONE-bit packets, the RRR algorithm routes the two types of packets differently. More specifically, packets from unusual events are routed along the shortest paths, while the routine data are randomly shunted to slower secondary path.

Let us first describe a simple scheme that may be used by any node, and in particular by source nodes, to identify unusual events as indicated earlier. It suffices that each sensor node keeps a running average over a short time window  $T$  of the value of the measurement that it is sending; if the measurement at time  $t$  is very similar to the *average* evaluated over time  $[t - T, t[$  then the packet sent out at time  $t$  is marked with a ZERO-bit indicating a routine packet. If the measurement is significantly different from the previous average then the packet is marked as an unusual event and given a ONE-bit in its header. This simple mechanism allows for an on-line classification of each successive packet from a given source. However it can also be used by intermediate nodes if they themselves wish to decide whether a packet is a routine or unusual event packet, as long as they are able to distinguish between the source-destination pairs contained in the packets as well as keeping track of content values.

The RRR algorithm operates in the following manner at any source or intermediate node:

- If at time  $t$  the node locally observes a level of arriving traffic  $\tau_t$  which is below a given threshold  $\tau_0$ , then traditional geographic routing [19] is applied to all packets.
- However, if  $\tau_t > \tau_0$ , then ZERO-bit packets and ONE-bit packets, whether they arrive from some other node for forwarding, or are generated internally, will be routed differently as follows.
  - Each node  $i$  ranks its neighboring nodes  $i_1, \dots, i_H$  so that  $i_1$  is located closest to the sink in number of hops, and  $i_H$  is the one which is farthest away. Node  $i$  forwards ONE-bit packets to neighbors  $i_1, \dots, i_K$ , and
  - The node forwards all ZERO-bit packets to the remaining neighbors  $i_{K+1}, \dots, i_H$ . Note that in general we will select one of these output nodes at random among the given set, and also we may choose not to use some of the nodes at the tail end of the ranking, because they may lead to excessively long paths.

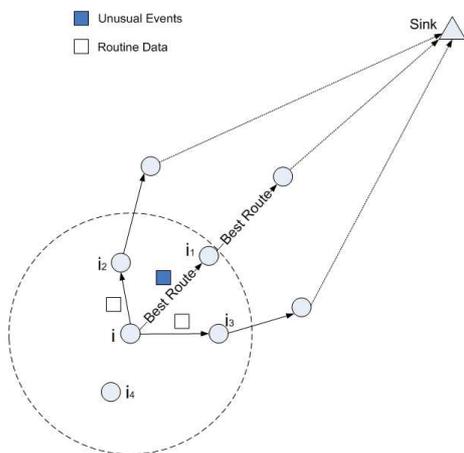


Fig. 1. Randomized Re-Routing for routine data and unusual events.

Figure 1 illustrates how the RRR algorithm operates. Here node  $i$  has four neighbors  $i_1, i_2, i_3$ , and  $i_4$  which are ordered according to their distances to the sink. When  $i$  receives a ONE-bit packet and forwards it to its neighbor which is closest to the destination, while it forwards ZERO-bit packets to  $i_2$  and  $i_3$  with equal probability. Note that  $i_4$  is not chosen as it is located further away from the sink than  $i$  itself. From this example, we see that our algorithm allocates the best route for transmitting important event data, while routine data are pushed aside to the remaining routes to achieve quality of service.

## V. PERFORMANCE ANALYSIS BASED ON THE DIFFUSION MODEL

Our evaluation includes both a theoretical component and simulations. We will present some additional analytical results here, followed by the simulation results in the next section.

Consider an example with  $D=10$ ,  $m=0.1$ ,  $\mu=30\text{pkt/s}$ ,  $\tau=5\text{pkt/s}$  and let us compute the travel delay of packets in a network with unusual events. When unusual events occur, network traffic increases suddenly and RRR is applied. The ONE-bit packets are routed along the shortest paths, while ZERO-bit packets are routed probabilistically along the remaining paths.

Define the advancement vector  $H = [-1, -1, 0, 0, 1, 1]$ , for a node with six neighbors, where the first two are one hop closer to the destination, two are at the same distance, and two are one hop further away from the sink than the current node. We use probability vectors for the direction taken by ONE-bit and ZERO-bit packets as  $P_U = [1, 0, 0, 0, 0, 0]$  and  $P_R = [0, 1/3, 1/3, 1/3, 0, 0]$ , respectively, indicating that ONE-bit packets are routed to the neighbor that is one hop closer to the sink, while the ZERO-bit packets are routed to the neighbors with -1, 0, and 0 hops of advancement with equal probability  $1/3$ .

Figure 2 shows the travel delay of RRR after unusual events occur. The travel delay of ONE-bit packets is much lower than that of the ZERO-bit packets. Similarly, Figure 3 and 4 show the travel delay of packets in RRR with  $D=10$ ,  $m=0.1$ ,  $\tau=10\text{pkt/s}$ ,  $\mu=50\text{pkt/s}$  and  $\mu=100\text{pkt/s}$  respectively. Again, they indicate that ONE-bit packets achieve shorter travel delay than ZERO-bit packets.

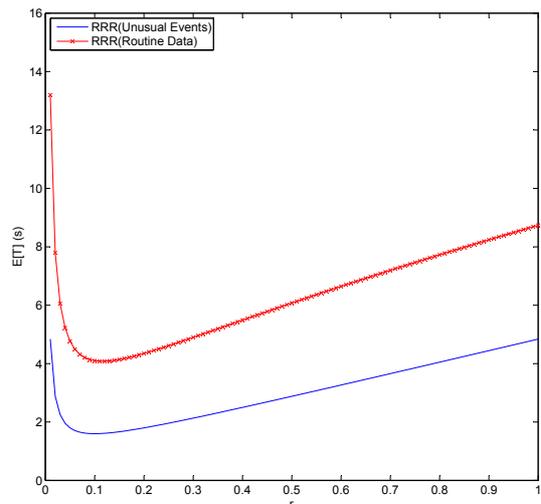


Fig. 2. After unusual events occur, travel delay with  $f=0.1$ ,  $D=10$ ,  $m=0.1$ ,  $\mu=30\text{pkt/s}$ ,  $\tau=5\text{pkt/s}$ .

### A. The traffic arrival rate $\tau$ at nodes

Even though RRR can provide better QoS to ONE-bit packets at the cost of lower QoS for the remaining packets, it does have the drawback of potentially increasing the total average arrival rate of packets as a whole at each node because ZERO-bit packets will visit more nodes than if they had taken the shortest path.

$\tau$  the total traffic arrival rate of packets per node depends on the packet arrival rates of unusual and routine packets. If

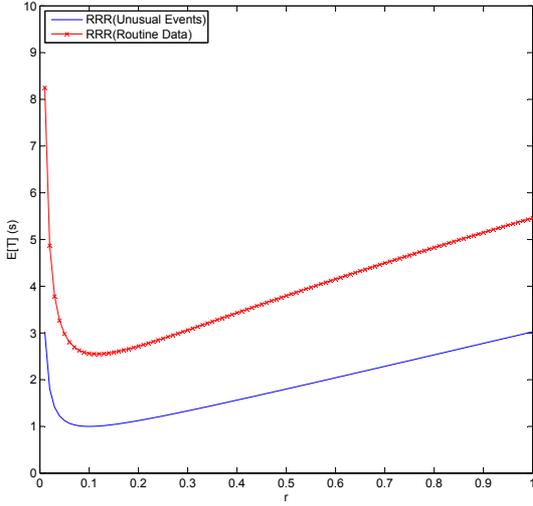


Fig. 3. After unusual events occur, travel delay with  $f=0.1$ ,  $D=10$ ,  $m=0.1$ ,  $\mu=50\text{pkt/s}$ ,  $\tau=10\text{pkt/s}$ .

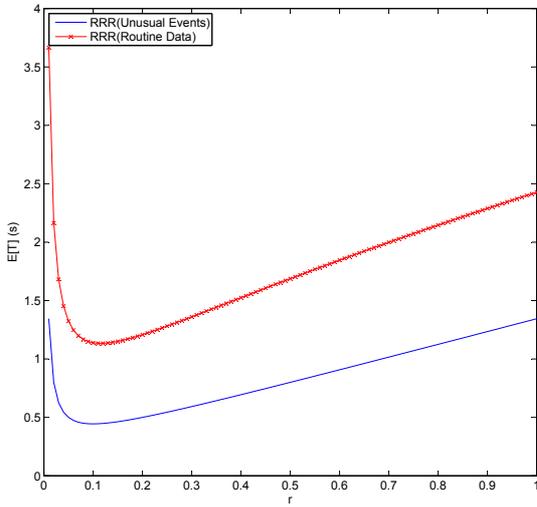


Fig. 4. After unusual events occur, travel delay with  $f=0.1$ ,  $D=10$ ,  $m=0.1$ ,  $\mu=100\text{pkt/s}$ ,  $\tau=10\text{pkt/s}$ .

the network as a whole has  $n$  nodes and there are  $s_R$  nodes that are sources of routine data at rate  $\lambda_R$ , with an average number of hops to the sink of  $d_R$ , the total average ZERO-bit traffic generated per node is then

$$\tau_R = \frac{s_R * d_R * \lambda_R}{n}. \quad (2)$$

Now if there are  $s_U$  nodes which are sources of unusual events each generating  $\lambda_U$  packets per second, and if they travel on the average  $d_U$  hops to the destination, they will now generate an additional average traffic rate per node of

$$\tau_U = \frac{s_U * d_U * \lambda_U}{n}. \quad (3)$$

Both  $d_R$  and  $d_U$  also depend on the loss probability and other

parameters such as the time-out.

The average incoming traffic rate per node is then

$$\tau = \tau_R + \tau_U \quad (4)$$

over the set of all  $n$  nodes in the network. Since  $d_R$  increases when RRR is used, it follows that *all* packets may experience greater delays per node when RRR is used.

However the preceding analysis is based on averages over the whole network, and RRR will in practice *reduce* the traffic experienced by nodes which carry ONE-bit traffic while increasing the traffic at nodes that are used by ZERO-bit packets.

### B. Travel delay for the RRR algorithm

Using the vectors  $H$  and  $P$  defined in Section V, we can compute the  $b_U$  parameter for the travel delay (in number of hops) for the ONE-bit packets as:

$$b_U = \sum_j p_j^U h_j, \quad (5)$$

where  $p_j^U$  is the probability that a node selects the  $j$ -th next neighbor as next hop to forward a ONE-bit packet. Similarly,

$$\begin{aligned} c_U &= \sum_j p_j^U h_j^2 - (b_U)^2 \\ &= \sum_j p_j^U h_j^2 - \left(\sum_j p_j^U h_j\right)^2 \end{aligned} \quad (6)$$

and the diffusion model will yield  $d_U$  the average number of hops for ONE-bit packets to reach a destination which is at distance  $D$  hops.

In order to consider the total travel delay, we also have to compute the average delay  $Q$  (queueing plus transmission) through each hop for each type of traffic which we will approximate using a M/M/1 queue yielding:

$$\overline{Q_U} = \frac{\mu^{-1}}{1 - \rho_U}, \quad (7)$$

where  $\rho_U = \tau_U / \mu$  and  $1/\mu$  is the link transmission delay per packet plus any processing delay through a node.

The total source-destination travel delay of the  $U$  packets is then

$$E[T_U] = \left[ \frac{2D\mu^{-1}}{1 - \frac{\tau_U}{\mu}} \right] \frac{1 + \frac{f+r}{m} + \frac{f}{r}}{|b_U| + \sqrt{b_U^2 + 2c_U(f+r)}}. \quad (8)$$

When forwarding ZERO-bit packets, nodes have a different set of probabilities for selecting neighbors as discussed in Section V. Using a similar approach as in the previous paragraph we will get:

$$b_R = \sum_j p_j^R h_j \quad (9)$$

and

$$\begin{aligned} c_R &= \sum_j p_j^R h_j^2 - (b_R)^2 \\ &= \sum_j p_j^R h_j^2 - \left(\sum_j p_j^R h_j\right)^2 \end{aligned} \quad (10)$$

so that the total average travel delay for ZERO-bit packets which originate at distance  $D$  hops from their destination is:

$$E[T_R] = \left[ \frac{2D\mu^{-1}}{1 - \frac{\tau_R}{\mu}} \right] \frac{1 + \frac{f+r}{m} + \frac{f}{r}}{|b_R| + \sqrt{b_R^2 + 2c_R(f+r)}}. \quad (11)$$

Note that we have assumed that  $\tau$  is the same for all nodes, which is a worst case assumption for the nodes carrying ONE-bit packets.

### C. Merit of the RRR algorithm

In order to evaluate the effectiveness of the RRR algorithm, we evaluate the preferential treatment of ONE-bit packets with respect to ZERO-bit packets, when both types of packets originate from nodes situated at the same distance to the destination.

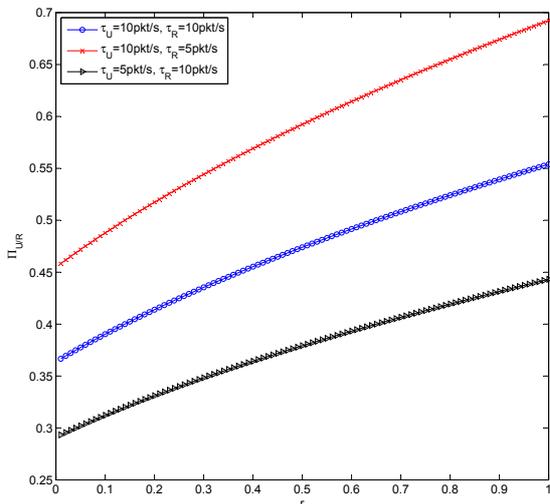


Fig. 5. Ratio  $\Pi_{U/R}$  with  $\mu=30\text{pkt/s}$  showing merit of the RRR algorithm.

Figure 5 shows the ratio  $\Pi_{U/R} = E[T_U]/E[T_R]$  which is:

$$\Pi_{U/R} = \frac{1 - \frac{\tau_R}{\mu} |b_R| + \sqrt{b_R^2 + 2c_R(f+r)}}{1 - \frac{\tau_U}{\mu} |b_U| + \sqrt{b_U^2 + 2c_U(f+r)}}, \quad (12)$$

where it is assumed that with RRR distinct nodes carry different types of packets are distinct.

## VI. SIMULATIONS OF THE RRR ALGORITHM

We have conducted simulations using the *ns-2* tool [20] to evaluate the RRR algorithm. The simulation parameters have been selected according to similar studies in SNs [21], [6], [8]. We focus on a WSN which collects and reports routine data to the sink constantly. There are a total of 100 sensors, each of which is uniformly distributed in an area whose size

is  $200 \times 200$  meters. Their communication range is  $40m$ . A single sink is located at the center position  $(100, 100)$  of the area. Any of the sensors has a probability  $p$  to be the source of routine data and generates data independently of the other nodes. We have also introduced a probability of packet loss at each node given by the parameter  $f$ .

At the beginning, only routine data is generated and after 1000 seconds four nodes located at  $(50, 50)$ ,  $(50, 150)$ ,  $(150, 50)$  and  $(150, 150)$  start generating the unusual traffic. We set  $p = 0.5$  and  $f = 0$ . The data rates of routine data and unusual events are  $\lambda_R = 1\text{pkt/s}$  and  $\lambda_U = 5\text{pkt/s}$ , respectively. Figure 6 shows how the travel delay of packets changes with time. The travel delay of packets emanating from four routine data sources located at  $(30, 100)$ ,  $(170, 100)$  and  $(100, 30)$ ,  $(100, 170)$ , and hence located at the same distance to the sink as the unusual events, is also plotted for comparison.

When the simulation starts, each node is forwarding its packets to a set of neighbors with equal probability as if the network had in the past experienced some unusual events. After some time, the nodes switch back to shortest path geographic routing for the routine data packets since they have learned that no unusual event data packets are arriving to them, and the travel delay of routine data drops significantly. The travel delay of the routine packets from the four reference source nodes is lower than the overall average since the reference nodes are closer to the sink than other nodes on average. When the unusual events occur at time 1100s, RRR starts operating. The ONE-bit packets are routed along the shortest paths while the ZERO-bit packets, which are still generated at the same rate, use the randomized routing scheme and hence suffer higher delay.

Figure 7 shows the results of the same experiment with another set of reference points of routine data sources at  $(30, 30)$ ,  $(30, 170)$ ,  $(170, 30)$ ,  $(170, 170)$ . Again, it shows that our proposed RRR algorithm yields the results we expect both in terms of adaptation and in providing far better travel delays to the ONE-bit packets. Since in this case the reference points of the routine data sources are located further away from the sink than the ‘‘average node’’, their travel delays are higher than the average delay.

## VII. CONCLUSIONS

In this paper we have proposed a simple randomized algorithm that is designed to provide better quality of service to the packets carrying the novel or unusual data in SNs, and also achieving overall good performance by distributing the secondary or routine traffic across a wider area of the network.

We have suggested a scheme to detect the occurrence of unusual events by observing a significant change on the content of packets being conveyed, and have proposed the adaptive Randomized Re-Routing (RRR) algorithm to provide differentiated quality of service to routine and unusual events.

We have evaluated the RRR algorithm using a theoretical model based on diffusion approximations, and also presented some simulation results. Both the analysis and simulation

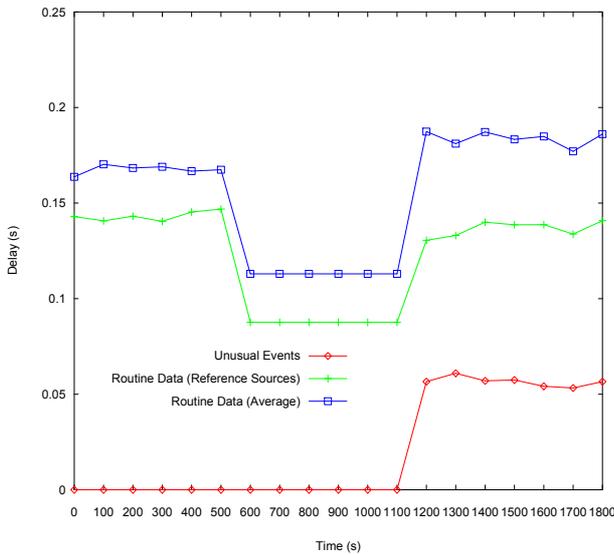


Fig. 6. Travel delay in three phases with  $f=0$ ,  $\lambda_U=5\text{pkt/s}$ ,  $\lambda_R=1\text{pkt/s}$ , reference routine data sources at (30, 100), (170, 100), (100, 30), and (100, 170).

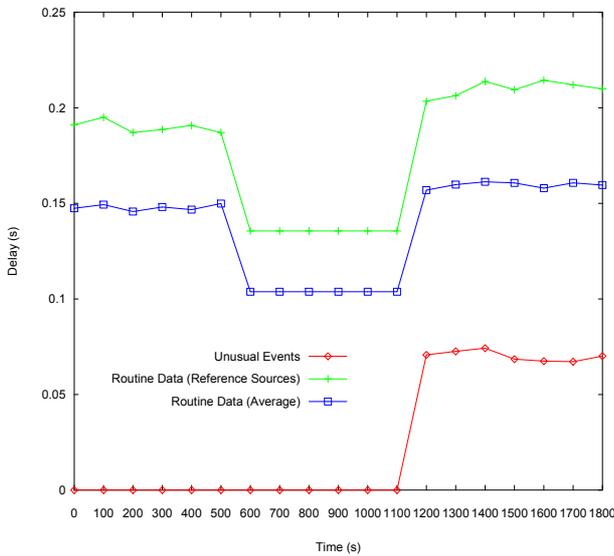


Fig. 7. Travel delay in three phases with  $f=0$ ,  $\lambda_U=5\text{pkt/s}$ ,  $\lambda_R=1\text{pkt/s}$ , reference routine data sources at (30, 30), (30, 170), (170, 30), (170, 170).

demonstrate that RRR can achieve significant QoS improvements for high priority traffic, while offering acceptable QoS levels to secondary traffic streams. When unusual events occur in the network, RRR has the added advantage of distributing the “routine” traffic streams randomly over secondary paths in the network so as to reduce congestion on the more heavily used shortest paths.

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