Shadowing and Mobility Effect on Proactive and Reactive Routing Protocol in MANET

ISTIKMAL\textsuperscript{1}, SUGONDO HADYIYOSO\textsuperscript{2}, INDRARINI DYAH IRAWATI\textsuperscript{2} ARIF INDRA IRAWAN\textsuperscript{1}

\textsuperscript{1}Fakultas Teknik Elektro, Universitas Telkom, Bandung  
\textsuperscript{2}Fakultas Ilmu Terapan, Universitas Telkom, Bandung  
Email: istikmal@telkomuniversity.ac.id

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ABSTRACT

In this paper, we compared the proactive routing protocols Optimized Link State Routing (OLSR) and the reactive routing protocols Ad-hoc on-demand distance vector (AODV) and Dynamic Source Routing (DSR) in shadowing propagation and mobility model on mobile adhoc network (MANET). There are two test scenarios, such as the impact of user movement velocity and variations in user pause time with Random Way Point (RWP) mobility. In user velocity testing, the greater the speed, the throughput performance will decrease, while the delay and Normalized Routing Load (NRL) will increase. The pause time test describes the dynamics of changing network topology. The longer the pause time, the longer the topology changes, as a result, the throughput parameter increases while the end-to-end delay and NRL decrease. The proactive OLSR shows the best performance compared to other protocols based on throughput parameters, while the reactive AODV outperforms for end-to-end delay and NRL parameters.

Keywords: routing protocol, shadowing, Mobility, MANET
1. INTRODUCTION

The routing protocol on the mobile Ad Hoc network responsibly to establish the communication path between the sender and the receiver. User movement and obstructions between the sender and receiver can cause the connection to deteriorate. The obstructions can cause a shadow effect. The shadow effect is explained as the effect of fluctuations in the received signal power due to the presence of obstructions between the receiver and the transmitter user or devices. The shadowing barriers catch up and prevent the sent signal, causing relatively sluggish signal changes compared to path loss model’s nominal values. Changes in connection quality degradation due to shadowing effects will increase the probability of packet errors and connection drops. With the communication line disconnected, the routing protocol must re-find the path.

In a real-world urban setting, researchers studied the impact of propagation model on routing algorithm performance on Ad Hoc network (Angeles et al., 2016)(Gutiérrez et al., 2016)(Istikmal et al., 2019)(Kuwahara et al., 2021). These studies are using end-to-end delay and delivery of packet rate to assess the routing performance. The impact of velocity speed to the protocol routing performance under the two way and the Nakagami-m of propagation loss model was investigated by (Aji Pratama et al., 2020), also modeled and analyzed of the shadow fading (Zhang et al., 2019). Using CORNER propagation model, the authors in (Guo et al., 2013)(Mukunthan et al., 2017), show the importance of the effect propagation model in network design using routing protocols. Moreover, the author in (Bakhtin et al., 2019), offers a way for implementing 3D obstacles, and their impact on signal propagation and one of the routing protocols for mobile Ad Hoc networks was tested. In (Huang et al., 2018), the authors elaborate a graph theory-based optimization approach for reducing the influence of the obstructive effect on the propagation of safety-related information.

In wireless network planning, the shadowing characterization data are commonly utilized to make adjustments to the link budget path loss calculations to account for statistical coverage estimations. Multiple reflections can generate somewhat slow change in Line-of-sight (LOS) components for some channels, causing an effect similar to blockage (Goldsmith, 2005; Liu et al., 2016). In Mobile Ad Hoc Networks, the author in (Amjad et al., 2015) study emphasizes the relevance of network performance and the physical layer using two-ray ground and shadowing propagation model. However, previous research did not analyze the impact of user’s mobility based on pause time and velocity with reference shadowing distance on routing protocol performance.

Therefore, in this paper, we analyze the impact of user variation pause time and the mobility with random way point (RWP) in shadowing propagation model. The pause time in a mobility model is the time it takes for the user to return to the new destination after arriving at the first one. The shorter the pause period, the more dynamic the user’s movements are. As a result, network topology changes more frequently. The user pause time movement and shadowing will have great impact on routing protocol performance.

2. RESEARCH METHOD

In this section, we investigates the correlation between shadowing and user movement mobility based on pause time and random way point to the routing protocol performance. In this research, we used proactive and reactive routing protocol. The Optimized Link State Routing (OLSR) known as proactive routing protocols, while Dynamic Source Routing (DSR)
and Ad-hoc on-demand distance vector (AODV) known as reactive routing protocols (Nayyar, 2018). We compare the performance of these routing protocols in mobility scenario under shadowing propagation model. The separation distance between the transmitter and the receiver is commonly used to determine the received power attenuation. The power of received signal according to the time of measurement based on geographic location can be expressed as follows (Liu et al., 2016):

\[ P_{rx} = \frac{1}{W} \sum_{n=1}^{W} \sum_{i=0}^{L-1} |h(nt_{rep}, l\Delta t)|^2. \]  

(1)

Simple empirical path loss modeling in decibels can be used to predict power attenuation after acquiring the received power versus the separation distance between the transmitter and the receiver, can be calculated as (Liu et al., 2016):

\[ PL(d) = A + 10.n.\log_{10}(d) + X_{\sigma}. \]  

(2)

According to (Goldsmith, 2005; Amjad et al., 2015), the authors combine path loss in the shadowing propagation model. Path loss is modeled simply and then combined with log-normal shadowing. This combined model produces power between transmitter-receiver which is expressed in dB as follows (Amjad et al., 2015):

\[ \frac{P_r}{P_t} dB = 10 \log k - 10.\gamma.\log\frac{d}{d_0} - \varphi_{dB}. \]  

(3)

where \( \gamma \) and \( d_0 \) are path loss exponent and reference distance, respectively. Gaussian distribution random variable and a unitless are \( \varphi_{dB} \) and \( k \). Moreover, to simplify in network simulator environment, there are two aspects to the shadowing model. The path loss model, which also estimates the mean received power at distance \( d \), represented by \( P_r(d) \). As a reference, it employs a close-in distance of \( d \). \( P_r(d) \) is calculated as follows in relation to \( P_r(d_0) \).

\[ \frac{P_r(d_0)}{P_r(d)} = \left(\frac{d}{d_0}\right)^{\beta} \]  

(4)

The path loss exponent, \( \beta \), is normally established empirically by field measurement. As distance increases, greater numbers equate to more obstructions, resulting in a faster fall in average received power. Table 1 shows the value of path loss exponent \( \beta \) in typical environments and shadowing deviation \( \sigma_{dB} \), refer to (Srinivasa and Haenggi, 2009).

**Table 1. Path loss exponent and shadowing deviation**

| Typical        | Environments                | \( \beta \) | \( \sigma_{dB} \)       |
|----------------|-----------------------------|-------------|--------------------------|
| Outdoor        | Free Space area, LOS        | 2.0         | 4 until 12               |
|                | Shadowed in urban place     | 2.75        | until 5.0                |
| Indoor         | In building, LOS            | 1.6         | until 1.8                |
|                | Obstructed indoor           | 4.0         | until 6.0                |
|                | Office with hard partition  | 7.0         |                          |
|                | Office with soft partition  | 9.6         |                          |

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Finally, the shadowing model's second part reflects the change in received power at a given distance. It has a Gaussian distribution when measured in decibels since it is a log-normal random variable. The shadowing model as a whole is represented by

\[
\frac{P_r(d)}{P_t(d_0)}_{dB} = -10\beta \log \left( \frac{d}{d_0} \right) + X_{dB}
\]

where \( \sigma_{dB} \) is a standard deviation and \( X_{dB} \) is a Gaussian random variable with a zero mean. The shadowing deviation, commonly known as \( \sigma_{dB} \), is determined by measurement. Based on this shadowing propagation model, we investigated the performance of AODV, DSDV, and OLSR routing protocol in velocity and pause time scenarios. In this investigation, there are 30 nodes, where one sender will send data to one recipient through 28 other nodes which move randomly. Table 2 shows the simulation parameter in the network simulator.

| Mobility                  | Random Way Point |
|--------------------------|------------------|
| Number of Nodes          | 30               |
| Packet Size              | 1024 Bytes       |
| Sender/receiver node     | 1                |
| Velocity of nodes (m/s)  | 2, 4, 6, 8, 10   |
| Propagation Model        | Shadowing        |
| Routing Protocols        | AODV, DSR, and OLSR |
| Reference Distance       | 1 m, 50 m        |
| Pause Time (second)      | 0, 10, 20, 30, 40, 50 |
| Power Transmits Tx       | 0.28183815 W     |
| Frequency                | 914.e6 Hz        |
| High of the node (ht,hr) | 1,5 m            |

![Figure 1. Research Method and simulation scenario](image-url)
Figure 1 shows the research method and simulation scenario. We use throughput and end-to-end delays parameter to measure the network performance and normalized routing load (NRL) to see the routing protocol efficiency. The queuing, route discovery, latency, and buffering delays are assessed in milliseconds and included in the end-to-end delay performance analysis. We can calculate the throughput \( B \), NRL by (Amjad et al., 2015):

\[
B = \frac{\text{data packets successfully delivered}}{\text{observation time}}
\]

\[
\text{NRL} = \frac{\text{The total number of routing packets generated}}{\text{the number of data packets successfully delivered to destination}}
\]

In this simulation, the mobility of nodes using random waypoint, where the node freely to move to any direction. In the first scenario, we want to investigate the routing protocol performance against the shadowing and the increase of the node velocity. The impact of the pause time in dynamically change of topology and at the same time the node has shadowing effect on routing protocol performance is the purpose of the second scenario. The nodes choose a random destination anywhere in the network and begin moving towards it at a constant speed in this mobility model. After arriving at the destination, the node will wait for the amount of time indicated by the "pause time" argument. Once the pause time has passed, the node will restart moving in a new direction until the simulation is completed.

3. RESULTS AND DISCUSSION

In reactive routing protocol, when the source wants to transmit the data, if there is no valid route, the route request (RREQ) will be broadcast by the source to the all neighbors. When the RREQ is received at the destination node, it is replayed as a route reply (RREP) and forwarded to the source node over a unicast link. The source node begins transmitting the data after receiving RREP. Meanwhile, proactive routing protocol will maintain the valid path periodically, even though the source is not sending data.
Figure 2 dan 3 shows the throughput and end-to-end delay performance of AODV, DSR, and OLSR routing protocol in velocity scenario with reference distance in shadowing 1 m. The OLSR routing protocol outperforms in average throughput than AODV and DSR. However, AODV gain advantages with smallest end-to-end delay. The OLSR used the concept of multi-point relay (MPR) to improve network performance, while in AODV, it can reduce the rate of increase in end-to-end delay, but does not obtain optimal throughput.

Figure 4 shows the difference in efficiency between reactive and proactive routing protocols based on NRL values. Although, OLSR and DSR get better performance in throughput and end-to-end, they produce higher NRL values than AODV. The smaller the NRL value, the more efficient the routing protocol algorithm. The velocity and shadowing lead the routing protocol to face signal quality connection degradation, this is also reporting in (Istikmal et al., 2015), path break and higher bit error rate. Increasing the node velocity will decrease average throughput and increase NRL and average end-to-end delay. The throughput and end-to-end delay AODV, DSR, and OLSR routing protocol performance in pause time scenario with reference distance in shadowing 50 m is shown in Figure 5 and 6, respectively.
The shorter the pause time duration, the more dynamically the network topology will change. These will lead path break and makes routing protocol enter the route discovery process. At the same time, the shadowing influences the degradation of the signal quality connection. This will lead path break and makes routing protocol frequently enter the route discovery process. Fig. 7 shows the normalized routing load of AODV, DSDV, OLSR in pause time scenario with $d = 50$ m.
The simulation result shows that AODV outperform than DSR and OLSR in term of NRL performance. As a reactive routing protocol, the AODV generates routing packets when forming a new topology. This is the reason that AODV has the lowest NRL. The number of routing packets is much less than the number of data packets received at the destination. Based on these results, a new method of routing protocols is needed to be able to adapt to the shadowing environment, among others are the probability-based reliable routing (Sun et al., 2019)(Xu et al., 2020) and cross layer scheme (Istikmal et al., 2017)(Wu et al., 2021) between physical layer and routing protocol. The modification of these methods in routing protocol algorithm, can be considered to improve network performance.

4. CONCLUSION

We have successfully simulated Ad Hoc network topology and tested the performance of AODV, DSR, and OLSR routing protocols based on shadow propagation models. Routing protocol performance is measured by parameters of throughput, end to end delay, and NRL with random user movements. The speed change simulation shows that the higher the speed, the lower the throughput, the higher the end-to-end delay and the NRL. Based on experiments, users with a speed of 6m/s showed the best performance on the three routing protocols. In the scenario of topology changes that are simulated with changes in pause time, it shows that the longer the pause time, the higher the throughput and the lower end-to-end delay and NRL. The proactive OLSR has superior performance based on throughput parameters because this protocol is always adaptive to update the routing table and has multi-point relay strategy. However, OLSR has a large NRL due to the large number of routing packets sent for this purpose and results in long delays. The reactive protocol AODV has superior performance in end-to-end delay and NRL. Cross layer scheme and probability-based reliable routing can be considered as methods to improve routing protocol performance for future research.

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