

On Accurate Measurement of Link Quality in Mobile MIMO-OFDM Systems

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I. BACKGROUND

The wireless networks have experienced rapid development in recent years, which leads to the increase of traffic loads and users requirements. The wide expansion of wireless services has brought serious challenges to the issues of spectrum efficiency and required data rate. Recently, WLANs (Wireless Local Area Networks) based on 802.11n have enjoyed tremendous growth due to the ever-increased demands of high-bandwidth applications. Moreover, with the increasing popularity of smartphones, the growth of mobile 802.11n is expected to continue unabated [Balasubramanian et al., 2010]. On the other hand, the power consumption of wireless communications is becoming increasingly vital to a more efficient and longer lasting mobile battery in mobile wireless systems. The continued success of mobile 802.11n depends on their ability to efficiently configure different PHY/MAC enhancements based on MIMO-OFDM technology, and here comes the question of how to get energy-efficient trade-off between reliability and data rate in mobile MIMO-OFDM systems?

In order to solve this problem, some approaches on energy-efficient rate adaption have been carried out based on simulations [Kim and Daneshrad, 2010] [Eraslan and Daneshrad, 2012] or experiments [Peng et al., 2011] [Li et al., 2012]. Although these works present different solutions such as convex optimization, packet delivery probing or channel state prediction, they are all based on the information of channel state or link quality. So a basic consideration is the **accurate channel state estimation and link quality measurement with low overhead**. This is challenging in that multi-configuration in mobile 802.11n not only requires far more samples to acquire sufficient information for all possible channel settings, but also introduces significant complications in channel modeling. Furthermore, channels are more vulnerable to environmental variability and terminal mobility in mobile 802.11n. Therefore, accurate channel measurement and prediction is becoming increasingly important in mobile 802.11n networks.

II. MOTIVATIONS

The 802.11n standard incorporates PHY/MAC enhancements to achieve higher throughput and wider coverage. In the PHY layer, 802.11n networks adopts MIMO technology to achieve spatial multiplexing and diversity. 802.11n utilizes channel bonding technology, with which two adjacent 20MHz channels are united to a new 40MHz one, to realize higher data rates. In the MAC layer, 802.11n also employs Short Guard Interval (SGI) and frame aggregation to reduce overhead and improve efficiency. All these PHY and MAC enhancements not only play an effective impact on the performance of higher layers, but also make link quality measurement and analysis more complicated. For above configurations of 802.11n, the PDR-RSS model shows different characteristics.

A. Time-varying and Location Differences

Static EWMA is widely used in PDR measure [Atheros, 2012] [Fietkau, 2005] [Wong et al., 2008], whose update time interval is set fixed to 50ms or 100ms. However, the temporary and spatial diversity of PDR would reduce the measurement accuracy of EWMA, especially when operating at high data rates. The following experiments show that PDR is vulnerable to spatial and temporal diversity in mobile 802.11n networks. RSS and PDR measurements were conducted along different routes by Atheros's 2T2R 802.11n module AR9382 at 5GHz band. Fig. 1a illustrates that both RSS and PDR encounter with sudden decline in short time scale. The Cumulative Distribution Functions (CDF) of RSS measures with the same 802.11n cards are given in Fig. 1b, which illustrates the spatial diversity feature of network status. Fig. 1c shows a specific measurement example for data rate of 78Mbps measured by the same way in Fig. 1a, where EWMA will overestimate 20% of PDR when there is a sudden decline. Thus, it is expected to develop dynamic PDR measurement method to increase the accuracy in mobile environments.

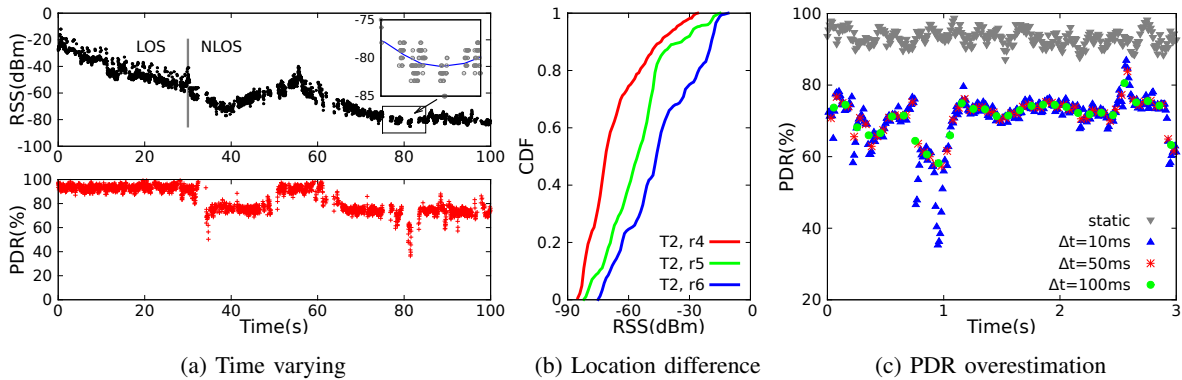


Fig. 1. Characteristics of PDR and RSS in mobile 802.11n, composed of both of LOS and NLOS scenarios.

B. Transition Windows

Since 802.11n standard incorporates several enhancements including channel bonding, spatial multiplexing, frame aggregation and SGI, it is a basic problem that how to switch between different operating configurations for better suiting the radio environments. The complexity stems from the large transition windows of PDR-RSS model in 802.11n [Halperin et al., 2010], as shown in Fig. 2a. This situation will be worse when PDR is overestimated by static measurement approach in fast changing channels.

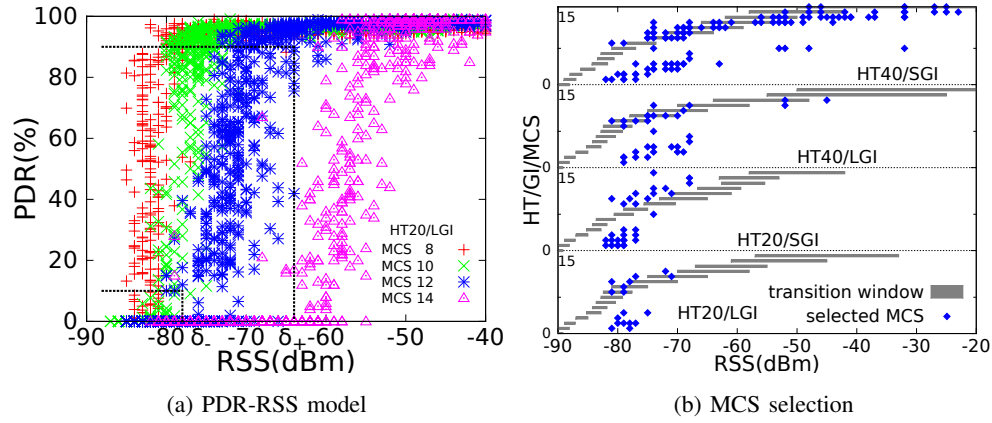


Fig. 2. (a) EWMA with update periods of 50/100ms will overestimate 20% of PDR when there is a sudden decline; (b) Large parts of selected HT/GI/MCS fall within the transition window, especially for high data rates.

Fig. 2b gives an example of rate selection results, which separates the HT/GI/MCS selection and RSS into three regions. The gray lines are the transition windows for each individual HT/GI/MCS selection, and locating just the right of gray line will efficiently improve achieved throughput with high reliability. For every four parts separated by HT/GI options, the selected MCS falling on the right of transition region means it can get high PDR under this setting. There are about 34% falling into the transition window, and even 8% are worse than locating on the left of the transition window. However, the good news is that the transition windows exhibit diversity distribution for different MCS selection and channel assignment from Fig. 2b. This indicates that there exists certain configuration(s) for current RSS that can ensure PDR being out of transition window. It provides a hint that we can configure different PHY/MAC enhancements at the right of gray line by jointly utilizing the real-time PDR and RSS, which improves the network throughput with high reliability.

The general framework for current channel measurement and prediction methods based on PDR-RSS model is shown in Fig. 3. The main features in the static framework are as follows: (1) static EWMA

measure for PDR; (2) static data set and PDR-RSS model; (3) single measurement metric (PDR or RSS) input. It will lead to new problems when the static PDR-RSS framework is applied in mobile 802.11n networks, which will be investigated in the following.

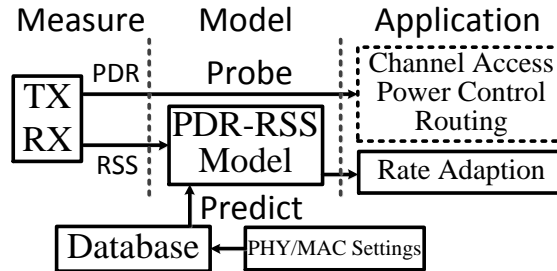


Fig. 3. General static PDR-RSS modeling framework.

To sum up, 802.11n PHY/MAC enhancements make the PDR measurement and modeling more complicated, and mobile wireless connections are highly changing. These factors significantly reduce the prediction accuracy and rate selection efficiency. The key is to design an on-line PDR-RSS model to address the following issues: (1) high-accuracy PDR measurement; (2) dynamic PDR-RSS model update; (3) efficient configuration(s) output. That is:

- 1) *How to accurately measure PDR with low overhead in fast changing channels?*
- 2) *How to characterize the relationship among PDR, RSS and 802.11n multi-configuration?*
- 3) *How to update the characterized relationship on-line and derive the set of configurations with certain reliable performance guaranteeing at current PDR and RSS?*

III. LITERATURE REVIEW

A. Related Works

1) *Practical Packet Delivery Models:* There are numerous work on practical delivery and interference modeling. Some early researches paid attention to offline models in static wireless networks [Kolar et al., 2011] [Reis et al., 2006], and it is widely used in upper layer applications such capacity analysis [Kashyap et al., 2007] and rate control [Chen et al., 2012] [Judd et al., 2008]. The authors in [Lenders and Martonosi, 2009] proposed repeatable measurement in mobile wireless networks that aimed at combining real experiments and simulations. The Sybot in [Kim et al., 2010] also conducts mobile spectrum survey which only makes RSS measurement but ignore the link level quality. These works are all based on traditional 802.11a/b/g that can not be directly applied in 802.11n networks. A number of

studies have investigated the experimental features of 802.11n networks recently [Halperin et al., 2010] [Rayanchu et al., 2011]. The authors in [Halperin et al., 2010] provided accurate delivery prediction for MIMO-OFDM of 802.11n based on CSI. But the channel estimation of CSI [Crepaldi et al., 2012] requires too much PHY/MAC operations which makes it more complicated for on-line measurement and modeling.

2) *Rate Adaption*: There are extensively large number of approaches on 802.11n rate control based on simulations or experiments [Kim et al., 2009] [Pefkianakis et al., 2010] [Zhang et al., 2008]. Some works have been deployed default on Linux platforms, for instance Minstrel [Fietkau, 2005] for `mac80211` and Atheros for `ath9k` [Wong et al., 2008]. However, these works are designed for static 802.11 networks, which utilize fixed EWMA calculations to process PDR and spend look around frames to detect available data rates. Some approaches were proposed on rate adaption in mobile environments, but most of them are concentrated on RSS measurement [Chen et al., 2012] [Judd et al., 2008]. Some upper layer applications such as intrusion detection [Boppana and Su, 2011] and congestion control [Floyd et al., 2000] employ on-line PDR measurement methods, but the above proposals do not focus on PDR-RSS modeling and rate selection related issues.

Recently, many researchers are working on energy-efficient issues in MIMO-OFDM wireless networks including WiMAX, LTE, WLANs. Some approaches are carried out based on convex optimization [Kim and Daneshrad, 2010] or CSI probing [Eraslan and Daneshrad, 2012] with mathematical analysis and simulations. Others are designed for 3G [Peng et al., 2011] or 802.11n [Li et al., 2012] [Zhang and Shin, 2011] networks by practical implementation and evaluation. In addition to utilizing physical-layer and packet-layer metrics, some studies aim at automatically adaption algorithm based on traffic patterns of various upper layer applications [Han et al., 2012] [Jang et al., 2011] or users' requirements [Zhuang et al., 2010] [Schulman et al., 2010]. Although these works present different solutions, they are all based on the information of channel state and link quality. Obviously, accurate information of channel state and link quality has significant influence on performance of rate adaption algorithms.

B. Statement of Significance

The measurement-based Packet Delivery Ratio and Received Signal Strength (PDR-RSS) model is widely used for performance analysis in static wireless networks [Reis et al., 2006]. It has been successfully exploited in 802.11a/b/g for upper layer applications such as capacity analysis [Kashyap et al., 2007] and spectrum allocation [Rayanchu et al., 2011]. However, PDR-RSS model exhibits a large transition window in wireless channels due given the fact that the frequency selectivity of the wideband 802.11

channel is not captured by RSS. Halperin *et al.* [Halperin et al., 2010] proposed predictable model to determine the subcarrier Signal to Noise Ratio (SNR) using Channel State Information (CSI) and then aggregate into a global metric called eSNR, which provided better characterization for link quality. But CSI-based measurement in 802.11n can dramatically increase the complexity of channel estimation and modeling. Furthermore, commonly used 802.11n wireless devices only provide CSI reports for unicast packets, which clearly affect the efficiency of collecting CSI matrices. Hence, the tradeoff exists between measurement overhead and accuracy which is worthy to be explored, especially in mobile environments.

Prior works on the PDR-RSS modeling are mostly static [Kashyap et al., 2007] [Kolar et al., 2011] [Reis et al., 2006]. Moreover, a single measurement metric, either packet-level (PDR) or physical-level (RSS), is utilized by upper layer applications [Judd et al., 2008] [Zhang et al., 2008]. The PDR-RSS model can overcome the channel quality capturing problem if PDR and RSS are jointly considered. This is further supported by exploiting the multi-configuration properties in 802.11n, in which the transition window exhibits diversity distribution for different configurations from our extensive experiments. It indicates that there exists certain configuration(s) for current RSS that can ensure PDR being out of transition windows. This key observation motivates us developing an on-line PDR-RSS modeling framework, which utilizes real-time PDR and RSS to update PDR-RSS model dynamically and configure PHY/MAC settings in mobile 802.11n.

IV. RESEARCH METHODOLOGY

Fig. 4a presents the on-line PDR-RSS modeling framework, which is composed of three components:

- **Database:** contains the raw data of PDR and RSS along with different 802.11n PHY/MAC settings.
- **PDR-RSS model:** a set of data pair for transition windows' lower and upper bound.
- **HT-GI-MCS index:** the configuration selection sequence.

Fig. 4b shows the software architecture, which is composed of network and device layer components:

- **Network layer:** conducts DSWA calculations to determine averaging intervals and sliding factor, and makes update to get HT-GI-MCS indexes.
- **Device layer:** makes PDR computation and RSS averaging driven by TX/RX events respectively, and selects rate indexes according to network layer results as PDR or RSS lower than certain threshold.

The construction of the online framework is consist of the following three steps.

- 1) First, the database is initialized through empirical experiments.
- 2) Second, it will be updated to the PDR-RSS model under different settings in realtime operating.

- 3) Finally, the HT/GI/MCS selection sequence can be generated according to the online PDR-RSS model and current status, which can provide optional HT/GI/MCS sorted in order.

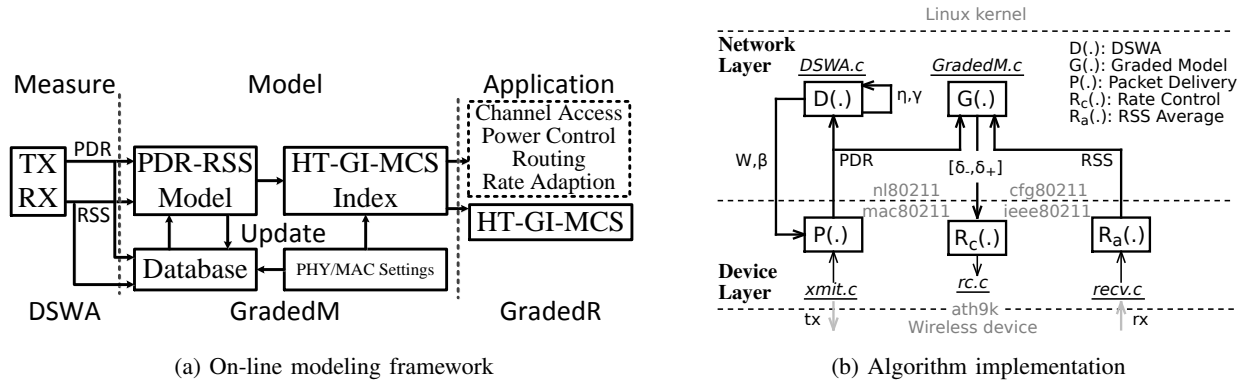


Fig. 4. On-line modeling framework and implementation

Compared to the general static PDR-RSS modeling framework in Fig. 3, the on-line framework has the following distinctive features.

- 1) First, it has two inputs into both PDR-RSS model and database.
- 2) Second, both PDR-RSS model and database are updated online.
- 3) Third, it exploits both real-time PDR and RSS, along with the diversity property to derive the set of configuration(s) with certain reliable performance guaranteeing.

Through the distinctive features, the online PDR-RSS modeling framework can provide a systematic solution for the channel quality capturing problem in static PDR-RSS models. In the following, the report gives a thorough design process for the on-line framework to demonstrate how to address the three critical issues: high-accuracy PDR measure, dynamic PDR-RSS model update and efficient configuration output, in the on-line modeling framework.

A. Packet Delivery Measurement

For the statistical PDR-RSS model based on realistic measurement in mobile wireless networks, there is a trade-off between measurement accuracy and overhead. The measurement period should be set short enough and responding rapidly to changing network status, or be long cycle to reduce overhead when link quality is steady and reliable. At the same time, both RSS and PDR have time varying and location difference features in mobile 802.11n. And the diversification in data rates and packet size will significantly complicate the programming process.

In the traditional rate control algorithms of Atheros's Linux wireless drivers, both `Madwifi` for 802.11a/b/g and `ath9k` [Atheros, 2012] for 802.11n, EWMA is used to process PDR of each configuration, as shown in Fig. 5a. EWMA has some deficiencies when applied in mobile 802.11n.

- The weighting coefficient α is set fixed to 0.125 or 0.25 in practical implementation [Atheros, 2012] [Fietkau, 2005], and it is hard to respond promptly to link quality changes [NIST/SEMATECH, 2012].
- The update cycle is also fixed to 50ms or 100ms, which can not achieve effective control on accuracy and overhead. This will lead to PDR overestimation at high data rates as shown in Fig. 1c.

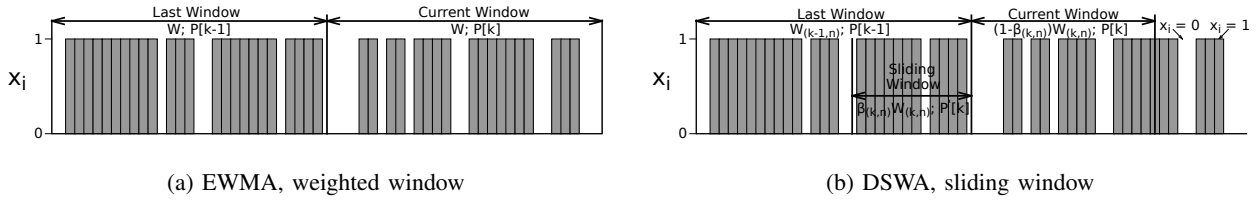


Fig. 5. Averaging window length in PDR measurement

The DSWA approach can be employed to resolve this dilemma. Unlike EWMA, the window length W is event driven and unrelated to data rates, and DSWA takes advantage of sliding window averaging to emphasize on the most recent status. It can be inferred that the measurement errors are closely associated with channel state changes, which have different features in static or mobile wireless networks. Assuming the same average intervals, it will increase the measurement errors when channel state encounters with sudden changes in short time scale.

B. Channel State Estimation

C. Packet Delivery Modeling

The on-line PDR-RSS modeling is to update the PDR-RSS model and database, and generate the HT/GI/MCS selection sequence according to current network conditions. Obviously, the reliability can be identified by the distance between transition windows' upper bound and current RSS. Then the database is sorted in order by this distance, and the HT/GI/MCS selection sequence is generated accordingly.

D. Rate Adaption

DSWA can be adopted to get accurate PDR measurement with low overhead, then the suitable configuration can be chosen according to HT/GI/MCS index. The PDR threshold in GradedM.c is set to $\{P_{thrl}, P_{thrh}\} = \{10\%, 90\%\}$. When the selected MCS is far away the right bound of its transition window,

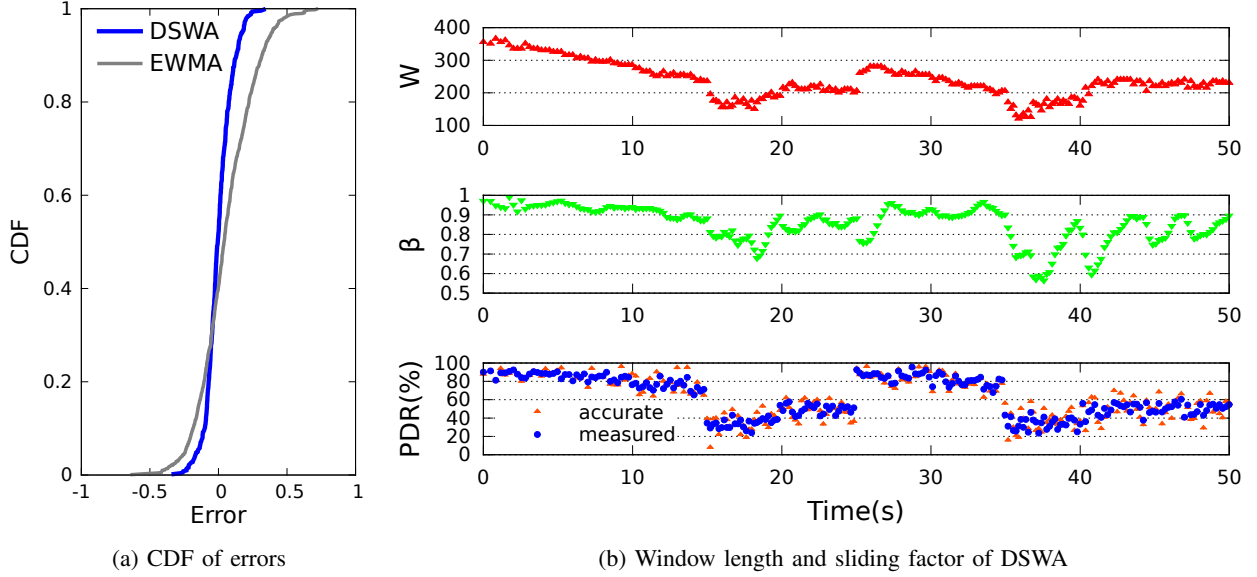


Fig. 6. Measurement results of EWMA and DSWA in mobile scenarios.

GradedR will choose a new configuration to acquire a higher data rate. On the contrast, it will reduce the data rate when current PDR falls into the transition window. Given the characterization results of PDR-RSS model, SGI is only selected as the final rate adaptation step when the link quality is still poor running at the highest rates of LGI.

V. OUTCOMES AND VALUES

A. Measurement

1) *Accuracy*: For the PDR measurement in realistic networks, EWMA can hardly get sufficient measurement accuracy compared to DSWA. In mobile wireless networks, the propagation environments are complex and communication terminals are on the move particularly, which means the RSS and interference are changing during PDR measurement. This will make the packets received probability changes in short time scale. In this case, it can be characterized by a Generalized Bernoulli process that the probability of $x_i = 1$ is different for all values of i . Fig. 6a illustrates the CDF of measurement errors for EWMA and DSWA when applied in mobile scenarios. The measurement errors are within ± 0.008 for DSWA, and change from -0.019 to 0.032 for EWMA. The errors of EWMA show that it tends to overestimate the actual PDR, which can also be seen from Fig. 6a that the overall CDF curve of EWMA errors shift to the right of line $Error = 0$. Compared with the traditional EWMA method, DSWA can improve the overall measurement accuracy of 89% higher in mobile scenarios.

2) *Overhead*: In addition to meet the accuracy requirements, it also deserves attention to reduce measurement overhead, since more sample packets will lower the throughput achieved. The sampling intervals of DSWA are weighted average of last n results so that it can reduce mutations caused by noise and respond quickly to real changes of PDR values. Moreover, the averaging intervals of DSWA are associated with PDR changes to allow a more timely response to sustained decreasing in link quality, and make less frequent samples as network conditions are in steady continuously. Fig. 6b shows an example of DSWA for measuring PDR adaptive to different network conditions. The packet delivery has a sudden decrease at the time of about 15s, and both average interval W and sliding factor β drop accordingly. When PDR increases and getting stable from 40s to 50s, W changes from 100 to 200 which will reduce measurement overhead significantly. The average window length of EWMA are approximately constant for certain rate that $W = 20$ for 6.5Mbps and $W = 500$ for 300Mbps. EWMA can not respond timely when $W = 500$ and result in unnecessary errors, especially when there is sudden PDR decline.

B. Performance

1) Reliability:

2) *Throughput*: As is illustrated in Fig. ?? of 2x3 MIMO, the throughput is 5-20Mbps higher before the time of 8s, and it will be even more than 30Mbps higher for 3x3 MIMO before 15s in Fig. ?. The reason is that GradedR updates the sensitivity table realtime and chooses the most suitable rate indexes for current conditions rather than randomly select a lower rate. The achieved PDR also has significant impact on throughput. For all the experiments along **r5**, at least 91% of PDR values are greater than 90% for GradedR, but more than 63% are lower than 90% for Minstrel. GradedR's rate selection is more smooth and stable, which can avoid concussion of network status, when the network conditions are good enough. And another aspect which will obviously affect the throughput is the measurement overhead. Since Minstrel spends 10% percentage of frames, doing "look around" i.e. randomly trying other rates, to gather statistics, the rate being looked around can hardly meet the current situation and it will increase unnecessary load of wireless link. For GradedR, it first adopts DSWA measurement method to get accurate and efficient packet delivery prediction adaptive to network conditions, and then sorts the GradedT to get the parameter configuration according to current PDR and RSS. These two procedures not only reduce the measurement overhead but also improve rate selection efficiency.

3) Energy-efficiency:

REFERENCES

[Atheros, 2012] Atheros, Q. (2012). Atheros linux wireless drivers. <http://linuxwireless.org/en/users/Drivers/ath9k>.

- [Balasubramanian et al., 2010] Balasubramanian, A., Mahajan, R., and Venkataramani, A. (2010). Augmenting mobile 3G using WiFi. *ACM MOBISYS '10*.
- [Boppana and Su, 2011] Boppana, R. and Su, X. (2011). On the effectiveness of monitoring for intrusion detection in mobile Ad Hoc networks. *IEEE Transactions on Mobile Computing*, 10(8):1162–1174.
- [Chen et al., 2012] Chen, X., Gangwal, P., and Qiao, D. (2012). RAM: Rate adaptation in mobile environments. *IEEE Transactions on Mobile Computing*, 11(3):464–477.
- [Crepaldi et al., 2012] Crepaldi, R., Lee, J., Etkin, R., Lee, S., and Kravets, R. (2012). CSI-SF: Estimating wireless channel state using CSI sampling & fusion. *IEEE INFOCOM 2012*.
- [Eraslan and Daneshrad, 2012] Eraslan, E. and Daneshrad, B. (2012). Practical energy efficient link adaptation for mimo-ofdm systems. In *IEEE WCNC 2012*, pages 480–485.
- [Fietkau, 2005] Fietkau, F. (2005). Minstrel rate control algorithm for mac80211. http://madwifi-project.org/browser/madwifi/trunk/ath_rate/minstrel.
- [Floyd et al., 2000] Floyd, S., Handley, M., Padhye, J., and Widmer, J. (2000). Equation-based congestion control for unicast applications. *ACM SIGCOMM '00*.
- [Halperin et al., 2010] Halperin, D., Hu, W., Sheth, A., and Wetherall, D. (2010). Predictable 802.11 packet delivery from wireless channel measurements. *ACM SIGCOMM '10*.
- [Han et al., 2012] Han, H., Liu, Y., Shen, G., Zhang, Y., and Li, Q. (2012). Dozyap: power-efficient wi-fi tethering. *MobiSys '12*, pages 421–434, New York, NY, USA. ACM.
- [Jang et al., 2011] Jang, K.-Y., Hao, S., Sheth, A., and Govindan, R. (2011). Snooze: energy management in 802.11n w lans. In *CoNEXT '11*, pages 12:1–12:12, New York, NY, USA. ACM.
- [Judd et al., 2008] Judd, G., Wang, X., and Steenkiste, P. (2008). Efficient channel-aware rate adaptation in dynamic environments. *ACM MOBISYS '08*.
- [Kashyap et al., 2007] Kashyap, A., Ganguly, S., and Das, S. R. (2007). A measurement-based approach to modeling link capacity in 802.11-based wireless networks. *ACM MOBICOM '07*.
- [Kim and Daneshrad, 2010] Kim, H. S. and Daneshrad, B. (2010). Energy-constrained link adaptation for mimo ofdm wireless communication systems. *IEEE Transactions on Wireless Communications*, 9(9):2820–2832.
- [Kim et al., 2010] Kim, K., Min, A. W., and Shin, K. G. (2010). Sybot: an adaptive and mobile spectrum survey system for wifi networks. *ACM MOBICOM '10*.
- [Kim et al., 2009] Kim, W., Khan, O., Truong, K., Choi, S., Grant, R., Wright, H., Mandke, K., Daniels, R., Heath, R., and Nettles, S. (2009). An experimental evaluation of rate adaptation for multi-antenna systems. *IEEE INFOCOM 2009*.
- [Kolar et al., 2011] Kolar, V., Razak, S., Mähönen, P., and Abu-Ghazaleh, N. B. (2011). Link quality analysis and measurement in wireless mesh networks. *Ad Hoc Netw.*, 9(8):1430–1447.
- [Lenders and Martonosi, 2009] Lenders, V. and Martonosi, M. (2009). Repeatable and realistic experimentation in mobile wireless networks. *IEEE Transactions on Mobile Computing*, 8:1718–1728.
- [Li et al., 2012] Li, C.-Y., Peng, C., Lu, S., and Wang, X. (2012). Energy-based rate adaptation for 802.11n. *Mobicom '12*, pages 341–352, New York, NY, USA. ACM.
- [NIST/SEMATECH, 2012] NIST/SEMATECH (2012). e-handbook of statistical methods. <http://www.itl.nist.gov/div898/handbook/>.
- [Pefkianakis et al., 2010] Pefkianakis, I., Hu, Y., Wong, S., Yang, H., and Lu, S. (2010). MIMO rate adaptation in 802.11n wireless networks. *ACM MOBICOM '10*.

- [Peng et al., 2011] Peng, C., Lee, S.-B., Lu, S., Luo, H., and Li, H. (2011). Traffic-driven power saving in operational 3g cellular networks. *MobiCom '11*, pages 121–132, New York, NY, USA. ACM.
- [Rayanchu et al., 2011] Rayanchu, S. K., Shrivastava, V., Banerjee, S., and Chandra, R. (2011). FLUID: improving throughputs in enterprise wireless lans through flexible channelization. *ACM MOBICOM'11*.
- [Reis et al., 2006] Reis, C., Mahajan, R., Rodrig, M., Wetherall, D., and Zahorjan, J. (2006). Measurement-based models of delivery and interference in static wireless networks. *ACM SIGCOMM '06*.
- [Schulman et al., 2010] Schulman, A., Navda, V., Ramjee, R., Spring, N., Deshpande, P., Grunewald, C., Jain, K., and Padmanabhan, V. N. (2010). Bartendr: a practical approach to energy-aware cellular data scheduling. *MobiCom '10*, pages 85–96, New York, NY, USA. ACM.
- [Wong et al., 2008] Wong, M., Gilbert, J., and Barratt, C. (2008). Wireless LAN using RSSI and BER parameters for transmission rate adaptation. US Patent 7,369,510.
- [Zhang et al., 2008] Zhang, J., Tan, K., Zhao, J., Wu, H., and Zhang, Y. (2008). A practical SNR-guided rate adaptation. *IEEE INFOCOM 2008*.
- [Zhang and Shin, 2011] Zhang, X. and Shin, K. G. (2011). E-mili: energy-minimizing idle listening in wireless networks. *MobiCom '11*, pages 205–216, New York, NY, USA. ACM.
- [Zhuang et al., 2010] Zhuang, Z., Kim, K.-H., and Singh, J. P. (2010). Improving energy efficiency of location sensing on smartphones. *MobiSys '10*, pages 315–330, New York, NY, USA. ACM.