

On-line and Dynamic Estimation of Rician Fading Channels in GSM-R Networks

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1 Introduction

- Background
- Problem Formulation

2 On-line Estimation

- Measurement Framework
- Local Power Estimation

3 Performance Evaluation

- Implementation
- Evaluation

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2 On-line Estimation

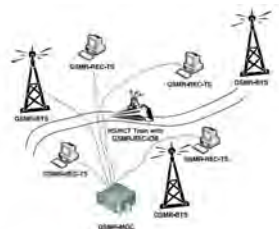
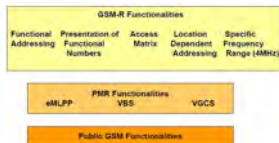
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1. GSM-R for high-speed railway

- The high-speed railway is critical for transporting commodities and passengers, and it has experienced rapid development recently.
- The primary consideration of high-speed railway is **safety**, which increasingly relies on the information and communication system.
- So it requires **realtime** measurement to ensure the reliability and stability of GSM-R networks and the high-speed railway system.^[1]



[1] G. Baldini et al. "An early warning system for detecting GSM-R wireless interference in the high-speed railway infrastructure". In: [International Journal of Critical Infrastructure Protection](#) (2010).

2. Require: On-line Monitoring System for GSM-R Networks

- ① It is crucial to reduce the estimation overhead so that the **on-line monitoring** can be implemented and ensure the realtime reliability.
- ② It is necessary to make **dynamic measurement** due to the feature of propagation environments along the high-speed railway routes.

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 - ② It is necessary to make **dynamic measurement** due to the feature of propagation environments along the high-speed railway routes.
- Difficulties:
 - Speed** 250-300km/h for China's high-speed railway;
 - Terrains** mountains, viaducts, plains, etc. along the routes;
 - Interface** vulnerable to changes of propagation environments;
 - Services** the communication may be affected by measurement.

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 - Terrains** mountains, viaducts, plains, etc. along the routes;
 - Interface** vulnerable to changes of propagation environments;
 - Services** the communication may be affected by measurement.
 - Advantages:
 - Flat** the propagation environments are generally flat;
 - Fixed** the trajectory and speed of trains are relatively fixed.

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1. Propagation Model

- 1 Since GSM-R networks are deployed along the railway routes with varied terrains, the propagation environments are very complex.
- 2 The cell radius is normally designed short, so the multi-path fading should be characterized by Rician fading in this case.



(a) Viaduct



(b) Tunnel



(c) Mountain



(d) Plain

Figure 1: Propagation environments and terrains of GSM-R networks

1. Propagation Model: $p_r^2(x) = s(x)h(x)$

1 Shadowing fading:

$$s(x) \sim N\left(m(x), \sigma_s^2\right) \quad (1)$$

2 Multi-path fading:

$$h(x) = \underbrace{\frac{1}{\sqrt{1+K}} \lim_{M \rightarrow \infty} \frac{1}{\sqrt{M}} \sum_{m=1}^M a_m e^{j\left(\frac{2\pi}{\lambda} \cos(\theta_m x) + \phi_m\right)}}_{\text{NLOS Components}} + \underbrace{\sqrt{\frac{K}{1+K}} e^{j\left(\frac{2\pi}{\lambda} \cos(\theta_0 x + \phi_0)\right)}}_{\text{LOS Component}} \quad (2)$$

2. Measurement Procedures

The procedures of propagation measurement in GSM-R networks is typically composed of the local mean power estimation, propagation prediction and model correction, as is demonstrated in Fig. 2.

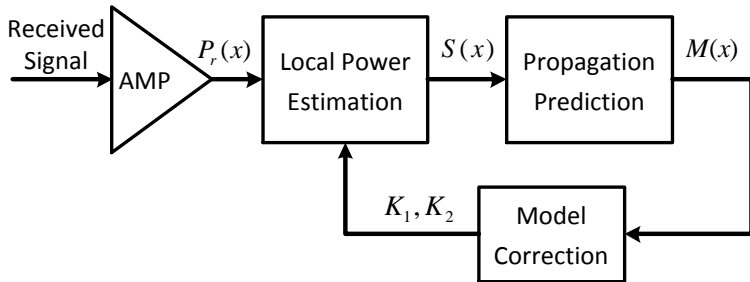


Figure 2: Basic Procedures of Radio Propagation Measurement

3. Sampling Frequency

- 1 $P_r(x)$ is influenced by different environments as shown in Fig. 3a and Fig. 3b, it should be adaptive to the networks status.
- 2 $P_r(x)$ is changing in both large and small time scale as shown in Fig. 3c, it should also be adaptive to this realtime fluctuation.

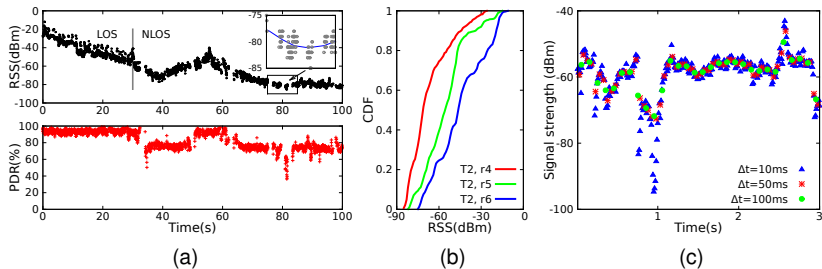


Figure 3: Character of RSS in mobile networks, composed of both of LOS and NLOS scenarios.

Traditional Algorithms on Local Power Estimating

- Lee's method proposed a standard process of local mean power estimation, which is determined in Rayleigh fading channels.^[2]
- Other works are based on confidence degree or ML estimation, but are also analyzed in Rayleigh channels.^[3]
- For the estimation of the received signal strength in Rician fading channels, the estimation overhead are usually high for GSM-R.^[4]
- The Generalized Lee method does not need a priori knowing of distribution function, but the optimal length of averaging interval is calculated by all the routes of the data with high overhead.^[5]

[2] W.C.Y. Lee. "Estimate of local average power of a mobile radio signal". In: [IEEE Trans. on Vehicular Technology](#) (1985), pp. 22–27.

[3] Bo Ai et al. "Theoretical analysis on local mean signal power for wireless field strength coverage". In: [WCSP '2009](#).

[4] C. Tepedelenlioğlu et al. "Estimation of Doppler spread and signal strength in mobile communications with applications to handoff and adaptive transmission". In: [Wireless Commun. and Mobile Computing](#) (2001), pp. 221–242.

[5] D. de la Vega et al. "Generalization of the Lee Method for the Analysis of the Signal Variability". In: [IEEE Trans. on Vehicular Technology](#) 58.2 (2009), pp. 506–516.

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On-line Estimating Procedure

The on-line estimating algorithm adopts the Lee's standard procedure in the case of Rician fading channels. Fig. 4 illustrates the basic steps which mainly consist of the determination of **proper length of statistical interval** and **required number of averaging samples**.

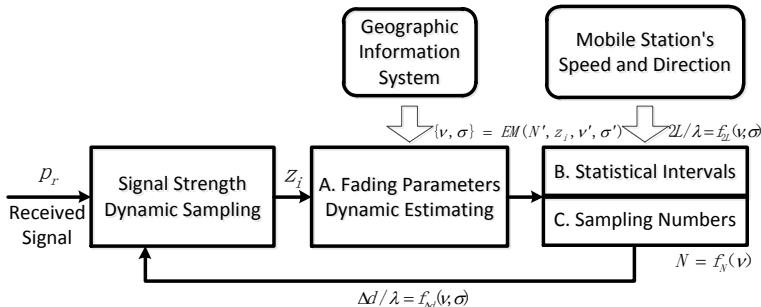


Figure 4: On-line and Dynamic Estimation of Rician Fading Channels

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1. EM algorithm for Rician estimation^[6]

$$\nu_{k+1} = \frac{1}{N} \sum_{i=1}^N \frac{I_1\left(\frac{\nu_k z_i}{\sigma_k^2}\right)}{I_0\left(\frac{\nu_k z_i}{\sigma_k^2}\right)} z_i \quad (3)$$

$$\sigma_{k+1}^2 = \max \left[\frac{1}{2N} \sum_{i=1}^N z_i^2 - \frac{\nu_k^2}{2}, 0 \right] \quad (4)$$

where N is the number of samples. The initial values are:

$$\nu_0 = \left(2 \left(\frac{1}{N} \sum_{i=1}^N z_i^2 \right)^2 - \frac{1}{N} \sum_{i=1}^N z_i^4 \right)^{1/4} \quad (5)$$

$$\sigma_0^2 = \frac{1}{2} \left(\frac{1}{N} \sum_{i=1}^N z_i^2 - \nu_0 \right) \quad (6)$$

[6] T.L. Marzetta. "EM algorithm for estimating the parameters of a multivariate complex Rician density for polarimetric SAR". In: *International Conference on Acoustics, Speech, and Signal Processing, 1995. IEEE. 1995*, pp. 3651–3654.

2. Length of Statistical Intervals

The local mean power can be estimated by the integral spatial average of $p_r^2(x)$:

$$\hat{s} = \frac{1}{2L} \int_{y-L}^{y+L} p_r^2(x) dx = \frac{s}{2L} \int_{y-L}^{y+L} h(x) dx \quad (7)$$

$$\sigma_{\hat{s}}^2 = \frac{2(n-1)}{n^2(1+K)^2} \int_0^n g(K; \rho) d\rho \quad (8)$$

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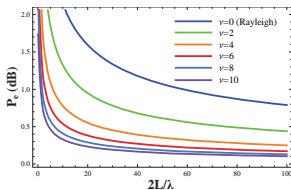
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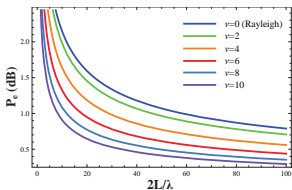


$$P_e = 10 \log_{10} \left(\frac{\hat{s} + \sigma_{\hat{s}}}{\hat{s} - \sigma_{\hat{s}}} \right) = 10 \log_{10} \left(\frac{\frac{2\sigma^2 + \nu^2}{2\sigma^2} n + \sqrt{2(1+n) \int_0^n g\left(\frac{\nu^2}{2\sigma^2}; \rho\right) d\rho}}{\frac{2\sigma^2 + \nu^2}{2\sigma^2} n - \sqrt{2(1+n) \int_0^n g\left(\frac{\nu^2}{2\sigma^2}; \rho\right) d\rho}} \right) \quad (9)$$

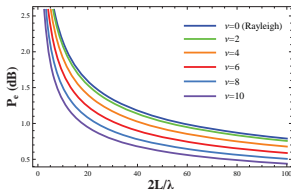
Local Power Estimation



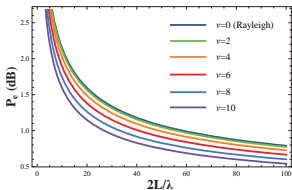
(a) $\sigma = 1$



(b) $\sigma = 3$



(c) $\sigma = 5$



(d) $\sigma = 7$

$$P_e = 1 \text{ dB}$$
$$\Downarrow$$
$$2L = f_{2L}(\lambda; \nu, \sigma)$$

Figure 5: Proper Length of Statistical Intervals

3. Number of Averaging Samples

The received power $r^2 = 2\sigma^2 + \nu^2 \approx \frac{1}{N} \sum_{i=1}^N z_i^2$ can be calculated by (3) and (4), then the expectation and variance of r^2 can be calculated:

$$\bar{r}^2 = E[r^2] = \frac{1}{N} E\left[\sum_{i=1}^N z_i^2\right] = \frac{\sigma^2}{N} (2N + \nu^2) \quad (10)$$

$$\sigma_{\bar{r}^2} = D[r^2] = \frac{1}{N^2} D\left[\sum_{i=1}^N z_i^2\right] = \frac{\sigma^4}{N^2} (4N + 4\nu^2) \quad (11)$$

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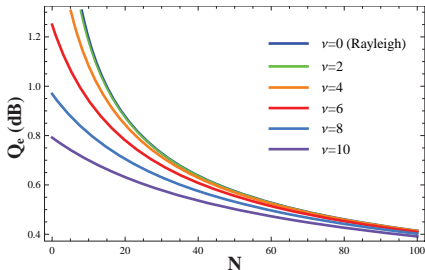
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$$\begin{aligned} Q_e &= 10 \log_{10} \left(\frac{\bar{r}^2 + \sigma_{\bar{r}^2}}{\bar{r}^2} \right) = 10 \log_{10} \left(\frac{\frac{\sigma^2}{N} (2N + \nu^2) + \frac{2\sigma^2}{N} \sqrt{N + \nu^2}}{\frac{\sigma^2}{N} (2N + \nu^2)} \right) \\ &= 10 \log_{10} \left(\frac{2N + \nu^2 + 2\sqrt{N + \nu^2}}{2N + \nu^2} \right) \end{aligned} \quad (12)$$

Local Power Estimation



Number of Samples

$$Q_e = 1 \text{ dB}$$



$$N = f_N(\lambda; \nu, \sigma)$$

Figure 6: Required Number of Averaging Samples

Local Power Estimation

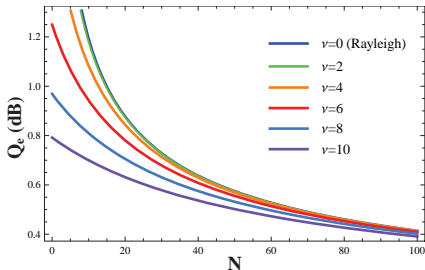


Figure 6: Required Number of Averaging Samples

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Sampling Intervals

$$\Delta d = 2L/N$$

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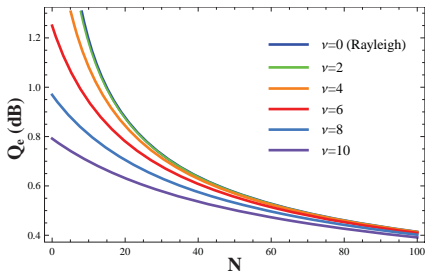


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$$\Delta d = 2L/N = f_{2L}(\lambda; \nu, \sigma) / f_N(\lambda; \nu, \sigma) = f_d(\lambda; \nu, \sigma)$$

- $\Delta d \Leftarrow$ statistical interval $2L$ and number of averaging samples N ;
- $\Delta d \Rightarrow$ measurement accuracy and overhead of on-line estimation.

Distance Driven: Δd

- 1 SDU: the radars and speed sensors are required
- 2 GPS: the accuracy is limited with additional overhead of communication

Time Driven: Δt

- 1 Accuracy: the speed and wave length are steady
- 2 Overhead: the system only needs velocity information from speed sensor

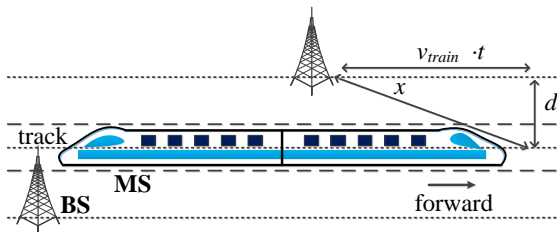


Figure 7: The distance between MS and BS.

Relative Distance

Since BSs are settled so close to the railway track, the relative distance of MS and BS can be deemed as $\Delta x = v_{train} \cdot \Delta t$.

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1. Platform of On-line Monitoring System

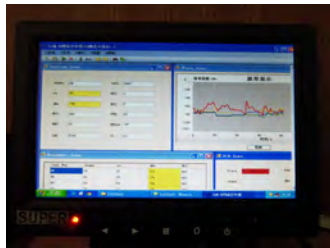
- Hardware: The CPU is RTD's CME137686LX-W, and the GSM-R module is COM16155RER-1 using Triorail's engine TRM:3a.
- Software: The software is developed by Microsoft .NET Compact Framework in C#, and it can run on Windows XP/CE/Mobile.

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(a) Hardware Design



(b) Software Development

Figure 8: Um Interface Monitoring System for GSM-R Networks

2. Algorithm Implementation and System Design

- The raw data is processed by the on-line estimation algorithm to provide current network status and conduct next signal sampling.
- The algorithm can also provide received signal strength prediction, and it will give the warning information when it is necessary.

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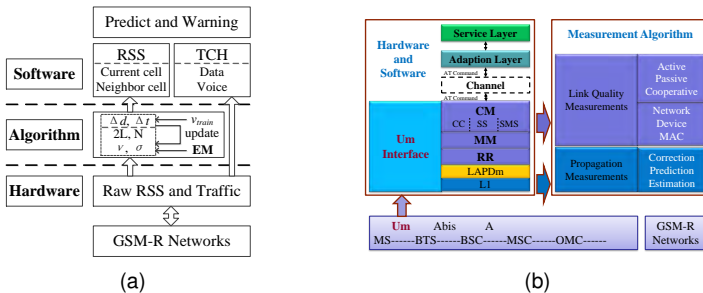


Figure 9: Measurement framework and algorithm implementation

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1. Experiments

- The experiment is carried out by the on-line monitoring system.
- The data was collected on Beijing-Shanghai high-speed railway.
- The collected data is also analyzed and evaluated by simulation.

1	TS	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z
1	2011-9-25 14:25:06	2	11	-99	460	0	2	0	0	3	8	-102	460	0	2	-9	-2	23	7	-102					
2	2011-9-25 14:25:07	2	9	-101	460	0	2	-2	-2	3	8	-102	460	0	2	-9	-3	23	7	-102					
3	2011-9-25 14:25:08	2	9	-101	460	0	2	-2	-2	3	8	-102	460	0	2	-9	-3	23	8	-102					
4	2011-9-25 14:25:09	2	8	-102	460	0	2	-3	-3	3	8	-102	460	0	2	-9	-3	23	8	-102					
5	2011-9-25 14:25:10	2	9	-101	460	0	2	-2	-2	23	8	-102	460	0	0	-3	-1	2	7	-102					
6	2011-9-25 14:25:11	23	9	-101	460	0	0	0	0	3	9	-101	460	0	2	-2	-2	2	6	-104					
7	2011-9-25 14:25:12	23	10	-100	460	0	0	1	1	3	7	-102	460	0	2	-4	-4	2	7	-102					
8	2011-9-25 14:25:30	23	10	-100	460	0	0	1	1	29	8	-102	460	0	0	-7	-7	2	8	-102					
9	2011-9-25 14:25:31	23	11	-99	460	0	0	2	2	3	9	-101	460	0	2	-2	-2	29	8	-102					
10	2011-9-25 14:25:32	23	11	-99	460	0	0	2	2	2	9	-101	460	0	2	-2	-2	29	8	-102					
11	2011-9-25 14:25:33	23	10	-100	460	0	0	1	1	2	8	-102	460	0	2	-2	-2	24	8	-102					
12	2011-9-25 14:25:34	23	10	-100	460	0	0	1	1	23	8	-102	460	0	2	-2	-2	24	8	-102					
13	2011-9-25 14:25:34	23	10	-100	460	0	0	1	1	23	8	-102	460	0	2	-2	-2	24	8	-102					
14	2011-9-25 14:25:35	2	10	-100	460	0	2	-1	-1	22	10	-100	460	0	0	3	1	29	7	-102					
15	2011-9-25 14:25:43	29	8	-102	460	0	0	-7	-7	23	6	-104	460	0	0	-5	-5	24	5	-105					
16	2011-9-25 14:25:44	548	7	-160	460	0	5	-4	2	29	6	-104	460	0	0	-9	-9	23	5	-105					
17	2011-9-25 14:26:03	2	8	-102	460	0	2	-3	-3	23	7	-102	460	0	0	-2	-2	29	7	-102					
18	2011-9-25 14:26:04	2	10	-100	460	0	2	-1	-1	23	7	-102	460	0	0	-2	-2	29	7	-102					
19	2011-9-25 14:26:05	2	12	-98	460	0	2	1	1	29	8	-102	460	0	0	-7	-7	23	7	-102					
20	2011-9-25 14:26:07	2	14	-96	460	0	2	3	3	23	9	-101	460	0	0	0	0	29	8	-102					
21	2011-9-25 14:26:08	2	15	-95	460	0	2	4	4	23	10	-100	460	0	0	1	1	29	9	-101					
22	2011-9-25 14:26:09	2	15	-95	460	0	2	4	4	29	11	-99	460	0	0	-4	-4	23	10	-100					
23	2011-9-25 14:26:10	2	16	-94	460	0	2	5	5	21	12	-98	460	0	0	3	3	29	12	-98					
24	2011-9-25 14:26:11	2	17	-93	460	0	2	6	6	22	13	-97	460	0	0	4	4	29	12	-98					
25	2011-9-25 14:26:12	2	17	-93	460	0	2	6	6	29	15	-95	460	0	0	0	0	23	13	-97					
26	2011-9-25 14:26:13	2	17	-93	460	0	2	6	6	29	15	-95	460	0	0	0	0	23	12	-98					
27	2011-9-25 14:26:14	2	17	-93	460	0	2	6	6	29	15	-95	460	0	0	0	0	23	12	-98					
28	2011-9-25 14:26:15	2	17	-93	460	0	2	6	6	29	15	-95	460	0	0	0	0	23	12	-98					
29	2011-9-25 14:26:16	2	17	-93	460	0	2	6	6	29	15	-95	460	0	0	0	0	23	12	-98					
30	2011-9-25 14:26:17	2	17	-93	460	0	2	6	6	29	16	-94	460	0	0	1	1	23	12	-98					
31	2011-9-25 14:26:18	2	17	-93	460	0	2	6	6	29	16	-94	460	0	0	1	1	23	11	-99					
32	2011-9-25 14:26:19	2	17	-93	460	0	2	6	6	29	16	-94	460	0	0	1	1	23	11	-99					

Figure 10: Experimental Results along Beijing-Shanghai High-speed Railway

1. Experiments

- The experiment is carried out by the on-line monitoring system.
- The data was collected on Beijing-Shanghai high-speed railway.
- The collected data is also analyzed and evaluated by simulation.



Figure 11: Experimental Results along Beijing-Shanghai High-speed Railway

2. Results

- Δd is more larger compared to Lee's method when $K = 0$, which means the multi-path fading is Rayleigh distributed.
- Δd may be not so small although $2L$ decreases, for $n < 5$ when the terrains gradually become flat until $\nu > 10$.

Table 1: Summary of Experiment Results

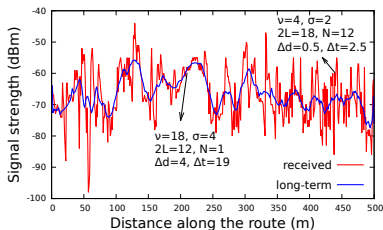
Terrain	K(dB)	ν	σ	$2L(\lambda)$	N	$\Delta d(\lambda)$	$\Delta d(m)$	$v_{train}(km/h)$		
								200	250	300
								$\Delta t(ms)$		
NLOS*	0	-	-	40	36	1.1	0.367	2.20	1.76	1.47
Intensive	0	0	1	55	15	3.7	1.222	7.33	5.86	4.89
	2	4	2	18	12	1.5	0.500	3.00	2.40	2.00
	4	5.6	2	9	9	1.0	0.333	2.00	1.60	1.33
	6	6	3	20	7	2.9	0.967	5.80	4.64	3.87
	8	12	3	8	1	8.0	2.667	16.00	12.80	10.67
Open	10	18	4	12	1	12.0	4.000	24.00	19.20	16.00

* Calculated by Lee's method in the case of Rayleigh fading

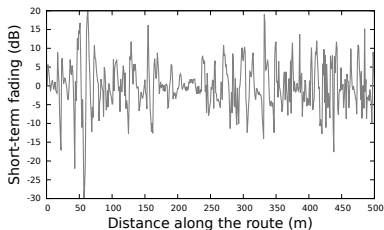
Evaluation

The long-term and short-term fading are differentiated separately.

- 1 Long-term: propagation prediction by ML or MMSE estimator.^[7]
- 2 Short-term: hysteresis selection in handoff algorithms.^[8]



(a) Signal Strength



(b) Short-term Fading

Figure 12: Measurement Results

[7] L. Gopal et al. "Power Estimation in Mobile Communication Systems". In: *Comp. and Info. Science* (2009), P88.

[8] K.I. Itoh et al. "Performance of handoff algorithm based on distance and RSSI measurements". In: *IEEE Trans. on Vehicular Technology* (2002), pp. 1460–1468.

3. Conclusions

- 1 The on-line and dynamic estimation algorithm and Um monitoring system is designed, and be tested by experiments&simulations.
- 2 EM algorithm is employed to reduce the estimation overhead: only the most recent samples instead of all routes of the database;
- 3 The measurement accuracy is guaranteed without unnecessarily frequent sampling: 12λ compared to Lee's 1.1λ for LOS signal.

The estimation algorithm can be used in upper layer applications:

- network planning with lower overhead, e.g., coverage assessment;
- real-time operating with dynamic adjustment to the time and space changes, e.g., channel allocation, power control and handoff;^[1]
- Since Rician fading is the generalized model of multi-path fading channels, the algorithm can also be introduced to other networks.

[1] G. Baldini et al. "An early warning system for detecting GSM-R wireless interference in the high-speed railway infrastructure". In: [International Journal of Critical Infrastructure Protection](#) (2010).

THANKS!

<http://yongsen.github.com>