Characterization, Signaling, Capacity and Coding Based on Fading Channels: A Review

Zhongli Wang¹, Guo Sheng², and Xu Zhang³

¹College of Electrical and Information Engineering, Beihua University, Jilin 132013, China
²Department of Communication Engineering, Zhejiang Post and Telecommunication College, Shaoxing 312016, China
³Department of Mechanical Engineering and Materials Science, Duke University, Durham, USA

Email: wqwaszx13579@163.com; sg@zptc.cn; xuzhangduke@gmail.com

Abstract—In this paper, the characterization, diversity, signaling, capacity and coding associated with fading channels are reviewed in an intuitive manner. We first introduce the basic concepts of fading and fading channels with an adequate number of derivations and simulations, by which the effects of fading on wireless communication can be analyzed comprehensively. Then, we also give the performance analysis of signals propagating over such a channel is presented and several special cases are discussed in depth with simulations. After this, we focus on diversity techniques, a powerful tool to overcome deep fading and improve the reliability of wireless communication. In addition, information-theoretic issues, e.g. channel capacity, outage probability and coding based on fading channels are also expatiated. By reading this paper, the readers are expected to have a broad understanding of fading phenomenon, a variety of types of fading channels, effects of fading on wireless communication and the methodologies to overcome the disadvantageous effects. Meanwhile, we also point out the future potential research directions in the relevant fields.

Index Terms—Fading, fading channel, diversity, channel capacity

I. INTRODUCTION

Wireless communication has become one of the most vibrant and dynamic research areas in electrical engineering [1]. In recent decades, wireless communication has eased people’s life dramatically and a series of novel mobile terminal devices, e.g. smartphone, tablets and laptop, have been ubiquitous [2]. However, until now, fading is still one of the main obstacles preventing the development of communication engineering [3]. The reason is that deep fading degrades the reliability of the communication systems and sometimes even makes the wireless communication impossible [4]. Therefore, the research into fading is of high importance and attracts a large number of researchers’ interest in the relevant fields [5]. In addition, fading channel is an important concept different from traditional channel models. Conventionally, to simplify the analysis of communication system and its performance, we simply use a classical Additive White Gaussian Noise (AWGN) channel model with a fixed attenuation factor or a linear filter with a fixed frequency response to model a communication channel [6]. However it is not always the case in practice [7]. In fact, a more complex and general channel model should consider the time-varying property, i.e. the randomness of the channel condition, e.g. path loss and shadowing, as well as the multipath effect [8]. This channel model is called fading multipath channel or simply fading channel. Due to the significance of fading and fading channel, we introduce, review and analyze them and the relevant techniques in depth in this paper. Moreover, we also point out the research directions worth further investigating in the relevant field in this paper.

The rest of this paper is organized as follows. Section II first introduces the fundamentals of fading and fading channels as well as their physical characteristics and mathematical models. Then, a crucial technique, diversity, is explained in Section III. Also, the signaling techniques of fading channel, as the important technique to transmit and receive signals, are analyzed and discussed in Section IV. As for the information theoretic issues, the capacity, outage probability and coding techniques are subsequently detailed in Section V. Finally, we conclude this paper and point out the future research directions in Section VI.

II. FUNDAMENTALS OF FADING AND FADING CHANNEL

A. Basic Concepts and Explanations of Fading and Fadingchannels

The most distinctive characteristics of fading channel is the randomness of channel condition and the multipath effect, which are also known as time-varying property and time spread respectively [9]. The former is used to characterize the randomness of the frequency response of a fading channel [10]. That is, when the same pulse is transmitted for many times, for each time, we measure it and will almost surely receive a different scaled and delayed version of the original transmitted signal. This can be vividly depicted in Fig. 1. The latter is caused by signal reflection, diffraction and scattering [11]. By these propagation mechanisms, not only a single path exists between the transmitter and the receiver, instead, a large number of paths exist. Therefore, when a single pulse is transmitted, a train of pulses will be received over such a fading channel. This can be graphically shown in Fig. 2.
Therefore, consider both characteristics jointly. The typical form of received random pulse train corresponding to the transmitted signal over the fading channel is shown in Fig. 3.

From (3), it is obvious that the equivalent baseband received signal is

\[ r(t) = \sum_{n=1}^{N} \alpha_n(t) e^{-j2\pi f_c(t)\delta[t - \tau_n(t)]} \]  (4)

Because \( \tau(t) = s(t) * c(t, t) \), the time-variant impulse response \([8]\) is

\[ c(t, t) = \sum_{n=1}^{N} \alpha_n(t) e^{-j2\pi f_c(t)\delta[t - \tau_n(t)]} \]  (5)

Besides, we can also assume \( \theta_n(t) = -2\pi f_c \delta[t - \tau_n(t)] \) and (6) becomes

\[ c(t, t) = \sum_{n=1}^{N} \alpha_n(t) e^{j2\pi f_c(t)\delta[t - \tau_n(t)]} \]  (6)

In addition, if the number of paths, i.e. \( N \) tends to infinity, we might as well use integral to express (3) and thus obtain \([8]\):

\[ r(t) = \text{Re} \left[ \int_{-\infty}^{\infty} c(t, t) e^{-j2\pi f_c(t)\delta[t - \tau_n(t)]} e^{j2\pi f_c(t)} dt \right] \]  (7)

Therefore, the corresponding time-variant impulse response is

\[ c(t, t) = e^{-j2\pi f_c(t)} \]  (8)

where \( c(t, t) \) is the impulse response of the fading channel at time \( t \) to an impulse applied at time \( t - \tau \).

Additionally, we can also assume \( \theta = -2\pi f_c \) and (8) becomes

\[ c(t, t) = e^{j2\pi f_c(t)} \]  (9)

Now, let us take a close look at the effects of fading channel on the received signal. Without considering initial value and channel noise, assume an all-one pulse as shown infra is transmitted:

\[ s(t) = 1, \quad \forall t \in \mathbb{R} \]  (10)

Then substituting (10) into (4), for the discrete fading channel mode given in (6), we have the equivalent baseband received signal:

\[ r(t) = \sum_{n=1}^{N} \alpha_n(t) e^{j\theta_n(t)} \]  (11)

From (11), it is obvious that the resultant received signal would fluctuate tempestuously due to the alternate constructive and destructive superpositions of received signals from multiple paths with random and time-variant multiplicative factor \( \alpha_n(t) e^{j\theta_n(t)} \). This phenomenon is termed signal fading \([8]\). To sum up, the primary cause of signal fading is due to the time-variant multipath characteristics of the fading channel \([9]\). Moreover, it can be proved that \( \theta_n(t) - U(0, 2\pi) \) and the distribution of the angle can be viewed as independent of \( \tau_n(t) \) \([14]\); \( \alpha_n(t) \) is distributed as Weibull distribution \([15]\). Besides, by central limit theorem, it can be further proved that
\[
\sum_{n=1}^{N} \alpha_n(t) \cos \theta_n(t) \sim N(\mu_n, \sigma_n)
\]
\[
\sum_{n=1}^{N} \alpha_n(t) \sin \theta_n(t) \sim N(\mu_n, \sigma_n)
\]

Therefore, according to the definition of Rayleigh distribution, it is easy to see that \( r \) the envelope of the resultant received signal:

\[
|E(t)| = \sqrt{\left[ \sum_{n=1}^{N} \alpha_n(t) \cos \theta_n(t) \sim N(\mu_n, \sigma_n) \right]^2 + \left[ \sum_{n=1}^{N} \alpha_n(t) \sin \theta_n(t) \sim N(\mu_n, \sigma_n) \right]^2}
\]

is a random Rayleigh distributed variable as long as both Gaussian random components share the same variance. Otherwise, the envelope is Rician distributed. Whether Rayleigh or Rician distribution should be adopted depending on whether there is a Light of Sight (LOS) path between the transmitter and the receiver [16].

Consider the case of a Rayleigh fading channel, i.e.

\[
\sum_{n=1}^{N} \alpha_n(t) \cos \theta_n(t) \sim N(0,1) \quad \text{and} \quad \sum_{n=1}^{N} \alpha_n(t) \sin \theta_n(t) \sim N(0,1)
\]

with 100 Hz sampling frequency, we can simulate and plot the amplitudes of transmitted signal \( s(t) \) and the resultant received signal \( r(t) \) with \( N = 20, N = 200 \) and \( N = 2000 \) in Fig. 4, Fig. 5 and Fig. 6 respectively. From these three figures, several important features associated with signal fading and multipath fading channel can be found and summarized. First, overall, a multipath fading channel has a positive effect on the signal transmission. Because of the constructive supposition of received signals from multiple paths, in most cases the amplitude of the resultant received signal is larger then the original transmitted signal, which provides a higher Signal-to-Noise Power Ratio (SNR) and thus a better performance than single path channel. Also, with an increasing number of paths, the average amplitude of the resultant received signal has been enlarged and the situation of signal fading has also been improved accordingly. However, the detrimental effect of a multipath fading channel should not be overlooked. As we can see from these figures, the amplitude of the resultant received signal is frequently lower than the amplitude of the transmitted signal, which will result in a degradation of the performance of communication systems. This degradation is termed deep fading and has a significant and negative impact on the system reliability, especially for received signals with memory [8]. To make our summary more reliable, we can also plots the Probability Density Functions (PDF) of these three cases in Fig. 7 and they further validate our analysis.
Based on this analysis of Rayleigh fading channel, we can easily discuss the scenario when there exists a LOS path. In this case, we should use Rician fading channel model. For the simplification and demonstration purposes, we can simply assume the amplitude of the received signal from LOS path is distributed as N(0; 1), whereas the amplitudes of other received signals from N indirect paths are distributed as N(0; 1). By this simulation configuration, we can simulate and plot the PDF of the amplitude of resultant received signal over Rician fading channel in Fig. 8. From this figure, it is clear that with a LOS path, the overall performance of the signal transmission can be improved, but with an increasing number of paths, this improvement effect is weakened.

### B. Time Domain and Frequency Domain Properties of Fading Channels

Based on the assumptions of wide-sense-stationary (WSS) channel and the uncorrelated scattering, i.e. the received signals from different paths with different delays can be regarded as statistically uncorrelated in terms of attenuation and phase shift, we can thus calculate the autocorrelation function of \( c(t) \) by [8]:

\[
R_c(\tau_1, \tau_2; \Delta \tau) = \mathbb{E}[c^*(\tau_1) c(\tau_2; t + \Delta \tau)] = R_c(\tau_1; \Delta \tau) \delta(\tau_2 - \tau_1)
\]

where \( \tau_1 \) and \( \tau_2 \) are the time delays of two received signals transmitted at time \( t + \Delta \tau \) and \( t \), respectively.

If we assume two signals are transmitted at the same time and received with a time difference \( \tau \), then the multipath intensity profile or called delay power spectrum of the channel can be defined as [8]:

\[
R_c(\tau; 0) = R_c(\tau)
\]  

Also, we can define the multipath spread \( T_m \) of a channel as [8]:

\[
\exists \tau \in (0, T_m) \text{ s.t. } R_c(\tau) \neq 0
\]  

In order words, the variable \( T_m \) characterizes the time spread of a channel and once a sharp and narrow pulse is sent, the received signals could spread within \( (0, T_m) \) in time domain [17].

Additionally, if we define the channel frequency response as [18]:

\[
C(f; t) = \int_{-\infty}^{\infty} c(\tau; t)e^{-j2\pi f \tau} d\tau
\]

Based on this definition, analogously, we can define the autocorrelation function of two frequency responses with different carrier frequencies, termed spaced-frequency spaced-time correlation function of a channel [8]:

\[
R_c(f_2, f_1; \Delta f) = \mathbb{E}[C^*(f_1; t) C(f_2; t + \Delta f)] = R_c(\Delta f; \Delta t)
\]

where \( \Delta f = f_2 - f_1 \).

Similarly, if there is no difference between transmission time instants of two signals, we have

\[
R_c(\Delta f, 0) = R_c(\Delta f) = \int_{-\infty}^{\infty} R_c(\tau)e^{-j2\pi f \tau} d\tau
\]

We ditto define the coherence bandwidth of a channel as [8]:

\[
\exists f \in (-\Delta f / 2, \Delta f / 2) \text{ s.t. } R_c(\Delta f) \neq 0
\]

According to (14), (17) and (18), we can approximate the coherence bandwidth of a fading channel [19]:

\[
\Delta f_c \approx \frac{1}{T_m}
\]

That is, if the carrier frequencies of two transmitted signals are separated larger than \( \Delta f_c \), they will be affected differently when propagating over the wireless fading channel in terms of attenuation factors and phase shifts and can be viewed as statistically uncorrelated. Assuming \( W \) is the bandwidth of the channel, if \( \Delta f_c \ll W \), the fading channel is termed frequency-selective fading channel, in which a small variation of carrier frequency of the transmitted signal will result in a significant difference of the amplitude and phase shift of the received signal [20]. Otherwise, the channel is termed frequency-nonselective fading channel or flat fading channel, which indicates all signals with different carrier frequencies are treated without bias [21].

Also, the difference of transmission time between two transmitted signals denoted by \( \Delta t \), will also matter. To analyze this effect, we should first define the Fourier transform of \( R_c(\Delta f; \Delta t) \) with respect to \( \Delta t \) as [8]:

\[
S_c(\Delta f; \lambda) = \int_{-\infty}^{\infty} R_c(\Delta f; \Delta t)e^{-j2\pi \lambda \Delta t} d\Delta t
\]

Without considering the difference between carrier frequencies, we can define the Doppler power spectrum of a channel as [8]:

\[
S_c(0; \lambda) = S_c(\lambda) = \int_{-\infty}^{\infty} R_c(0; \Delta t)e^{-j2\pi \lambda \Delta t} d\Delta t
\]

We ditto define the Doppler spread of a channel as [8]:

\[
\exists \lambda \in (-B_d / 2, B_d / 2) \text{ s.t. } S_c(\lambda) \neq 0
\]

From the definition of Doppler spread of a channel \( B_d \), we can define the coherence time of a channel as [19]:

\[
\Delta f_c \approx \frac{1}{T_m}
\]
This variable \( \Delta t \) characterizes the time domain property of the fading channel. That is, if the difference between the transmission time of two signals is greater than \( \Delta t \), they will be treated differently by the channel in terms of attenuation factor and phase shift and can be regarded as statistically uncorrelated [22]. If \( \Delta t \ll E[\tau] \), the channel is termed fast fading channel; otherwise, it is termed slow fading channel [8].

Overall, we can conclude that if the channel condition \( T_\alpha B_\beta \ll 1 \) is satisfied, it is possible to transmit a signal over this channel with flat time and frequency responses [8].

Finally, based on the analysis in this subsection, we can summarize the scattering function of a fading channel with time delay \( \tau \) and Doppler frequency shift \( \lambda \) by [8]:

\[
S_C(\tau; \lambda) = \int \int R_C(\Delta t; \Delta f) e^{j2\pi\Delta t} e^{j2\pi\Delta f} d\Delta t d\Delta f
\]

(24)

The variable \( S_C(\tau; \lambda) \) can be used to measure the average power output of a fading channel.

C. Statistical Models of Fading Channels

Although the fading channel has been referred many times as \( c(\tau; t) \) in the previous sections, we have not constructed the model of its generic form and the selection of parameters systematically. Therefore, in this subsection, we will dig into and construct its mathematical expressions.

The most common case of a fading channel is that there exist a large number of multiple indirect paths between the transmitter and the receiver. In this case, as analyzed before, the envelope of the channel frequency response is Rayleigh distributed and the phase angle is uniformly distributed ranging from 0 to \( 2\pi \). Although Rayleigh distribution is simple and easy to implement, it only contain a single characteristic parameter, which makes it sometimes inaccurate. Because the applicability of Rayleigh distribution for envelop modeling is based on the central limit theorem, if the number of multiple paths reduces, the Rayleigh distribution will become inaccurate and cannot fit the fading channel well [23]. In this case, Nakagami-m distribution, as a two-parameter distribution, would provide a better fit for wireless fading channel modeling [24].

In addition of a relatively infrequent case [25], if there is a LOS path between the transmitter and the receiver, Rician distribution would be proper to use for fading channel modeling, because it takes the unfairness of the received signals from multiple paths into consideration [26].

For a special case in which a LOS path and a small number of indirect paths exist, we can use some simplified and closed forms to model the channel impulse response \( c(\tau; t) \). For example, consider the airplane to ground communication as shown in Fig. 9. There are only two multipath components between the transmitter and the receiver, which are LOS path (marked in blue) component and a single multipath component (marked in red) owing to terrian reflections. In this case, the channel impulse response can be expressed as [8]:

\[
c(\tau; t) = \alpha c(\tau) + \beta(t) c(\tau - \tau_0(t))
\]

(25)

where \( \alpha \) is a constant attenuation factor of the LOS path; \( \beta(t) \) is the time-variant attenuation factor of the single indirect path; \( \tau_0(t) \) is the time-variant path delay of the single indirect path. According to (15) and (25), we have:

\[
C(f; t) = \alpha + \beta(t) e^{-j2\pi f \tau_0(t)}
\]

(26)

Based on (26), a more sophisticated and specific model used for microwave LOS radio channel of long-distance signal transmission can be constructed as below [27]:

\[
C(f) = \alpha \left[ 1 - \beta e^{-j2\pi f \tau_0} \right]
\]

(27)

where \( \alpha \) is an overall attenuation parameter; \( \beta \) is the shape parameter caused by multipath components; \( \tau_0 \) is the frequency of the fade minimum; \( \tau_0 \) is the relative time delay between the LOS path and the indirect path.

In this model, we assume that the difference of time delays among different indirect paths is rather small in comparison with the time delay between the LOS path and any indirect path. Also, it is clear that this model is an approximation of time-invariant channel model and can ease the analysis for this special case.

D. Performance Analysis of Signals Propagating over Flat and Slow Fading Channel

As long as \( T_\alpha B_\beta \ll 1 \), it could be proved that signals transmitted over a flat and slow fading channel will experience a multiplicative distortion with random but time-invariant attenuation factor and phase shift within one or multiple signaling intervals [8]. Therefore, consider the equivalent baseband signal model given above. Within a signaling interval \( [0, T] \), we have:

\[
\tau(t) = \alpha e^{j\phi} z(t) + z(t)
\]

(28)

where \( \alpha \) is the constant attenuation factor; \( \phi \) is the phase shift; \( z(t) \) is the complex-valued AWGN term.

Because the signals are transmitted over flat and slow fading channel, the channel frequency response, i.e. \( \alpha e^{j\phi} \), can be estimated easily by pilot signals or training...
sequences [28], [29]. Hence, we can use matched filter and coherent detection techniques to determine the received signal $r(t)$ [30]. Obviously, the error probability of coherent detection only depends on the randomness of the attenuation factor. Assuming for a fixed attenuation factor, the error probability is denoted as $P_{e}(\alpha)$. Then, if the attenuation factor is random and distributed by a PDF $p_{\alpha}(\alpha)$ average error probability can be expressed by [31]:

$$P_{e} = E[P_{e}(\alpha)] = \int_{-\infty}^{\infty} P_{e}(\alpha) p_{\alpha}(\alpha) d\alpha$$

(29)

Meanwhile, by fundamentals of signal processing [18], we can also use SNR per bit to express the relations given above:

$$\gamma = \frac{\alpha^2 \varepsilon_0}{N_0}$$

(30)

where $\varepsilon_0$ is the energy in each information bit; $N_0$ is the power spectral of complex-valued AWGN.

Therefore, based on the relation given in (30), (29) can be rewritten as:

$$P_{e} = E[P_{e}(\gamma_b)] = \int_{-\infty}^{\infty} P_{e}(\gamma_b) p_{\gamma_b}(\gamma_b) d\gamma_b$$

(31)

For BPSK signal, the fixed attenuation error probability is [32]:

$$P_{e-BPSK}(\gamma_b) = Q(\sqrt{2\gamma_b})$$

(32)

where $Q(\cdot)$ is the tail probability of the standard normal distribution.

Further assume that the transmitted signal propagates over a flat and slow Rayleigh fading channel. We can determine the PDF of $\alpha$ [33]:

$$p_{\alpha}(\alpha) = \frac{\alpha}{\sigma^2} e^{-\alpha^2/2\sigma^2}, \quad \alpha > 0$$

(33)

$$0, \quad \text{otherwise}$$

where $\sigma$ equals to 1 for standard Gaussian random process [33].

Due to (30) and (33), we have:

$$p(\gamma_b) = \frac{1}{\gamma_b} e^{-\gamma_b/\gamma_0}, \quad \gamma_b > 0$$

(34)

where $\gamma_0 = \varepsilon_0 E(\alpha^2)/N_0$.

Also, for a standard Gaussian random process, we can determine $E(\alpha^2)$ by [8]:

$$E(\alpha^2) = \text{var}(\alpha) + [E(\alpha)]^2 = \left(2 - \frac{\pi}{2}\right) \sigma^2 + \left(\sigma \sqrt{\frac{\pi}{2}}\right)^2 = 2$$

(35)

According to (29), (32) and (34), we can obtain the average error probability for BPSK [8]:

$$P_{e-BPSK} = \frac{1}{2} \left(1 - \frac{\gamma_b}{1 + \gamma_b}\right)$$

(36)

Following a similar derivation procedure of the BPSK case, for BFSK with coherent detection, BFSK with noncoherent detection and DPSK, we have [8]:

$$P_{e-BFSK(C)} = \frac{1}{2} \left(1 - \frac{\gamma_b}{2 + \gamma_b}\right)$$

(37)

$$P_{e-BFSK(NC)} = \frac{1}{2 + \gamma_b}$$

(38)

$$P_{e-DPSK} = \frac{1}{2(1 + \gamma_b)}$$

(39)

Based on these relations derived above, we can simulate and plot the results in Fig. 10. From this figure, it is clear that BPSK outperforms other three signaling schemes due to its constant envelope property, i.e. $|s_1(t)| = |s_2(t)|$. It can also be observed that the performances of BFSK with coherent detection and DPSK are similar, especially when SNR is relatively high. However, because the channel state is not estimated by noncoherent detection, the BFSK with noncoherent detection has the worst performance in comparison with other three signaling schemes [8].

![Fig. 10. Relations between $\gamma$ and $P_e$ for BPSK, BFSK with coherent detection, BFSK with noncoherent detection and DPSK.](image)

III. DIVERSITY AND RECEIVING TECHNIQUES FOR SIGNALS TRANSMITTED OVER FADING CHANNELS

A. Concept and Classification of Diversity

As the transmission and signaling techniques have been analyzed in the previous section, we focus on diversity and receiving techniques of signals transmitted over fading channels in this section.

Diversity technique is based on the viewpoint that if a large number of replicas of the same signal are transmitted, even though several replicas experience deep fading and are severely distorted when received, the transmitted signal is still recoverable from the combination of all replicas [34]. This idea is feasible,
because fading channels of different signal replicas are statistically independent, simultaneous deep fading for all signal replicas is almost surely impossible as long as the number of signal replicas is sufficiently large. Theoretically, the performance and reliability of the communication system can be improved by this way. By fundamental statistics [33], assuming \( p_f \) is the probability of the occurrence of deep fading of a transmitted signal replica and the number of replicas is \( L \), we can determine the probability of the event that all signal replicas are deep faded over statistically independent fading channels is

\[
P_f = \prod_{l=1}^{L} p_f^l
\]

As the merit of diversity has been analyzed, it is the time to move forward to the implementation of diversity techniques. There are a number of approaches to provide diversity for the receiver, e.g. frequency diversity, time diversity and space diversity. For frequency diversity, we can employ a set of \( L \) carriers with frequency separation greater than \( \Delta f \), and thus they are statistically uncorrelated when received [35]. Alternatively, time diversity can also be utilized. To achieve time diversity, we can also transmit signal replicas by \( L \) time slots with a time separation greater than \( \Delta t \) [36]. As for space diversity, it is also known as antenna diversity and utilizes multiple antennas to transmit and receive signal replicas [37]. To guarantee the feasibility of space diversity, we should ensure different antenna pairs are placed with a sufficiently large distance so that the channels among them can be viewed as statistically independent [8]. More complicated, we can transmit a wideband signal with bandwidth \( W \gg \Delta f \). Therefore, in different correlated frequency intervals, the frequency components of the entire wideband signal will be treated differently and form a large number of statistically independent signal segments with different central frequencies at the receiver [8]. Owing to this statistical independence, they can provide frequency diversity as well. Accordingly, a RAKE correlator or called RAKE matched filter can be applied to receive and detect these signal segments [38].

B. Communication System and Channel with Diversity

Following the assumption of flat and slow Rayleigh fading channel stated in the last section, with such \( L \) i.i.d. diversity channels, we can express the equivalent baseband received signal of a \( M \)-ary modulation scheme by [39]:

\[
r_b (t) = \alpha_k e^{j k} s_m (t) + z_k (t)
\]

where \( k \in \{1, 2, \ldots, L\} \); \( m \in \{1, 2, \ldots, M\} \); \( \{s_m (t)\} \) represents the equivalent baseband \( m \)-th transmitted signal over the \( k \)-th channel; \( \alpha_k e^{j k} \) represents the attenuation factor and phase shift of \( k \)-th channel; \( z_k (t) \) is the AWGN term of the \( k \)-th channel.

Because \( \{s_m (t)\} \) is manipulated by a certain modulation scheme, it is also known to the receiver. Hence, the receiver can utilize \( M \) matched filters with impulse responses infra to detect the received signals [40]:

\[
b_m (t) = \sum_{k=0}^{L} s^*_{m,k} (T-t)
\]

Once the channel can be perfectly estimated, i.e. \( \{\alpha_k\} \) and \( \{\phi_k\} \) are known to the receiver, the index of transmitted signal \( m \) can be optimally determined by the algorithm of maximal ratio combination (MRC) as illustrated in (43) [41]:

\[
m_k = \arg \max_{m=0, 1, \ldots, M} \left\{ \sum_{k=0}^{L} \alpha_k e^{j \phi_k} s^*_m (t) + z_k (t) \right\}
\]

The multiplicative term \( \{\alpha_k e^{j \phi_k}\} \) shown in this equation is used to compensate the effect of phase shift and ease the elimination of deep faded received signals [8]. Finally, such a detection procedure involving MRC and diversity channels can be summarized in Fig. 11.

Fig. 11. Detection of received signals propagating over diversity fading channels by MRC [8].

Without loss of generality, we can take BPSK as a specific example to illustrate this procedure in details. Following (31), (32) and (43), we can obtain the average error probability in this case [8]:

\[
P_{e-BFSK-L} = \frac{1}{2} \left[ 1 - \sqrt{\frac{1}{1 + 1 + 1}} \right]^L \times \sum_{k=0}^{L} \left( L - 1 + k \right) \left[ \frac{1}{2} \left( 1 + \sqrt{\frac{1}{1 + 1}} \right) \right]^k
\]

Fig. 12. Error probability of BPSK with different diversity.
In comparison with (36), we can simulate and illustrate the results in Fig. 12. From this figure, some key points can be summarized. First, diversity is a very powerful tool to combat deep fading and improve the reliability of wireless communication systems. However, its feasibility is based on the large SNR. Otherwise, when $\bar{\gamma}_b \ll 1$, the effect of diversity becomes inconspicuous.

Simulation of (45) using MATLAB yields the results shown in Fig. 13.

Analogously, for DPSK, following the deviation procedure of BPSK, we have [8]:

$$P_{b-DPSK-L} = \frac{1}{2^{2L-1}(L-1)!}(1+\frac{1}{\bar{\gamma}_b})^{L-k} \times \sum_{k=0}^{L-1} \left[ \binom{L-1+k}{k} \left( \frac{2L-1}{n} \right)^{L-k} \right]$$ (47)

Simulation of (47) using MATLAB yields the results shown in Fig. 15.

It is not surprising that all modulation schemes presented above follow the same trend when diversity technique is employed. This consistency validates the practicability and universality of diversity as well as its advantageous effects. That is, whichever modulation scheme is used, diversity will always improve the reliability of the communication system and with different SNR, the improvement effect could be obvious or trivial. Now let us move our attention to the comparison among different modulation schemes with the same number of diversity channels, say $L=16$. The comparison results are illustrated in Fig. 16. From this figure and Fig. 10, we can find that although the advantageous effects of diversity are suited for all of these four modulation schemes, BFSK with coherent detection obtains a more obvious improvement when...
diversity technique is performed and the improvement of BPSK with noncoherent detection is relatively small.

IV. SIGNALING TECHNIQUES OF FREQUENCY-SELECTIVE FADING CHANNELS

A. Wideband Signal and Tapped Delay Line Channel Model

In order to obtain diversity and overcome deep fading, the basic signaling philosophy of transmitted signal over fading channels can be described as follows. Either modulating a signal so that it has a bandwidth $W \ll \Delta f$, or subdividing its entire bandwidth $W$ so that each subdivision has a bandwidth $W_c \ll f_c$. Alternatively, a more intuitive and convenient way to achieve this goal is to employ a wideband signal with bandwidth $W \gg \Delta f$, and transmission pulse interval $T \ll t_c$ [8]. In this case, the fading channel can be viewed as a frequency-selective and slow fading channel. The frequency components of the transmitted signal over this channel are treated differently but the channel frequency response is WSS. By sampling theory, such a wideband signal has an equivalent baseband signal as expressed below:

$$s_i(t) = \sum_{n=-\infty}^{\infty} s_i \left( \frac{n}{W} \right) \sin \left[ \frac{\pi W(t - n/W)}{\pi W} \right]$$

(48)

For such a transmitted signal, according to (3) and (48), the noiseless equivalent baseband received signal over a frequency-selective facing channel can be expressed as:

$$r(t) = \frac{1}{W} \sum_{n=-\infty}^{\infty} s_i \left( t - \frac{n}{W} \right) c_i \left( \frac{n}{W}, t \right)$$

(49)

As long as the number of subchannels is large enough, for each frequency interval $1=W$ the channel frequency response can be regarded as flat and the channel impulse response can be viewed as only depending on time $t$ and thus we have the expression given infra [8]:

$$c_i(t) = \frac{1}{W} \left( \frac{n}{W}, t \right)$$

(50)

This provides a foundation for us to model the fading channel as a filter with a number of taps. In practice, the number of filter taps is a finite value and thus a filter system with infinite taps should be truncated for implementation purpose. Let $L = \lceil TnW \rceil + 1$ and the truncated tapped delay line channel model can be expressed by [18]:

$$r(t) = \sum_{n=1}^{L} c_i (t) s_i \left( t - \frac{n}{W} \right)$$

(51)

In accordance with the aforementioned assumptions of uncorrelated scattering and the specification of wideband signal, the tap coefficients $c_i(t)$ is statistically independent and time-variant. Therefore, $L$ signal replicas are provided and form a diversity for the receiver. This truncated tapped delay line channel model can also be graphically depicted in Fig. 17. By this way, the reliability of the communication system can be improved by employing wideband signals.

In Fig. 17, tapped delay line model of frequency-selective fading channels [8].

B. RAKE Demodulator

As we mentioned in Section III, RAKE demodulator is a powerful tool to process multiple signal replicas and detect the wideband signal. In this subsection, it will be analyzed in details.

For the simplification and demonstration purposes, we might first consider $M$-ary signaling scheme consisting of $M$ homenergic signal waveforms $(s_{in}(t))$ over the frequency selective fading channel, where $m \in \{1,2,\ldots,M\}$. As the specification of wideband signals, the intersymbol interference (ISI) caused by multipath effect can be ignored, since the transmission pulse interval $T$ of a wideband signal is far greater then the time spread $T_{aw}$. Because we also assume $T \ll \Delta f$, the fading channel varies slowly compared to the transmitted signal. As a consequence, it is reasonable to assume the fading channel can be estimated perfectly and thus $\{c_i(t)\}$ is known to the receiver. According to (43) and assuming the index of transmitted signal is $m$, the detection procedure of RAKE demodulator can be expressed in (52). This detection procedure involving RAKE demodulator can also be graphically depicted in Fig. 18. On the other hand, if ISI is not negligible, a channel equalizer could be inserted between the sampler and the integrator to compensate the nonlinear effect caused by ISI [8].

$$m = \arg \max_{n=1,2,\ldots,M} \left\{ \int_{t}^{t+T} r(t) \left[ \sum_{k=1}^{K} c_i \left( t - \frac{k}{W}, t \right) \right] \left[ T - \left( \tau - t + \frac{k}{W} \right) \right] \right\}$$

(52)
Having introduced the frequency domain property of OFDM over fading channel, it is the time to move towards time domain property. In general for practical communication systems and channels, we have $T \ll \Delta T_c$ [1]. Therefore, the fading channel can be regarded as slow and WSS during the transmission interval of each OFDM symbol. To model such a slow fading channel, we might use the two-term Taylor series expansion to approximate the channel attenuation factor $a_k(t)$ during each symbol interval $t \in [0, T]$ [45].

$$a_k(t) = a_k \left(\frac{T}{2}\right) + \left(t - \frac{T}{2}\right) a'_k \left(\frac{T}{2}\right)$$

(57)

Substituting (57) into (54) yields the general form of channel impulse response of $k$-th OFDM subchannel within a symbol interval:

$$c_k(\tau; t) = a_k \left(\frac{T}{2}\right) \delta(\tau) + \left(t - \frac{T}{2}\right) a'_k \left(\frac{T}{2}\right) \delta(\tau)$$

(58)

Based on (4) and (58), the equivalent baseband received signal can be determined by [8]:

$$r(t) = \frac{1}{\sqrt{T}} \sum_{i=1}^{N} a_k \left(\frac{T}{2}\right) s_k e^{j2\pi f/2} + \frac{1}{\sqrt{T}} \sum_{i=1}^{N} \left(t - \frac{T}{2}\right) a'_k \left(\frac{T}{2}\right) s_k e^{j2\pi f/2} + n(t)$$

(59)

where $\{s_k\}$ represents the complex valued signal constellation points stipulated by a certain modulation scheme; $n(t)$ is the AWGN term.

To demodulate these $N$ constellation points contained in the received signal $r(t)$, we can simply use $N$ parallel correlators with the correlation frequencies corresponding to the OFDM subcarriers’ frequencies [46]. Then the output of $i$-th correlator can be determined by [8]:

$$\hat{s}_i = \frac{1}{\sqrt{\frac{T}{2}}} \int_{0}^{T} r(t) e^{-j2\pi f}\,dt = \begin{cases} 1, & i = k \\ 0, & i \neq k \end{cases}$$

(60)

In (60), we have separated the effects of the transmitted signal, ICI and noise. Because for slow fading channel, $a_k(t)$ is estimable, we can simply send a scaled version of $\hat{s}_i$ to detector to evaluate the transmitted constellation point of this subcarrier. Involving mean-square values, we can calculate the energy of desired signal and ICI respectively by [47]:

$$S = E\left[|a_k(T/2)s_k|^2\right]$$

(61)

and

$$I = E\left[\frac{T}{2\pi} \sum_{k=1}^{N} a'_k(T/2) s_k^2 \right]$$

(62)
where \( f_m \) is the maximum Doppler frequency.

Therefore, we can determine the signal-to-interference ratio (SNIR) by:

\[
SNIR = \frac{S}{I} = \frac{1}{\left(\frac{f_m}{2}\right)^2} \sum_{k=-\infty}^{\infty} \frac{1}{(k - t)^2}
\]  

(63)

Finally, we are able to use (63) as a benchmark to evaluate the performance of an OFDM system in fading channel and quantify the effect of ICI caused by Doppler shift.

V. CAPACITY AND CODING TECHNIQUES OF FADING CHANNELS

A. Fundamentals of Channel Capacity

In this section, the information-theoretic issues related to fading channels will be discussed. Several important concepts, e.g. ergodic capacity, outage capacity and outage probability, will also be addressed. Considering the properties of fading phenomenon and the time domain properties of different types of fading channels, the calculation of channel capacity should be classified into two different cases. One is the fast fading channel and another is the slow fading channel. For the former case, the channel state varies sharply and vehemently during a single transmission interval of a information block. As a result, during a transmission interval of a single block, all possible channel states will be experienced by the waveforms in the block. Hence, the ergodic channel model and ergodic capacity would be more appropriate to use in this case [46], [48], [49].

For the latter case, the channel state is quasi-static during a transmission interval of a block. Therefore, for different blocks, different channel states will be experienced but within a block, all signal waveforms will be assumed to experience the same channel state. Therefore, capacity in different block transmission interval is different and random. As a consequence, outage capacity should be used to characterize the channel [50]–[52]. More specifically, from information theory, we might know that the channel capacity characterizes the randomness and uncertainty of a channel [53]. Therefore, whether the channel state information (CSI) can be estimated will affect the channel capacity [54]–[56]. As we can see from Fig. 16 of BFSK signals with coherent detection and noncoherent detection, if CSI is unobtainable, the performance for the communication system will be degraded even the same modulation scheme is employed. In general, CSI can be known to the transmitter or the receiver or both [57]. In terms of the transmitting end, the receiver can use CSI to extract the channel attenuation factor and phase shift so that correlators can be employed to effectively demodulate the received signals [60].

According to information theory founded by Shannon, the concept of channel capacity is defined as the supremum of the transmission rate at which reliable communication over the channel is possible [61]. Otherwise, whatever we design the transmission and detection techniques, we would not be able to obtain reliable communication [8]. That is, as long as the transmission rate is less than channel capacity, we can always find out a coding scheme to achieve an infinitesimal error probability. As definition, the general form of the capacity of a discrete memoryless channel is:

\[
C = \max \limits_{p(x)} I(X;Y)
\]  

(64)

where \( p(x) \) is the probability of the occurrence of event \( x \), i.e. the information symbol \( x \) is transmitted; \( I(X;Y) \) is the mutual information between events sets \( X \), i.e. the possible transmitted symbols and \( Y \), i.e. the possible received symbols.

From a practical point of view, for a band-limited power-constrained discrete-time AWGN channel, we might use a reduced version of (64) to determine the channel capacity [8]:

\[
C = W \log \left( 1 - \frac{P_t}{N_0 W} \right)
\]  

(65)

where \( W \) is the channel bandwidth; \( P_t \) is the transmitter power; \( N_0 \) is the power spectral density of complex-valued AWGN.

However, we should also note that the capacity models constructed above are based on the assumption that the number of channel states is infinite. In practice, it is not the case, instead, the channel state is randomly and independently determined from a set of finite possible states. This model is termed finite-state channel model [62]. Such a finite-state channel model with \( N \) states can be depicted in Fig. 19. We mainly consider this channel model in the following subsections.

![Finite-state channel model with N states.](image)

Fig. 19. Finite-state channel model with N states.

B. Ergodic Capacity

From Fig. 19, we can make the concepts of ergodic capacity more explicit. The ergodic capacity \( \tilde{C} \) should be
used if the random gate selector varies its output selecting index and thereby changes the channel states within one transmission interval of a single block. Also, since this variation is far faster than the transmission interval of a single block, almost all channel states can be experienced for the block, whereas, the outage probability \( C_s \) should be used if the random gate selector maintains its output selecting index during the entire transmission interval. Therefore, only a random channel state will be selected and experienced by the entire block.

Now let us analyze both in details. First, for ergodic capacity, we can assume that the random gate selector outputs selecting index according to a fixed probability density function and the probability of choosing i-th channel state with channel capacity \( C_i \) is a fixed value and denoted by \( \{\delta_i\} \). Obviously, the relation among all \( \{\delta_i\} \), given \( i \in \{1,2,\cdots,N\} \), exists:

\[
\sum_{i=1}^{N} \delta_i = 1 \tag{66}
\]

Therefore, without CSI, the ergodic capacity of each channel can be determined by the average of the channel capacities corresponding to all channel states:

\[
\bar{C} = \sum_{i=1}^{N} \delta_i C_i = \sum_{i=1}^{N} \delta_i \max_{p_{ki}} I(X_i;Y_i) \tag{67}
\]

On the other hand, if CSI is only obtainable at the receiver, then we can determine the ergodic capacity by [8]:

\[
\bar{C}_s = \sum_{i=1}^{N} \delta_i \max_{p_{ki}} I(X_i;Y_i | S_i) \tag{68}
\]

where \( \{S_i\} \) represents the known channel state.

In addition, if CSI is obtainable at both transmitter and receiver, we have [63]:

\[
\bar{C}_s = \sum_{i=1}^{N} \delta_i \max_{p_{ki}} I(X_i;Y_i | S_i) \tag{69}
\]

Taking the Rayleigh fading channel as an example, for i-th state, we can employ the baseband discrete-time equivalent channel model to ease our analysis:

\[
y_i = r_i x_i + n_i \tag{70}
\]

where \( \{x_i\} \) and \( \{y_i\} \) represent the complex input and output when the i-th channel state has been selected; \( \{r_i\} = \{e^{j\phi_i}\} \) characterizes CSI of the i-th channel state and can be simply regarded as a complex i.i.d. random variable with Rayleigh distributed magnitude and uniformly distributed phase; \( \{n_i\} \) is the i.i.d. complex AWGN term of i-th channel state.

Assuming the transmitted power and received power conserve and equal to \( P_t \), then a parametric expression can be utilized to calculate the ergodic capacity [64]:

\[
P_t = \mu e^{-\gamma - \mu\rho(\mu) - 1} + \log_2 \Gamma(\mu) \tag{71}
\]

where \( \mu \) is a parameter connecting \( P \) and \( \bar{C} \) without physical meaning; \( \gamma \) is the Euler’s constant; \( \Gamma(\cdot) \) is the Gamma function, \( \psi(\cdot) = \Gamma'(\cdot)/\Gamma(\cdot) \).

On the other hand, once CSI is detectable at the receiving end, according to (68), the ergodic capacity has a closed form and can be expressed as [8]:

\[
\bar{C}_s = \frac{1}{\ln 2} \sum_{i=1}^{N} \frac{\rho_i}{\rho} e^{-\gamma} dt \tag{72}
\]

Moreover, once CSI is detectable at both transmitting and receiving ends, according to (69), the ergodic capacity has a closed form and can be expressed as [8]:

\[
\bar{C}_s = \frac{1}{\ln 2} \int_0^\infty e^{-\gamma} dt \tag{73}
\]

where \( \rho_0 \) is a parameter which satisfies:

\[
\int_0^\infty \max \left( 0, \frac{1}{\rho_0} - \frac{1}{\rho} \right) e^{-\rho} d\rho = \frac{P_e}{N_0} \tag{74}
\]

C. Outage Capacity

As the definition of outage channel model, once a channel state is selected, it is assumed to be quasi-static during the entire transmission interval of a block. Therefore, due to the randomness of the gate selector, the channel capacity becomes random before the transmission of a block. However, the transmission rate \( R \) is normally constant regardless of the random channel capacity [65]. Therefore, there is a probability that \( R \) would exceed the capacity, in which case a outage of the channel will occur. This probability is termed outage probability and can be mathematically expressed by [8]:

\[
P_{out}(R) = \Pr[C < R] = \int_{R}^{\infty} P_e(C)dC \tag{75}
\]

where \( P_e(C) \) is the PDF of random capacity \( C \).

Theoretically, it is impossible to completely avoid outage as long as a positive transmission rate \( R \) is employed. Therefore, it would be wise to set a tolerable threshold of outage probability \( \varepsilon \in (0,1) \) and accordingly adjust \( R \) to ensure \( P_{out}(R) \leq \varepsilon \). In this case, the \( \varepsilon - \)outage capacity can be mathematically defined as [66]:

\[
C_{\varepsilon} = \max \{R: P_{out}(R) \leq \varepsilon \} \tag{76}
\]

Similar to the derivation procedure of ergodic capacity, we take Rayleigh fading channel as an example to illustrate the details of outage capacity and outage probability. From the analysis above, it is not difficult to find that the randomness of outage capacity is caused by the randomness of the channel attenuation factor \( \gamma \), which is a Rayleigh distributed random variable. Hence, according to (75), we have:
\[ P_{\text{out}}(R) = \Pr \left[ \log \left( 1 + \frac{|r|^2 P_r}{N_0} \right) < R \right] \]

\[ = \Pr \left[ |r|^2 < \left( 1 + \frac{(2^\epsilon - 1)N_0}{P_r} \right) \right] \]

\[ = \Pr \left[ \frac{(2^\epsilon - 1)N_0}{P_r} < |r| < \frac{(2^\epsilon - 1)N_0}{P_r} \right] \]

\[ = \Pr \left[ 0 < |r| < \frac{(2^\epsilon - 1)N_0}{P_r} \right] \]

\[ = \int_{0}^{(2^\epsilon - 1)N_0/P_r} |r|^2 \frac{H_1}{\sigma^2} e^{-2\sigma^2 P_r} \, dr \]

\[ = 1 - e^{-2\sigma^2 P_r} \]

Because of the assumption \( P_r = P_0 = 2\sigma^2 P_r \), we have \( 2\sigma^2 = 1 \). Consequently, (77) can be further reduced to:

\[ P_{\text{out}}(R) = 1 - e^{-2\sigma^2 P_r} \] (78)

Moreover, solving (78) yields the maximum transmission rate \( R_{\text{max}} \), i.e., the outage capacity \( C_{e} \) when the outage probability threshold \( \epsilon \) is given:

\[ C_{e} = R_{\text{max}} = \log \left[ 1 - e^{-2\sigma^2 P_r} \right] \] (79)

In order to enlarge the outage capacity, we can employ diversity techniques. Assuming \( L \) diversity channels are involved and following (33), (40), and (78), we can determine the outage probability in this scenario by [8]:

\[ P_{\text{out}}(R) = 1 - e^{-2\sigma^2 P_r} \sum_{k=1}^{L} \frac{(2^\epsilon - 1)N_0}{P_r} \] (80)

If we let \( P_{\text{out}}(R) = \epsilon \), we can obtain the maximum transmission rate, i.e., the outage capacity for the diversity-based case as well. However, because (80) is a transcendental equation, there is no closed expression of \( C_{e} \) [67]. By some approximation methods, a numerical solution could be provided [68]–[70].

D. Coding Techniques Associated with Fading Channels

As we can see from Section III, diversity techniques are very powerful to overcome the negative effects of fading. From the viewpoint of coding, diversity techniques can be regarded as coding techniques as well, because the original information is encoded into a number of replicas. To further improve the system performance, we might involve more sophisticated codewords to interleave the information bits in time-frequency space [8]. Such a digital communication system considering coding can be depicted in Fig. 20.

VI. CONCLUSIONS

In conclusion, the relevant properties and techniques of fading and fading channels are reviewed in this paper. Specifically, we have analyzed the characterization and classification of fading channels. Then, four types of commonly used fading channels, their unique time-frequency domain properties and applicability are introduced. With simulations, we have also discussed the performance of signals propagating over flat and slow fading channels. By the simulation results, we are able to observe the effects of fading on signal propagation. Moreover, in order to overcome the negative impact of fading, diversity techniques have been expatiated and the communication system with and without diversity are simulated and compared. The feasibility and wide applicability of diversity can be validated by these simulation results.

In addition, several signaling techniques, including wideband signaling, RAKE demodulation and OFDM, are presented with a series of mathematical derivations and explanations. Finally, we have delved into the information-theoretical issues relevant to fading channels and addressed several important concepts to investigate into the essence of fading. Therefore, a number of potential research directions as stated above have been revealed. By reading this paper, the readers are expected to have a broad understanding of fading phenomenon, a variety of types of fading channels, effects of fading on wireless communication and the methodologies to overcome the disadvantageous effects.

Moreover, through the review over a large amount of contents, we would also like to point out a series of potential research directions which are worth further investigating in the future. They are summarized as follows:

- Channel modeling considering pass loss, shadowing and small-scale fading comprehensively.
- Fading channel modeling of multihop systems and multicarrier systems.
- Fading analysis and explanation from the perspective of electromagnetism.
- Design and test of receivers’ prototypes which are able to handle deep fading scenario and recover the signals.
• Fading channel modeling considering location distribution of communication nodes over random networks.
• Fading channel modeling of self-organized networks and considering mutual interference.

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Zhongli Wang graduated from Beihua University and University of Science and Technology Beijing for the majors of Communication Engineering and Control Theory & Control Engineering, respectively. He is currently an associate professor with College of Electrical and Information Engineering, Beihua University and dedicates to the investigations of pattern recognition and embedded systems. He has tens of papers published in international journals and proceedings and wrote three textbooks. He also participated in two research projects sponsored by National Natural Science Foundation of China, four research projects sponsored by the local municipal government, and three research projects sponsored by his university.

Guo Sheng graduated from Zhejiang University, is the associate professor with Zhejiang Post and Telecommunication College. He has published several papers related to the research into electronics and communication engineering. His current research interests include electronics, wireless communication and signal processing.

Xu Zhang received the BS degree from Department of Mechanical Engineering and Materials Science, University of Shanghai for Science and Technology, Shanghai, China. He is now a fourth year PhD candidate at Department of MEMS, Duke University. His research interests include artificial intelligence, neural networks, neurorobotics.