Molecular communication noise effect analysis of diffusion based channel for considering msk and mosk modulations

Abstract

One of the unaddressed and open challenges in the nano-networking is the characteristics of noise. The previous analysis, however, have concentrated on end-to-end communication model with no separate modelings for propagation channel and noise. By considering a separate signal propagation and noise model, the design and implementation of an optimum receiver will be much easier. In this paper we justify consideration of a separate additive Gaussian noise model of a nano-communication system based on the molecular communication channel for which are applicable for Msk and Mosk modulation schemes. The presented noise analysis is based on the Brownian motion process and advection molecular statistics, where the received random signal has a probability density function whose mean is equal to the mean number of the received molecules. Finally, the justification of received signal magnitude being uncorrelated with additive non-stationary white noise, is provided.

Keywords: molecular, noise, diffusion, channel, frequency shift keying

Introduction

Any molecular communication model similar to conventional digital communication system consist of three main components, they are the transmitter, receiver, and the channel. Figure 1 illustrates a schematic of a molecular communication model. In comparison with the digital communication system model, molecular communication system can also be implemented using different types of modulation schemes. The modulation techniques mainly used in digital communications\(^1\) Pulse Amplitude Modulation (PAM), Phase Shift Keying (PSK), and Frequency Shift Keying (FSK). In PAM, different amplitudes are used for message modulation. In PSK, to modulate different messages finite number of phases are used, and in FSK, finite number of frequencies are used for modulating the messages.

Figure 1 A schematic of a molecular communication system.

Considering nano-communication transmission and reception environment as shown in Figure 2 in general, transmitters can employ three modulation techniques:

Figure 2 Block diagram of a molecular communication system.
Concentration Shift Keying (CSK), which is similar to PAM. In this type of modulation, only one type of molecule is used and messages are encoded by different amounts of released molecules.\textsuperscript{6,7} Pulse Position Modulation (PPM), where only one type of molecule is used and messages are encoded by transmitting single pulse in different time shifts.\textsuperscript{6,7}

Molecular Shift Keying (MOSK), in which messages are encoded by using different type of molecules.\textsuperscript{6,7} To avoid interference the signal space will be chosen to be orthogonal. In\textsuperscript{8} it is shown that this type of modulation in comparison to CSK. It is more robust against noise.

The modeling and characterization of noise in a conventional communication is well defined and very broadly analyzed. But in contrast to this, in the molecular communications characterization of noise still one of the open challenges for researchers. Even though, noise sources of molecular communications are analyzed in terms of stochastic process, but most of the works have concentrated on a general end-to-end communication model with considering no separate model for the propagation channel and noise. It is then believed that, by having a separate signal propagation and noise model, if one wants to follow, the design of an optimum receiver will be much easier task.

Noise modeling in diffusion-based molecular communication

Consider the diffusion-based noisy channel of Figure 3; where at time zero, \( Q \) number of messenger molecules are propagated through the diffusion channel of impulse response \( h(t) \). Let \( n(t) \) be the sample time zero, \( Q \) number of messenger molecules are propagated through the sensing area of the receiver has Binomial distribution with mean \( (x(t),y(t)) \) and variance \( 2Dt \), with \( D \) being the diffusion coefficient. Hence the probability of having a messenger molecule in the sensing surface area of \( A \), is given by:

\[
P_e(x(t),y(t)) = \frac{1}{4\pi Dt} e^{-\frac{(x(t)-x_0)^2+(y(t)-y_0)^2}{4Dt}}
\]

with mean \( (x^*(t),y^*(t)) \) and variance \( 2Dt \), with \( D \) being the diffusion coefficient. Hence the probability of having a messenger molecule in the sensing surface area of \( A \), is given by:

\[
P_e(x(t),y(t)) = \frac{1}{4\pi Dt} \exp(-\frac{(x(t)-x_0)^2+(y(t)-y_0)^2}{4Dt})
\]

Let us now, assume that \( Q \) messenger molecules are emitted by the transmitter of the nano-system, then the reception of \( K \) messenger molecules in the sensing area of the receiver has Binomial distribution with probability mass function given by:

\[
P(k) = \binom{Q}{k} p^k (1-p)^{Q-k}
\]

As we know, if we choose \( Q \) large and \( p \) small enough, then the Eq.(6), can be approximated by a Gaussian distribution \( N(\mu, \sigma^2) \) where \( \mu = Qp \) and \( \sigma^2 = Qp(1-p) \). By combining of the Eqs.(5), we get the result

\[
\mu = Qp = Q \frac{1}{4\pi Dt} \int_A \int_B \exp(-\frac{(x(t)-x_0)^2+(y(t)-y_0)^2}{4Dt})dxdy
\]

Figure 3 Diffusion-based molecular channel model.

Figure 4 Model of the two separate noise components.

It is worth to know that, \( n(t) \), the Brownian noise is associated with the random thermal motion of the messenger molecules and \( n(t) \), the residual noise, is as a result of the interference due to the messenger molecules. We also should remind that \( n(t) \) is analogous to the inter-symbol interferences in the conventional communications. Now, considering \( z(t) \) being the input to the receiver or the output of the channel, we might be able to write

\[
Z(t) = s(t) + n(t) + n(t)
\]

Where \( s(t) \) is the mean number of the received messenger molecules.

Brownian noise in diffusion-based molecular channel

Since Brownian noise is produced by the random thermal motion of the transmitted messenger molecules, then the number of counted molecules at the receiver are random. Let \( X(t) = (x(t),y(t),z(t)) \) be a three dimensional position of a messenger molecule at time, \( t \), since the Brownian motion is modeled by the Wiener process, for a two dimensional reception surface, we may write,

\[
x(t) = x + \sqrt{2Dt} u(t)
\]

\[
y(t) = y + \sqrt{2Dt} u(t)
\]

where \( (x,y) \) being the original position of the transmitted messenger molecule with \( u(t) \) having a standard normal distribution, \( u(t) \sim N(0,1) \). Therefore, for, any fixed \( t=t_0 \), the mean or expected values of \( x(t) \) and \( y(t) \) are given respectively by;

\[
E(x(t)) = x + \sqrt{2Dt} E(u(t)) = x
\]

\[
E(y(t)) = y + \sqrt{2Dt} E(u(t)) = y
\]

and as a result, for each emitted molecule, the joint probability density function of being at point \( (x(t),y(t)) \), \( P_{xy}(x(t),y(t)) \) is a two dimensional Gaussian distribution function, where

\[
P_{xy}(x(t),y(t)) = \frac{1}{4\pi Dt} \exp\left(-\frac{(x(t)-x_0)^2+(y(t)-y_0)^2}{4Dt}\right)
\]

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Comparing the mean value \( u \) in the Eq.(7) with the Eq.(5), it is not hard to see that these two quantities are the same, that is \( u \) is the same as the mean number of received messenger molecules. This implies that the received random signal has a probability density function whose mean is the deterministic signal of the Eq.(5). Thus this equality justifies the facts that first, Brownian noise is additive, second signal in Eq.(5) is the response of the molecular channel. Let \( v(t) \) be the number of the messenger molecules corrupted by Brownian noise, as shown in Figure 5, to be defined as follows:

\[
v(t) = s(t) + n_b(t)
\]

Figure 5 Diffusion-based molecular channel model.

then, \( v(t) \) has a Gaussian distribution of mean \( s(t) \) and variance \( QP(1-P) \), that is \( N(s(t),Q(1-P)) \). Now if we assume \( p<<1 \), then we can say that 1-p~1 and only then the \( \text{var}(n_b(t)) = QP \). We also know that \( QP = E(s(t)) \) and thus \( n_b(t) = (s(t))N(0,1) \), which implies that Brownian noise follows a zero mean Gaussian process with variance equal to the mean number of received molecules in the sensing area. To investigate the statistical properties of the Brownian noise we want to compute the autocorrelation function of the Brownian noise, that is \( R_{nn}(t_1,t_2) \). Since the probability density function of Brownian noise is derived through the probability density function of \( v(t) \), let's compute the second order density function of \( v(t) \). Using the chain rule of probability,

\[
P_v(v_1,v_2) = P_v(v_1)P_v(v_2) = P_v(v_1)v_1P_v(v_2)
\]

with \( v_1 = v(t_i) \) for \( i = 1,2 \). According to Eq.(6), \( v(t) \) is a binomial random variable with parameters \( Q,P \), \( B(Q,P) \), and \( p \), being the probability of having \( v(t) \) molecules at the receiving area at time \( t_i \). The same argument is true for \( v(t) = v(t) \). But \( P(V_1|V_1) \), the probability of having \( v(t) \) molecules at the receiving area at time \( t_i \) given that \( v(t) = v(t) \), molecules have been transmitted at time \( t_i \), where \( t_i \leq t \), we then can write.

\[
P_{v_1}(v_1) = 2 \frac{(Q-P)^{v(t)}}{v(t)} \quad \text{and} \quad P_{v_2}(v_2) = \frac{(Q-P)^{v(t)}}{v(t)}
\]

where \( P_{v_1}(v_1) \) is the number of \( v_1 \) molecules staying in the sensing area and \( P_{v_2}(v_2) \) is the number of \( v_2 \) molecules staying in the sensing area.

\[
P_v(v_1,v_2) = \frac{1}{v_1!v_2!} \frac{(Q-P)^{v(t)}}{v(t)}
\]

which is a binomial probability function, with \( P_{v_1}(v_1) \) being the probability of having a messenger molecule at the sensing area at time \( t_i \), while \( v(t) \) is the number of \( v(t) \) molecules staying in the sensing area. Thus we have,

\[
P_v(v_1,v_2) = \frac{1}{v_1!v_2!} \frac{(Q-P)^{v(t)}}{v(t)}
\]

By assuming the remaining \( i \) molecules in the sensing area is negligible, that is \( i=0 \), the Eq.(7) will become

\[
P(v_2|v_1) = \frac{v_1}{v_2} \frac{(1-P)^{v_2}}{v_2} P(v_2)
\]

with \( v(t) \) being white Gaussian with mean zero and variance one, \( N(0,1) \). Now if we let the autocorrelation function of the Brownian noise, \( n_b(t) \), be \( R_{nn}(t_1,t_2) \), then

\[
R_{nn}(t_1,t_2) = E(n_b(t_1)n_b(t_2)) = \delta(t_1-t_2) \quad \text{for which we conclude that} \quad v(t) \quad \text{is a non-stationary noise.}
\]

Residual noise in diffusion-based molecular channel

The chemical receptors may contain residual molecules on the surface of the nano machine structure from the previous reception the same time there may be molecules from the current communication symbol, thus distortion might be developed. In the conventional communications, the interference caused by the previous signaling intervals is usually called inter-symbol interference and generally compensated by the receiver using, matched filter or error correcting codes or other means. To handle the so called inter-symbol interference problem in the molecular communication, we treat this as a corrupting noise source or residual noise, \( n(t) \). Thus residual noise, \( n(t) \), accounts for interference from messenger molecules that are remained in the receiver sensing area from the last transmission. Now to investigate the statistical characteristics of this noise, we need to drive the probability density function of a messenger molecule from the previous transmission time, \( nT \), in the receiver at time \( t \). To do that, we use the fact that according to the Brownian motion process, the displacement of a molecule during an infinitesimal interval of dt seconds, can be modeled by a zero mean Gaussian distribution process with variance \( 2Dt \).

\[
X(t+dt) = X(t) + \sqrt{2Dt} u(t)
\]

where \( D \) being the diffusion coefficient and \( X(t) = (x(t), y(t), z(t)) \) being the three dimensional molecular position at time \( t \), and \( u(t) \) representing a vector of multivariate normal distribution. Thus, using Eq.(12) for the displacement of a messenger molecule transmitted at time \( nT \), is

\[
X(t) = X_n + \sqrt{2D}(t-nT) \quad u(t)
\]

where \( X_n \) is the transmission location \( N(O_{12}, I_{12}) \), \( O_{12} \) and \( I_{12} \) are two-dimensional null and identity matrices. Hence, the probability function of being at location \( X(t) \) we have;

\[
P_s(x) = \frac{1}{4\pi D(t-nT)} \exp\left(-\frac{|X-X_n|^2}{4D(t-nT)}\right)
\]

and therefore the probability of a messenger molecule being in the
sensing area $A_r$ is

$$P^{(n)}_v(t) = \int \int \int \int \frac{1}{4\pi D(t-nT_r)} \exp\left(-\frac{|X-X_m|^2}{4D(t-nT_r)}\right) dx dy$$  \hspace{1cm} (21)

It is also true that probability distribution of $K$ molecules from $Q$ molecules being at the reception area is a binomial distribution with mean $QP^{(\alpha)}$ and the variance of $QP^{(\alpha)}(1-P^{(\alpha)})$. Since $Q>1$ which is obvious, then for small $Q$ this binomial distribution can be approximated by a Gaussian distribution function of $N(QP^{(\alpha)}, QP^{(\alpha)}(1-P^{(\alpha)}))$. So, the probability density function of the process $v(t)=s(t) - t$ is approximated by a Gaussian distribution function of $N(QP^{(\alpha)}, QP^{(\alpha)}(1-P^{(\alpha)}))$. Hence, the probability distribution of the residual noise is as follows:

$$P^{(n)}_v(t) = \sum_{n=0}^{[\frac{Q}{2}]-1} \frac{Q^n}{n!} \frac{1}{4\pi D(t-nT_r)} \exp\left(-\frac{|X-X_m|^2}{4D(t-nT_r)}\right)dx dy$$  \hspace{1cm} (22)

Finally, the overall number of residual molecules has distribution of Gaussian as well, $\eta \sim N(\eta, \beta)$ where

$$\eta = \sum_{n=0}^{[\frac{Q}{2}]-1} \frac{Q^n}{n!} \frac{1}{4\pi D(t-nT_r)} \exp\left(-\frac{|X-X_m|^2}{4D(t-nT_r)}\right)dx dy$$ \hspace{1cm} (23)

Conclusion

In this paper, additive noise model is considered only in one scenario, that is diffusion-based. In this case, two additive noise sources are considered; as a result of thermal motions of molecules Brownian noise and, as a result of residual molecules from the previous transmission, residual noise. It is shown that the corrupted noise in the diffusion-based channel must be additive and is Gaussian with having uncorrelated and non-stationary statistics with dependence on the signal magnitude. For the future works, the investigation of the electromagnetic and electrostatic effects on the diffusion-based channel is very interesting. Finally, analysis of the channel capacity and pulse-shaping principals are very important subjects which must be addressed with great care.

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Conflicts of interest

The author declares there is no conflicts of interest.

References


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