Amplify and Forward Relaying In Cognitive Radio Systems Under Rayleigh Fading Conditions

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Abstract- Cognitive Radio Systems have emerged as one of the key enablers in high capacity and high data rate networks. As wireless media are highly random in nature, therefore sensing the spectrum and using the Channel State Information (CSI) to suppress the frequency tones leading to low Signal to Noise Ratio (SNR) is crucial to improve upon the BER performance of the system. Such systems can be efficiently modelled as a Software Defined Radio (SDR). In this paper, a multi hop cooperative relaying technique is proposed and a subsequent BER analysis follows. The model follows the regulations laid down by software defined radio (SDR) and is shown to have low bit error performance.

Keywords- Software Defined Radio (SDR), Cognitive Network, Channel State Information (CSI), Bit Error Rate (BER).

I. INTRODUCTION

The electromagnetic spectrum is a characteristic asset. The present spectrum authorizing plan is not able to oblige quickly growing demand in wireless communication due to the static spectrum allocation strategies. This allocation prompts increment in spectrum scarcity issue. Cognitive radio (CR) technology is a propelled remote radio design which aims to expand spectrum utilization by distinguishing unused and under-used spectrum in rapidly evolving environments. Spectrum sensing is one of the key strategies for cognitive radio which detects the presence of primary client in authorized licensed frequency band utilizing dynamic spectrum assignment policies to utilize unused spectrum. In many areas cognitive radio frameworks coexist with other radio frameworks, utilizing the same spectrum yet without creating undue interference.[1],[3] The most simple and easy way to implement sensing technique is energy detection.

With Cognitive Radio being utilized as a part of various applications, the territory of spectrum sensing has become progressively vital. As Cognitive Radio technology is being utilized to provide a method for utilizing the spectrum all the more productively, spectrum sensing is key to this application. The ability of Cognitive Radio frameworks to get

to spare sections of the radio spectrum, and to continue observing the spectrum to guarantee that the Cognitive Radio framework does not create any undue interference depends totally on the spectrum sensing components of the framework.[4]

Advantages of Cognitive Radio

- Mitigate and solving spectrum access issues and Spectrum utilization improves.
- Improves wireless networks performance through increased user throughput and system reliability.
- More adaptability and less co-ordination.

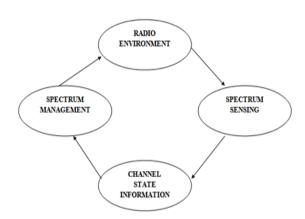


Fig.1 A Typical Cognitive Radio Environment.

For the overall framework to work viable and to provide the required change in spectrum efficiency, the Cognitive Radio spectrum sensing framework must have the capacity to adequately recognize some other transmissions, distinguish what they are and inform the central preparing unit inside the Cognitive Radio so that the required actions can be taken.

II. SPECTRUM SENSING IN SOFTWARE DEFINED RADIO (SDR)

Software defined radio (SDR) is a term typically used for the design of systems characterized by their channel state

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information. The principal step of spectrum sensing is that it decides the presence of primary user on a band. The cognitive radio has the capacity to impart the result of its detection with other cognitive radios in the wake of sensing the spectrum. The main objective of spectrum sensing is to discover the spectrum status and activity by periodically sensing the target frequency band. The cognitive radio system examines all level of flexibility (time, frequency and space) to predict spectrum usage. There are a few procedures available for spectrum sensing. Spectrum sensing is a system which figures out if a given frequency band is utilized. A wide range of routines are proposed to recognize the presence of signal transmission and can be utilized to improve the detection probability.

The aim of the spectrum sensing is to decide between two hypotheses which are

$$x(t) = w(t)$$
, H0 (Primary User absent)

$$x(t) = h * n(t) + w(t)$$
, (Primary User
H1 present)

Where x(t) is the signal received by the CR user, n(t) is the transmitted signal of the primary user , w(t) is the AWGN band, h is the amplitude gain of the channel. H0 is a null hypothesis, which states that there is no licensed user signal. Energy Detection is a simple detection method. The energy detection is said to be a blind signal detector in light of the fact that it overlooks the structure of the signal. Energy detection is based on the rule that, at the receiving end, the energy of the signal to be detected is computed. It estimates the presence of a signal by comparing the energy received and a known threshold λ derived from the statistics of the noise. If the frequency response of the system is available, then under such a condition, we possess the Channel State Information (CSI).

III. CHALLENGES IN UTILIZATION OF CSI FOR COGNITIVE RADIO SYSTEMS

Problem of BER: A major challenge for any wireless communication system is the frequency selective nature of wireless channels. For multi carrier modulation schemes, frequency selectivity poses a serious problem since different sub carriers are treated by the channel differently. Carriers which belong to the frequency range where the channel gain reduces drastically, undergo heavy attenuation. The attenuation varies with the frequency response of the channel. The result of this phenomenon is variable sub carrier gain or sub carrier strength. While some sub carriers may have satisfactory gain, some may have average gain while others may have extremely low sub carrier gain. The sub carriers

with low sub carrier gain or strength tend to adversely affect the Bit Error Rate of the system, since low strength of the carriers would imply low signal to noise ratio (SNR). This eventually introduces errors and results in non-reliable communication. Therefore it is important to suppress sub carriers with low gain or strength to improve the BER performance of the system. During the fades user experiences a signal outage. Outage probability determines the amount of signal strength which is less below the minimum noise power. So for proper reception outage probability should be as lowas possible. Thus it is important to improve the signal to noise ratio (SNR) for improving system performance.

IV. MATHEMATICAL MODELLING FOR BER AND OUTAGE OF A COGNITIVE RADIO SYSTEM

Outage: Let a cellular system is indicated for $\Phi = \{C, D\}$, C is cellular communication and D is D2D communication. It is also assumed that the distribution of device nodes of system follows a stationary Poisson point process (PPP) with density of λ_D in the finite two-dimensional plane. The channel model includes path loss and Rayleigh fading. Therefore, the power of the node i received from j can be expressed as: $P_i \delta_{ij} |X_{ij}|^{-\eta}$, $i, j \in \Phi$, P_{iis} the power of node i, δ_{ij} is the Rayleigh fading indexbetween i and j, and it has an exponential distribution with unitmean, X_{ij} is the distance between node i and j, η is the path loss exponent. Signal to interference plus noise ratio at the receiver k is given as:

$$SINR_{k} = \frac{P_{k} \delta_{k0} R_{k}^{-\eta}}{\sum_{j \in \phi} \sum_{X_{ij} \in \pi_{i}} P_{j} \delta_{ji} |X_{ji}|^{-\eta} + N_{0}}$$

where δ_{ji} is the fading factor on the power transmitted from the desired transmitter to the receiver, N_0 is the thermal noise power, spectrum sharing systems are interference limited systems, so the thermal noise is negligible in regimes of interest. So the SIR can be used instead of SINR, which allows simplification to:

$$SIR_{\mathbf{k}} = \frac{\delta_{\mathbf{k}0}R_{\mathbf{k}}^{-\eta}}{I_{\mathbf{k}}}$$

$$\begin{split} I_{\mathbf{k}} &= \sum_{j \in \varphi} I_{\mathbf{k} j} \text{ Where} \quad I_{\mathbf{k} j} \text{ is the sum of the power} \\ \text{normalized interference from transmitting nodes of system } j \text{ to} \\ \text{the receiver of system } k, \text{ which is defined} \\ I_{\mathbf{k} j} &= \left(\frac{\mathbf{p}_{i}}{\mathbf{p}_{\mathbf{k}}}\right) \sum_{X \mid J \in \pi_{j}} \delta_{ji} \left|X_{ji}\right|^{-\eta}, \text{ for QOS of system } k \text{ (D2D or cellular users)}, \text{ SIR required to meet a certain threshold:} \end{split}$$

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 $SIR_{k \ge V_k}$, where V_k is target SIR, then the probability of successful transmission can be defined as:

$$\begin{split} &P(SIR_k > V_k) = P\big(\delta_{k0} \geq V_k R_k^{\eta} I_k\big) \\ &= \int_0^{\infty} P\big\{\delta_{k0} \geq V_k R_k^{\eta} I_k\big\} f_{Ik}(I) \ dI \\ &= \psi_{Ik}\big(V_k R_k^{\eta}\big) \end{split}$$

 δ_{k0Is} exponentially distributed, then $\psi_{Ik}(V_k R_k^{\eta})$ is the Laplace transform of $f_{Ik}(I)$, which can expressed for $E[e^{-I_k\eta}]$.

 $I_{\mathbf{k}} = \sum_{j \in \Phi} I_{\mathbf{k}j}$, the interference $I_{\mathbf{k}j}$ is independent to each other,

$$E[e^{-l_k\eta}] = \prod_{j \in \phi} E[e^{-l_k\eta}]$$

then the probability of successful transmission can be changed to

$$\left(V_{\mathbf{k}}R_{\mathbf{k}}^{\eta}\right) = \prod_{i \in \Phi} \psi_{I\mathbf{k}j} \left(V_{\mathbf{k}}R_{\mathbf{k}}^{\eta}\right)$$

Especially, when the interferingnodes are Poisson pointdistributed:

$$\begin{split} &\psi_{Ij} \big(V_k R_k^{\eta} \big)_{=\mathrm{ex}} p \left\{ -\lambda_j \int_{R^2}^1 1 \, \mathrm{E} \left[\mathrm{e}^{-V_k R_k^{\eta} \delta |X|^{-\eta}} \right] \mathrm{d}x \right\} \\ &= \exp \left\{ -\lambda_j \int_{R^2}^1 \frac{V_k R_k^{\eta} \delta |X|^{-\eta}}{1 + V_k R_k^{\eta} \delta |X|^{-\eta}} \mathrm{d}x \right\} \\ &= \exp \left\{ -C_\alpha R_k^2 V_k^{2/\eta} \gamma_{kj} \lambda_j \right\} \\ &\gamma_{kj} = \left(\frac{p_j}{p_k} \right)^{2/\eta} \quad \text{Is power ratio, where } P_j \text{ and } P_k \text{ are power of system j and k,} \end{split}$$

$$\begin{aligned} \mathbf{C}_{\eta} &= \left(\frac{2\pi}{\eta}\right) \Gamma\left(\frac{2}{\eta}\right) \Gamma\left(1 - \frac{2}{\eta}\right), & \text{where} \\ \Gamma(x) &= \int_{0}^{\infty} y^{x-1} e^{-y} \, dy & \text{is a gamma function.} \end{aligned}$$

Therefore, success transmission probability of system k is:

$$P(\text{SIR}_k \ge V_k) = \exp\left\{-K_k \sum_{j \in \varphi} \gamma_{kj} \lambda_j\right\}$$

here, $K_{\mathbf{k}} = C_{\mathbf{k}} R_{\mathbf{k}}^2 V_{\mathbf{k}}^{2/\eta}$

Thus the outage probability of D2D receiver is

$$q(\lambda) = \exp \left\{ -\frac{2\pi^2}{\eta \sin\left(\frac{2\pi}{\eta}\right)} R_{\mathbf{k}}^2 V_{\mathbf{k}}^{2/\eta} \lambda \right\}$$

BER: We know that probability of error is given as:

$$P_e = P(0)P(1/0) + P(1)P(0/1)$$

Provided that system is assumed to be affected only by white noise than

$$P(^{1}/_{0}) = P(^{0}/_{1})$$
So, $P_{e} = P(^{1}/_{0})[P(0) + P(1)]_{(4.27)}$

$$P_{e} = P(^{1}/_{0})_{Or}P(^{0}/_{1})$$

$$P_e = \int_{v_{\rm th}}^{\infty} f(^z/_0) dz$$

$$=\int_{v_{th}}^{\infty}\frac{1}{\sqrt{2\pi\sigma_{n_0}^2}}e^{\left(\frac{-(z-a_z)^2}{\sigma_{n_0}^2}\right)}dz$$

Assume $\sigma_{n_0} = y_z - a_z = y_z - \sigma_{n_0}$

Differentiating above equation we get,

$$d_z = \sigma_{n_0} d_y$$

For $z = \infty$, $y \to \infty$ and $z = v_{th}$, $y = (v_{th} - a_2) / \sigma_{n_0}$

For
$$z = \omega_1$$
, $y = \omega_2$ and $z = v$ th, $y = (v)$

$$y = \frac{a_1 + a_2}{2\sigma_{n_0}} - a_2$$

$$= \frac{a_1 - a_2}{2\sigma_{n_0}}$$

We know that Q-function,

$$Q(x) = \frac{1}{\sqrt{2\pi}} \int_{x}^{\infty} e^{\frac{-y^2}{2}} dy$$

Where y = a dummy variable.

Thus, probability of error can be expressed as

$$P_{e} = \frac{1}{\sqrt{2\pi\sigma_{n_{0}}^{2}}} \int_{\frac{a_{1}-a_{2}}{2\sigma_{n_{0}}}}^{\infty} e^{\left(-y^{2}/_{2}\right)} \sigma_{n_{0}} dy$$

In terms of Q-function, probability of error can be expressed as

$$P_{e} = Q \left[\frac{a_1 - a_2}{2\sigma_{n_0}} \right]$$

In case of matched filter, Pe is given as

$$P_{e} = Q \left[\sqrt{\frac{(a_{1} - a_{2})^{2}}{4\sigma_{n_{0}}^{2}}} \right]$$

When more than one signal is transferred through the matched filter then h (t) = $x(\tau-t)$ where $x(t) = x_1$ (t) x_2 (t) $x_1 = x_2$

then $\frac{\sigma_{n_0}^2}{\sigma_n^2}$ corresponds to its maximum possible of $\frac{N_0}{2}$, so that $\frac{P_0}{2}$ corresponds to its minimum possible.

$$P_{e} = Q \left[\sqrt{\frac{1}{4} \frac{E_{d}}{\left(\frac{N_{0}}{2}\right)}} \right]$$

Where EDis energy of x (t). Thus in terms of signal power to

noise power,
$$P_e$$
 can be expressed as
$$P_e = Q \left[\sqrt{\frac{s}{N}} \right]$$
 Where $S = \text{signal power} = E_D/4$, $N = \text{noise power} = N_0 \times B$ and width

V. RESULTS

In the simulation, its assumed that a the primary user (PU) does not transmit energy and the signal at the receiver is due to the noise. In case the receiver detects an energy level above the energy threshold, and adjudges it as a signal transmitted by the primary user, then it counts for a false alarm. The probability for detecting a false user (False Alarm) decreases with the increase in the Energy threshold for Energy detection. The following graphs illustrate the concept. The subsequent section illustrates the various steps employed in the proposed system.

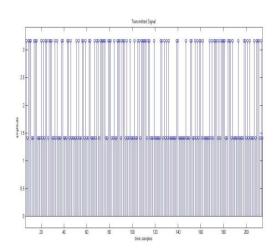


Fig.2 Binary Data of Sender

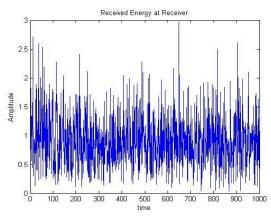


Fig.3 Energy variation at receiving end

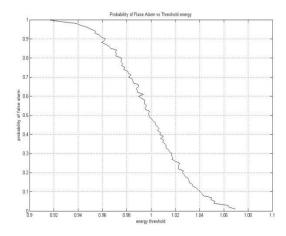


Fig.4 Probability of False alarm w.r.t. Practical Energy
Threshold

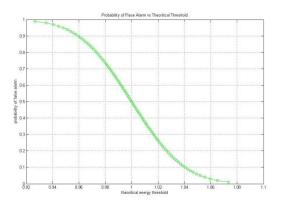


Fig.5 Probability of False alarm w.r.t. Theoretical Energy
Threshold

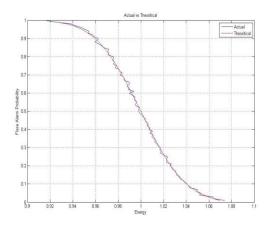


Fig.6 Comparative Probability of False Alarm for theoretical and practical cases

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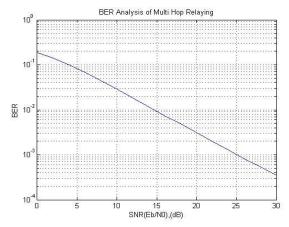


Fig.7 BER Analysis of Conventional Relaying

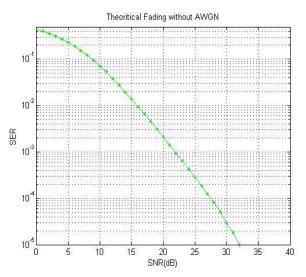


Fig.8 BER Analysis without Fading Conditions

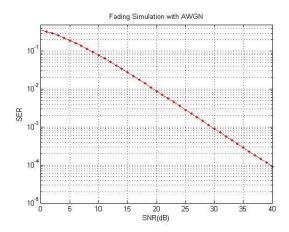


Fig.9 BER Analysis under Fading Conditions

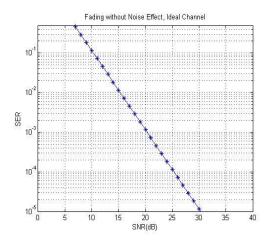


Fig.10 Fading under Ideal Zero Noise Conditions

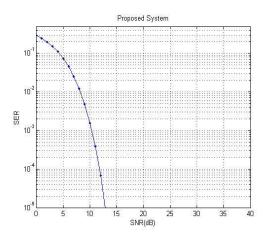


Fig.11 BER Analysis of Proposed System

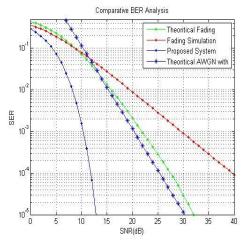


Fig.11 Comparative BER Analysis

VI. CONCLUSION

Here we have presented an algorithm that uses multi hop relaying in cognitive radio systems defined by software defined radio (SDR). The simulations have been run for different conditions. It has been shown that the proposed

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system achieves better BER performance compared to previously existing techniques and conventional systems.

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