

SMART TRANSPORTATION FOR NEXT GENERATION  
WIRELESS COMMUNICATION SYSTEM

by

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## Abstract

There have been an unprecedented increase in demand for high data rates and mobility required by new wireless applications which has led to intensive research on fifth generation (5G) wireless communication system. By 2020, 5G is projected to be employed worldwide supporting the connectivity of up to 20 billion devices and will be crucial in the success of vehicular networking and internet of things (IoT). It is also believed that 5G systems would be capable of providing significant improvements in cell capacity and will support high data rates up to 5 and 50 Gb/s for high-mobility and pedestrian users, respectively. However, to achieve such data rates for high-mobility scenarios there are still many challenges for wireless system engineers.

In this thesis, we propose two test-beds namely Heterogeneous Cooperative Spectrum Sensing (CSS) and Long Term Evolution for Railways (LTE-R) performance analysis test-bed to further advance the 5G system development. Heterogeneous CSS test-bed is implemented using Software-Defined Radios (SDRs) with different radio characteristics. We used both soft and hard data fusion schemes to compare the signal source detection performance in real-time fading scenario. For 5G technologies, the most effective solution is to use the underutilized spectrum as a secondary user via dynamic spectrum access (DSA). It is very challenging to get an accurate estimate of incumbent users with a single-sensor system under a practical fading environment. Various non-idealities such as shadowing, multipath and fluctuating noise variance can make it difficult to detect the primary user. Cooperative spectrum sensing can mitigate the effects of multipath and shadowing by utilizing the spatial and temporal diversity of a multiple radio network. LTE-R test-bed analyze the performance of high speed trains (HSTs) in a tunnel environment and can be used to test 5G systems for high mobility scenario in drastically impaired channels.

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# Contents

<b>List of Figures</b>	<b>5</b>
<b>List of Tables</b>	<b>7</b>
<b>1 Introduction</b>	<b>8</b>
1.1 Motivation . . . . .	8
1.2 State of the Art . . . . .	10
1.3 Thesis Contributions . . . . .	12
1.4 Thesis Organization . . . . .	12
<b>2 Smart Railway Communication System</b>	<b>14</b>
2.1 Positive Train Control . . . . .	15
2.2 Wireless Broadband (WiBro) . . . . .	15
2.3 T2T and T2G Communication System . . . . .	15
2.4 LTE-R Communication System . . . . .	15
2.5 Spectrum Regulation for LTE-R . . . . .	18
2.6 Vehicular Communication Using LTE . . . . .	20
2.6.1 Internet of Things (IoT) for Railways . . . . .	21
2.7 Leaky Coaxial Cable for LTE-R . . . . .	24
<b>3 Heterogeneous Cooperative Spectrum Sensing (CSS)</b>	<b>26</b>
3.1 Spectrum Sensing . . . . .	26
3.1.1 Energy Detection . . . . .	27
3.1.2 Cyclostationary Method . . . . .	29
3.1.3 Matched Filter Detection . . . . .	29
3.1.4 Wavelet Detection . . . . .	29
3.2 Cognitive Radios . . . . .	29
3.3 CSS in Heterogeneous Networks . . . . .	30
3.4 Software Defined Radios . . . . .	32
3.4.1 USRP N210 and RTL-SDR . . . . .	33
3.4.2 GNU Radio and Software-Defined Radio . . . . .	34
3.4.3 MATLAB . . . . .	36
3.5 Summary . . . . .	36

<b>4 Proposed LTE-R Test-bed Implementation and Results</b>	<b>37</b>
4.1 LTE-R Testbed in Matlab . . . . .	37
4.2 HST LTE-R in a Tunnel . . . . .	39
4.2.1 K-factor in a Tunnel . . . . .	40
4.2.2 BER Performance . . . . .	40
4.3 Channel Impairements Inside a Tunnel . . . . .	42
4.3.1 Multipath Fading . . . . .	43
4.3.2 Doppler Shift . . . . .	43
4.4 Proposed Channel Model . . . . .	45
4.5 Summary . . . . .	46
<b>5 CSS Implementation and Results</b>	<b>48</b>
5.1 Overview . . . . .	48
5.2 Experimental Setup for Heterogeneous CSS . . . . .	49
5.2.1 Sensor Node and Signal Source Setup . . . . .	51
5.3 Hard-Data Fusion Scheme . . . . .	53
5.4 Soft-Data Fusion Scheme . . . . .	53
5.4.1 Maximum Normalized Energy Scheme . . . . .	54
5.4.2 Equal Gain Combining Scheme . . . . .	55
5.5 Heterogeneous CSS Results . . . . .	56
5.5.1 Hard Decision Combining . . . . .	57
5.5.2 Soft Decision Combining . . . . .	57
5.6 Summary . . . . .	58
<b>6 Conclusions</b>	<b>61</b>
6.1 Research Outcomes . . . . .	61
6.2 Future Work . . . . .	62
<b>Bibliography</b>	<b>63</b>
<b>A Heterogeneous Cooperative Spectrum Sensing Code</b>	<b>71</b>
A.1 harddecisionpdroc.m . . . . .	71
A.2 softharddecisionpd.m . . . . .	73
A.3 spectrumsenseusrp.py . . . . .	75
A.4 gnuradiortlsdrsense.py . . . . .	83
<b>B LTE-R Analysis Code</b>	<b>89</b>
B.1 kfactordist.m . . . . .	89
B.2 bercalculation.m . . . . .	90

# List of Figures

1.1	Heterogeneous sensor network employing cooperative spectrum sensing. $RFFE_i$ and $SR_i$ represents different front end and sampling rates for the SDR units.	9
1.2	High speed train inside a tunnel for LTE-R. $D_{LOS}$ is the distance between transmitter and receiver, $d$ is the distance between the LCX cable transmission slots.	11
2.1	Proposed LTE-R Architecture for next generation High-speed Railways consisting of EPA and E-UTRAN.	16
2.2	Machine-to-Machine communication paradigm in 5G Cellular System.	18
2.3	Vehicular communication using D2D 5G cellular communication technology.	21
2.4	Machine-to-Machine communication paradigm in 5G Cellular System.	22
2.5	Leaky Coaxial Cable	25
3.1	The block diagram explaining the basic parts of Cognitive Radio system. The operating parameters are configured based on the characterization of the wireless environment.	30
3.2	Software defined radio pushes all the adaptive elements and data manipulation operation into software. The goal of SDR is to provide or define all of the radio operation in software.	33
3.3	GNURadio and Software-Defined Radio	35
4.1	Block diagram of a communication system through a HST channel using QPSK, 16-QAM and 64-QAM.	38
4.2	Received LTE-R OFDM signal under HST Ricean Fading Environment.	39
4.3	K-factor versus $D_{LOS}$ for different center frequencies $f_c = 2, 3$ and $5$ GHz.	40
4.4	Comparison of $E_b/N_0$ verus BER for LTE-R OFDM modulation with different $K$ -factors. The first three sub-figures shows the $E_b/N_0$ versus BER for individual modulation schemes employed in LTE-R and in last plot we compare all the modulation schemes for different $K$ -factors.	41
4.5	BER variation with time for HST with different modulation schemes of LTE-R. As the train moves towards the antenna the general trend of BER goes down with small-scale fluctuations due to varying K-factor.	42

4.6	Doppler spectrum for LTE-R at different train velocities $v$ (km/h) = 300, 400 and 500 and $f_c$ = 5 GHz. . . . .	44
4.7	HST channel model consisting of time-series K-factor and Doppler shift caused due to velocity of the train. . . . .	45
5.1	Experimental Test-Bed For Cooperative Sensing in Heterogeneous Network. Sensors 1, 2 and 4 are RTL-SDR units, sensor 3 is USRP N210 and TX is another USRP N210 unit which is used as a signal source for this work. . . . .	50
5.2	GNU Radio Flowgraph For Transmitter Running on USRP N210. . . . .	52
5.3	GNURadio Flowgraph For USRP and RTL-SDR sensor nodes. . . . .	52
5.4	Flowchart showing AND, OR and Majority Rule Fusion schemes. . . . .	54
5.5	Flowchart describing Maximum Normalized Energy Scheme in detail. . . . .	55
5.6	Flowchart describing Equal Gain Combining Scheme in detail. . . . .	56
5.7	Probability of Detection versus $SNR_{avg}$ For Hard Decision Combining. . . . .	58
5.8	ROC Characteristics for Hard Decision Combining with Different SNRs. . . . .	59
5.9	Probability of Detection versus $SNR_{avg}$ For Soft and Hard Decision Combining. . . . .	60

# List of Tables

2.1	Comparison of system parameters between GSM-R , LTE and LTE-R. . . . .	17
4.1	Tunnel and Tx/Rx Characteristics . . . . .	38
5.1	Operating Characteristics of Sensor Nodes . . . . .	52

# Chapter 1

## Introduction

### 1.1 Motivation

There has been an exponential growth in the Internet landscape. The devices like smartphones, computers, laptops, tablets will continue to increase along with multitude of other devices that are connected to Internet. This has facilitated the research in dynamic spectrum access (DSA) [1, 2] for efficient utilization of spectrum resources to sustain billions of Internet of Things (IoT) devices. There has been a significant increase in the study of cognitive radios for efficiently utilizing the electromagnetic spectrum. It has been observed that the spectrum occupancy is not uniform across all frequency bands, resulting in numerous spectral white spaces [3]. To opportunistically access the idle channel, spectrum sensing is considered to be a significant technology enabling DSA. Although several spectrum sensing techniques have been proposed in the open literature, energy detection is widely used due to its low implementation complexity [4]. We discuss some of the spectrum sensing techniques along with energy detection below.

- In energy detection (ED) scheme the energy of the signal is detected in the frequency location and based on the threshold value we decide whether the signal is present or absent.
- Cyclostationary Feature Detection is a complex scheme to implement compared to ED and it is mostly used when we need to also classify the signal present based on

their modulation scheme.

- When secondary user has a priori knowledge of primary user signal, matched filter (MF) detection is applied. Detection by using matched filter needs less detection time compared to ED but primary user information is required.

These spectrum sensing techniques can be used in a non-cooperative manner but it is very challenging to get an accurate estimate using a single-sensor system under a practical fading environment. Various non-idealities such as shadowing, multipath and fluctuating noise variance can make it difficult to detect the primary user [5, 6]. Cooperative spectrum sensing can mitigate the effects of multipath and shadowing by utilizing the spatial and temporal diversity of a multiple radio network [7, 8]. In cooperative spectrum sensing, each sensor node collects the spectral data and transmits it to a fusion center (FC) for decision making. Figure 1.1 shows how a heterogeneous sensor network exploits the spatial diversity.

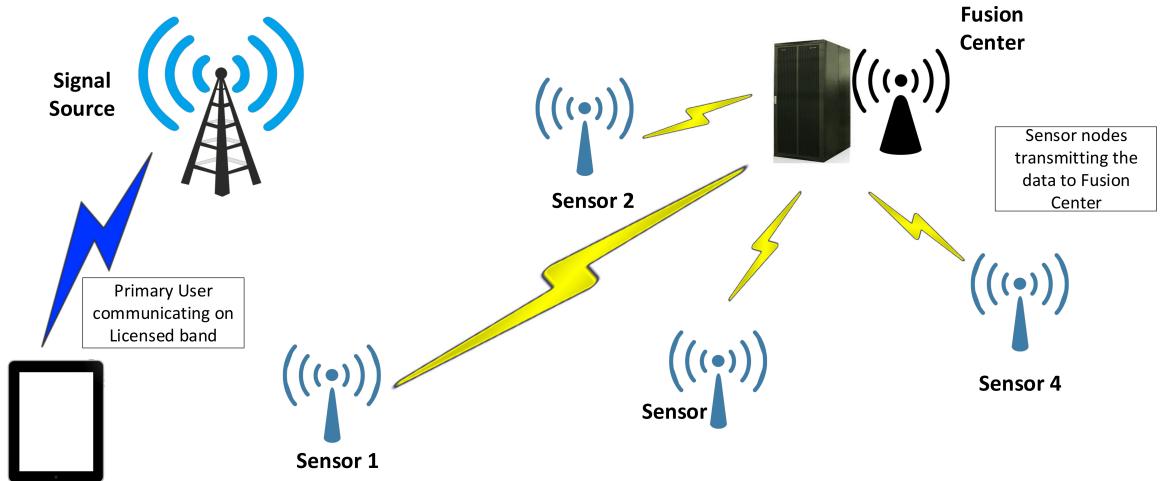


Figure 1.1: Heterogeneous sensor network employing cooperative spectrum sensing.  $RFFE_i$  and  $SR_i$  represents different front end and sampling rates for the SDR units.

We have also analyze the performance of Long Term Evolution for Railways (LTE-R) in a tunnel environment. In recent years, the use of trains have witnessed tremendous growth due to their increasing speeds, which has led to the demand for reliable wireless communication systems with these transportation systems. The development of a reliable wireless network for high speed trains is not a simple task and it is still an emerging technology. Global

System for Mobile Communication (GSM-R) [9], was a wireless communications standard designed for high speed trains, but it turned out not to be reliable enough and possess several limitations. Subsequently, LTE [10] proposed a promising solution for achieving broadband data rates in high speed trains that can overcome various GSM-R limitations [11, 12].

LTE-R is a high speed communication standard based on the existing LTE system architecture [12]. There has been several studies regarding the assessment of LTE-R as a viable choice for next generation high speed communications for railway applications [13, 14]. Most LTE systems operate at 1.8 GHz – 2.6 GHz bands, which possesses a high propagation loss and severe fading effects. Highly mobile trains inside tunnel environment makes the design of reliable communication links very challenging. To achieve reliable radio coverage inside tunnels, leaky feeder cables have been proposed [15]. With LCX, more uniform coverage can be achieved and installation is also comparatively simple. Each slot in the cable is equivalent to an antenna, which can transmit and receive signals. Figure ?? shows the LOS propagation environment inside a tunnel for a high speed train with velocity  $v$ .

## 1.2 State of the Art

The cooperative spectrum sensing testbed using normalized energy detection has been implemented and has been compared for both soft and hard data fusion scheme. Both soft data fusion and hard data fusion have been extensively studied [16–18], with several algorithms being implemented for each scheme. In a hard decision approach, each local decision statistic from sensor node is transmitted to an FC via overhead channels. The FC merges the sensing data and makes a global decision based on various algorithms such as majority rule, OR rule and AND rule [19]. For a soft decision scheme, each SU sends its local sensing data to the FC, which makes decision based on a global test statistic  $G$ . Soft decision combining improves the cooperative gain but it also possesses several limitations. With an infinite bandwidth, the real floating values can be transmitted to the FC, which can lead to a reliable decision mechanism. However, due to bandwidth constraint we have to quantize the data and this leads to error in the energy values. In hard decision combining, we can just transmit the decisions of the sensor nodes to the FC which can be binary values

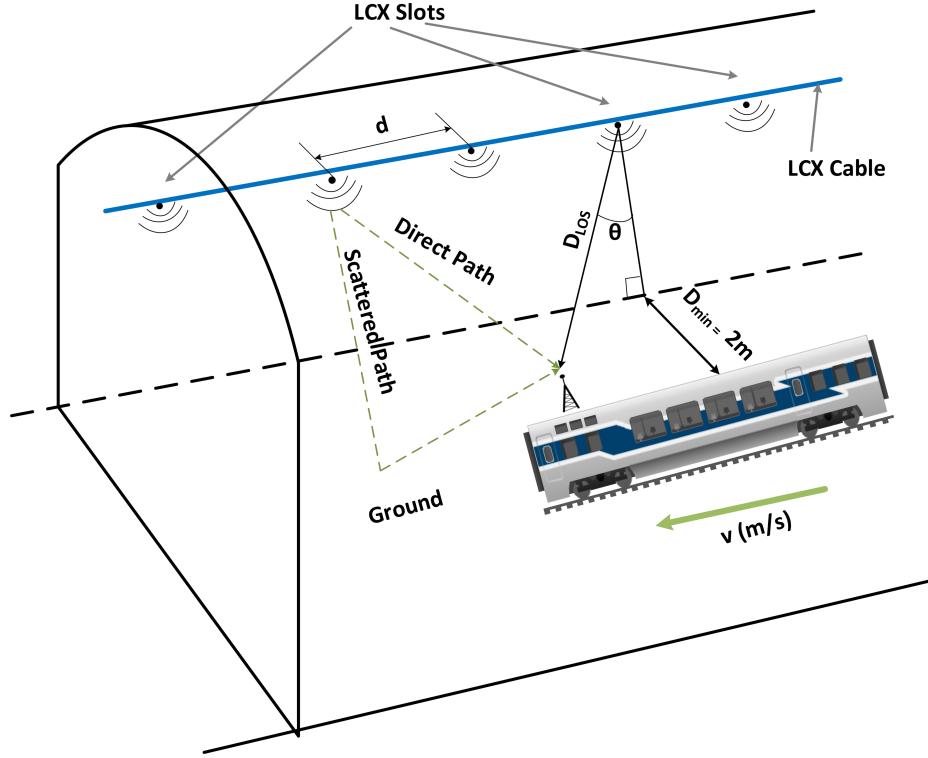


Figure 1.2: High speed train inside a tunnel for LTE-R.  $D_{LOS}$  is the distance between transmitter and receiver,  $d$  is the distance between the LCX cable transmission slots.

with 1 indicating that signal source is present and 0 indicating that a signal source is absent.

In open literature, most channel modeling have considered open areas with high speed trains [13, 14], while relatively little research has been conducted in tunnel environments. Due to various challenges presented by tunnel environments, it is important to derive a channel model for LTE-R involving high speed trains. In this paper, we analyze the effects of high Doppler shift and multipath due to tunnel environments. Experimental studies conducted inside tunnel environments have shown that the field amplitude distribution fits smoothly over a Rician distribution [20]. Several research efforts have been conducted on large-scale and small-scale fading of wideband communication systems inside tunnel environments. To the best of the authors knowledge, none of these studies have been conducted for LTE-R, which employs Orthogonal Frequency Division Multiplexing (OFDM) signals for data transmission inside tunnels. The large Doppler shifts caused by high speed

trains will potentially lead to ambiguity in extracting the carrier frequency, which can drastically increase the BER [21]. Therefore, it is important to study the effects of high Doppler shift and multipath fading for LTE-R communications in tunnel environments such that equalizers can be designed efficiently.

### 1.3 Thesis Contributions

This thesis will contribute the following to the cognitive radio communications and cellular wireless field:

- A cooperative spectrum sensing test-bed with normalized energy detection using both soft data fusion and hard data fusion implemented on available software defined radios.
- For soft data fusion, Maximum Normalized Energy (MNE) and Equal Gain Combination (EGC) algorithms are used. Hard data fusion is also implemented using majority rule, AND, and OR approaches. Both USRP N210s [22] and RTL-SDRs [23] are employed for the implementation of the heterogeneous sensor network.
- A performance assessment and simulation of LTE-R communications in a tunnel environment experiencing severe fading.
- Dynamic K-factor for a tunnel environment is derived using the classical two-ray propagation model [24] and is used to build Rician fading model for the tunnel.

### 1.4 Thesis Organization

This thesis will be organized into the following chapters: Chapter ?? provides background knowledge on heterogeneous cooperative spectrum sensing and focuses on heterogeneous networks, cooperative spectrum sensing and software-defined radios. Chapter 3 discusses LTE-R in detail and provides the necessary understanding of the LTE-R implementation in a tunnel environment. Leaky Coaxial Cable (LCX), channel impairments and two-ray propagation model is also discussed in detail. In chapter 4 and 5 implementation of heterogeneous cooperative spectrum sensing (CSS) test-bed and LTE-R performance

analysis in a tunnel is discussed. Chapter ?? presents the results of the implementation of heterogeneous CSS test-bed and LTE-R in a tunnel. Chapter 6 concludes this thesis, summarizing the accomplishments and outlines possible future work.

## Chapter 2

# Smart Railway Communication System

Gradually we are moving towards context awareness among automotive devices, where the vehicles are aware of their neighborhood. In modern trains a huge amount of wireless communications are used for safety features like train collision avoidance, railway management or customer oriented services such as passenger Internet access. To enhance reliability and safety of Railway systems while increasing accessibility and productivity, modern railway operations rely on ever increasing exchanges of information between the different wireless communication in terms of train-to-train (T2T) and train-to-ground (T2G). The integration of all the heterogeneous wireless networks deployed in the railway domain constitutes a key technical challenge. This can be answered by emerging Cognitive Radio (CR) technologies which can offer interoperability, reliability, dynamic spectrum access and lower deployment, and maintenance cost. The research is underway for cognitive radio enabled railway communication and two most famous projects are:

- Cognitive Radio for Railway Through Dynamic and Opportunistic Spectrum Reuse (CORRIDOR) [25] is a French research project that targets opportunistic spectrum access for railways. Due to rapid increase in demand for future railways in terms of control operations as well as providing high speed internet connectivity to the passengers, more bandwidth and spectrum is needed for railway communication.

- Rail-CR project [26] is a US-based Railway system project where the main effort is to implement a positive train control (PTC) technology to equip trains with wireless communication capabilities. The project aims to allow trains to communicate with wayside wireless stations, while moving to supply important information like speed, direction, etc. to improve safety and operations of the railways.

In the following section we discuss positive train control (PTC), wireless broadband (WiBro), train-to-train (T2T) and train-to-ground (T2G) communication, Long Term Evolution for Railways (LTE-R) communication system, spectrum regulation for LTE-R, LTE and Internet-of-Things (IoT) for railways which are important technologies for achieving smart railway transportation system.

## **2.1 Positive Train Control**

## **2.2 Wireless Broadband (WiBro)**

## **2.3 T2T and T2G Communication System**

## **2.4 LTE-R Communication System**

In recent years, the use of trains have witnessed tremendous growth due to their increasing speeds, which has led to the demand for reliable wireless communication systems with these transportation systems. The development of a reliable wireless network for high speed trains is not a simple task and it is still an emerging technology. Global System for Mobile Communication (GSM-R) [9], was a wireless communications standard designed for high speed trains, but it turned out not to be reliable enough and possess several limitations. The data rate of voice services which can reach up to 9.6 kbps, which can't meet the increasing demands of high-rate data transmission in railways communication. The limited data rate and quality of service (QoS) is not enough to support cellular communication for current generation. These reasons have made the evolution of high speed railway communication more and more urgent. Subsequently, LTE [10] proposed a promising solution for

achieving broadband data rates, flexible bandwidth allocation and high spectral efficiency in high speed trains that can overcome various GSM-R limitations [11, 12].

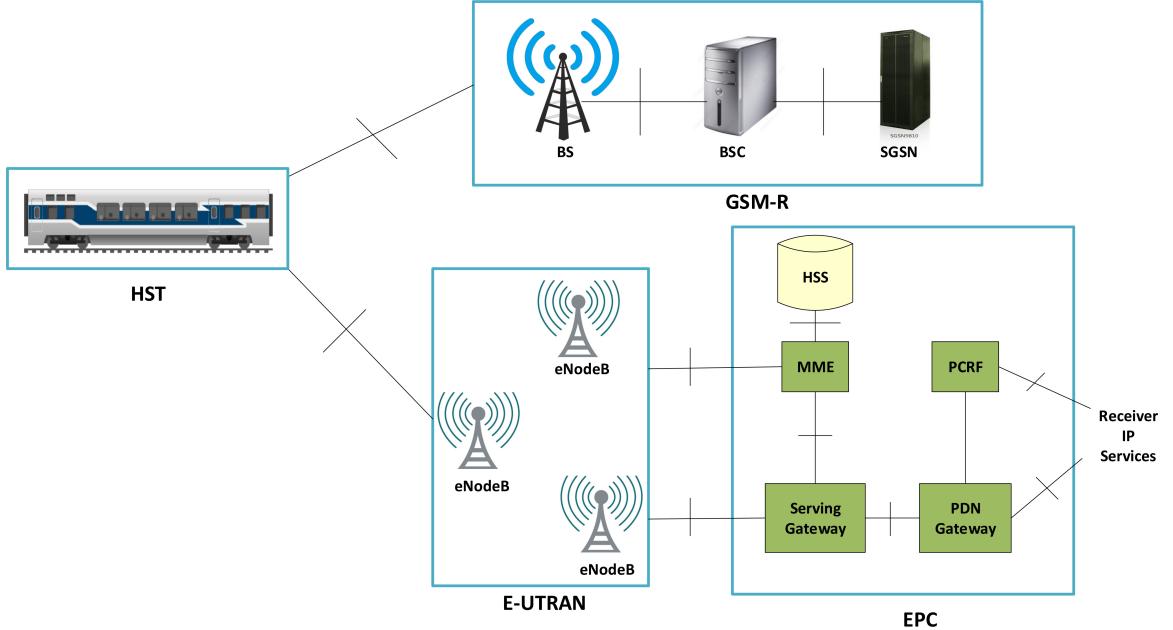


Figure 2.1: Proposed LTE-R Architecture for next generation High-speed Railways consisting of EPA and E-UTRAN.

LTE-R is a high speed communication standard based on the existing LTE system architecture [12]. There has been several studies regarding the assessment of LTE-R as a viable choice for next generation high speed communications for railway applications in [5, 6]. Conventional LTE includes a core network of evolved packet core (EPC) and a radio access network of Evolved Universal Terrestrial Radio Access Network (E-UTRAN). The Internet protocol (IP)-based EPC supports seamless handovers for both voice and data to cell towers, and each E-UTRAN cell will support high data and voice capacity by high-speed packet access (HSPA). As a candidate for the next-generation communication system of HSR, LTE-R inherits all the important features of LTE and provides an extra radio access system to exchange wireless signals with onboard units (OBUs) and to match HST-specific needs. Figure 2.1 shows the proposed architecture of LTE-R according to [10], and it shows that the core network of LTE-R is backward compatible with GSM-R. The network architecture of LTE-R is similar to that of LTE/SAE with Evolved Univer-

sal Terrestrial Radio Access Network (E-UTRAN) being the access network structure of LTE-R. There is evolved-Node B (eNodeB) that communicates directly with UEs like base transceiver station (BTS) in GSM network. It performs the transmission and reception of data packets using orthogonal frequency division multiplexing access (OFDM) for down-link and single carrier frequency division multiple access (SC-FDMA) for uplink at PHY layer. At the same time, as without base-station controller (BSC), it also has functions of radio resource control and wireless mobility management. The eNodeBs can be connected to network router directly without more intermediate control nodes, such as the BSC in GSM-R [27]. The core network of LTE-R is the Evolved Packet Core (EPC). The significant difference between EPC and the core network of GSM-R is that the EPC is an all-IP mobile core network.

Table 2.1: Comparison of system parameters between GSM-R , LTE and LTE-R.

System Parameters	GSM-R	LTE	LTE-R
Frequency	Uplink: 876–880 MHz downlink: 921–925 MHz	800, 1800, 2600 MHz	450, 800, 1400, 1800 MHz
Capacity	0.2 MHz	1.4-20 MHz	1.4-20 MHz
Modulation	GMSK	QPSK/16-QAM/64-QAM	QPSK/16-QAM
MIMO	No	2x2, 4x4	2x2
Cell Range	8 Km	1-5 Km	4-12 Km
Data Rates (DL/UL)	172/172 Kbps	100/50 Mbps	50/10 Mbps

Conventional LTE networks are different compared to LTE-R in many ways such as architecture, system parameters, network layout, services and quality of service (QoS). Table 2.1 summarizes the LTE-R parameters and describe the differences between LTE, GSM-R and LTE-R. Since the LTE-R environment has very severe fading and high Doppler shift, it is configured for QoS rather than higher data rates. Therefore, QPSK modulation is used for most number of subcarriers and the number of packet re-transmission must be kept low and achieved with user datagram protocol (UDP).

## 2.5 Spectrum Regulation for LTE-R

Departing from technical issues, we now discuss the crucial interactions that 5G will encounter with spectrum policy and allocation by the federal communications commission (FCC). As already discussed in Section ?? the spectrum allocated for cellular technologies is already saturated in peak markets due to massive amounts of wireless services and networks. Figure 2.2 shows the pronounced scarcity in the the wireless cellular bands as seen in the FCC frequency allocation chart. Large amount of unused spectrum is available in the mm-Wave realm and can be used for high data rate applications. Due to different propagation characteristics for signals at low and sub-terahertz frequencies, future systems will need to integrate a broad range of frequencies. Frequencies on a lower spectrum can be used for wide coverage, mobility support, control signaling and high frequencies for high data rate applications in small cells. This requires a new approach to spectrum policy and allocation methods for 5G standardization.

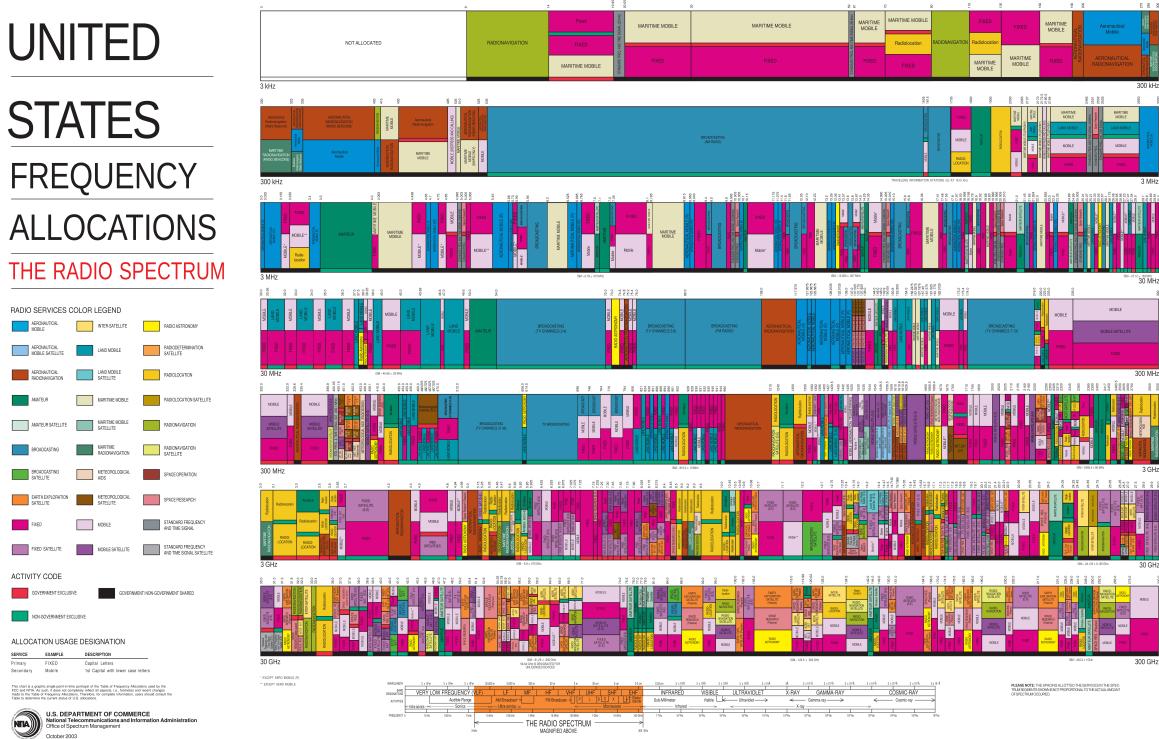


Figure 2.2: Machine-to-Machine communication paradigm in 5G Cellular System.

Cognitive radio is a promising technology that can solve the spectrum shortage problem arising due to the rapid increase in wireless networks and mobile devices. Recent advancements in software-defined radio technology and edge computing have enhanced the cognitive radio network (CRN) capabilities and, along with some adjustments in its operation, will be a key technology for 5G heterogeneous network deployment. The CR network is an innovative software-defined radio technique considered to be one of the key technologies to improve the utilization of the congested radio spectrum [28]. Integrating CR technology into 5G system is motivated by the fact that a large portion of the radio spectrum is underutilized most of the time. For achieving data rates in order of gigabits per second, we need to make an efficient use of available spectrum which can be achieved by utilizing CRNs. In CR networks, a secondary system can share spectrum bands with the incumbent primary system, either on an interference-free basis or on an interference-tolerant basis. The CR network should be aware of the surrounding radio environment and regulate its transmission accordingly. In interference-free CR networks, CR users are allowed to borrow spectrum resources only when licensed users do not use them. A key to enabling interference-free CR networks is figuring out how to detect the spectrum holes (white space) that spread out in wideband frequency spectrums and facilitate dynamic spectrum access (DSA) smoothly [29].

CR receivers should first monitor and allocate the unused spectrum via spectrum sensing (energy detection (ED), covariance absolute value (CAV) detection, etc.) or combining with geolocation databases and feed this information back to the central CR controller. A coordinating mechanism is required in multiple CR networks that try to access the same spectrum to prevent users colliding when accessing the matching spectrum holes. In interference-tolerant CR networks, CR users can share the spectrum resource with a licensed system while keeping the interference below a threshold. In comparison with interference-free CR networks, interference-tolerant CR networks can achieve enhanced spectrum utilization by opportunistically sharing the radio spectrum resources with licensed users, as well as better spectral and energy efficiency. However, it has been shown that the performance of CR systems can be very sensitive to any slight change in user densities, interference threshold, and transmission behaviors of the licensed system. However, the spectral efficiency can be improved by either relaxing the interference threshold of the primary system or considering

only the CR users who have short distances to the secondary BS (utilizing the spatial gain). Hybrid CR networks have been proposed in [30] for adoption in cellular networks to explore additional bands and expand the capacity. CR networks can only prove beneficial if the spectrum policies related to 5G are implemented in a robust manner.

## 2.6 Vehicular Communication Using LTE

With the development of mm-Wave and massive-MIMO, the spectral and energy efficiency for 5G wireless communications has been drastically improved. The development of driverless cars has posed some rigorous requirements for the safety of the passengers and pedestrians. For safety-critical applications the transmission delay needs to be less than 1 ms which is required for intelligent transportation systems (ITSs) and vehicular networks [31]. Considering the drawbacks of IEEE 802.11p networks, such as poor scalability, low capacity , and intermittent connectivity, the Long Term Evolution (LTE) mobile communication technologies were proposed to support vehicular applications [32]. Simulation and experiment results revealed that there is a trade-off between the proposed performance metrics and system parameters, such as base station (BS) and vehicle densities, radio coverage, and the maximum number of hops in a path. When LTE communication technologies have been integrated into vehicular networks, the interference has cut down the performance of LTE vehicular networks. When vehicle density is high, the beaconing signals of vehicular safety applications may easily overload the serving eNodeB. To handle this issue, a significant amount of such signals should be distributed directly among vehicles, without going through the eNB. In LTE-Advanced (LTE-A), device-to-device (D2D) communication is considered to allow direct message delivery between terminals in proximity to lighten the load of eNB [33].

Hence the infrastructure-aided D2D technologies can serve as a natural approach to enable reliable and efficient V2V communications without negatively affecting existing cellular systems. To meet the high performance requirements, such as low transmission delay and high throughput, a new architecture for 5G vehicular communication is required. Figure 2.3 describes the vehicular communication using D2D communication strategy where significant

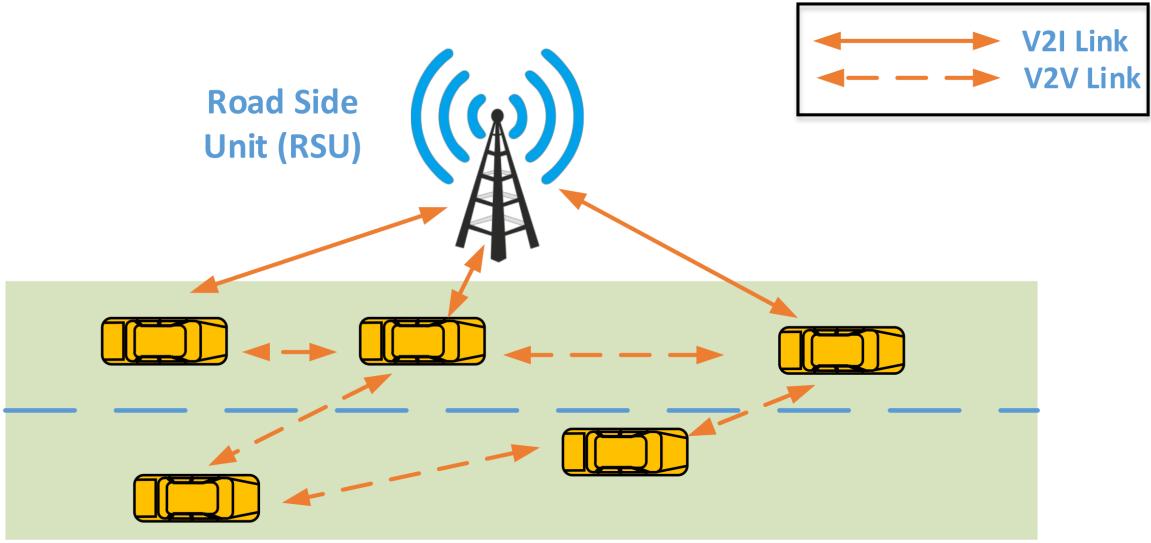


Figure 2.3: Vehicular communication using D2D 5G cellular communication technology.

amount of computation is offloaded from road side units (RSU) to vehicles to avoid increase spectral and efficiency. The communication environment in V2V is quite different than in D2D due to high mobility (Doppler shift) of vehicles. Network connectivity play a important role in V2V communication than D2D, compared with system throughput. These features can significantly affect D2D resource allocation strategies and system parameters, and thus should be modified for vehicular communication.

### 2.6.1 Internet of Things (IoT) for Railways

Machine to machine (M2M) communication is a new paradigm proposed to be a key technology in 5G wireless systems that enables the ubiquitous connectivity between a set of devices with different network stack. Thus, the autonomous connection of devices facilitates the emergence of a wide range of intelligent M2M applications. M2M exhibits a strong potential to improve human life in different fields such as e-Health, smart grids, smart cities, intelligent transportation and surveillance enabling internet of things (IoT) applications. A native inclusion of M2M communication in 5G involves satisfying three fundamentally different requirements associated with different classes of low data rates and power consumption services such as: enabling massive number of low-rate devices simultaneously, a

minimal data rate in virtually all scenarios, even in worst fading environment and finally managing the above two requirements with a very low latency.

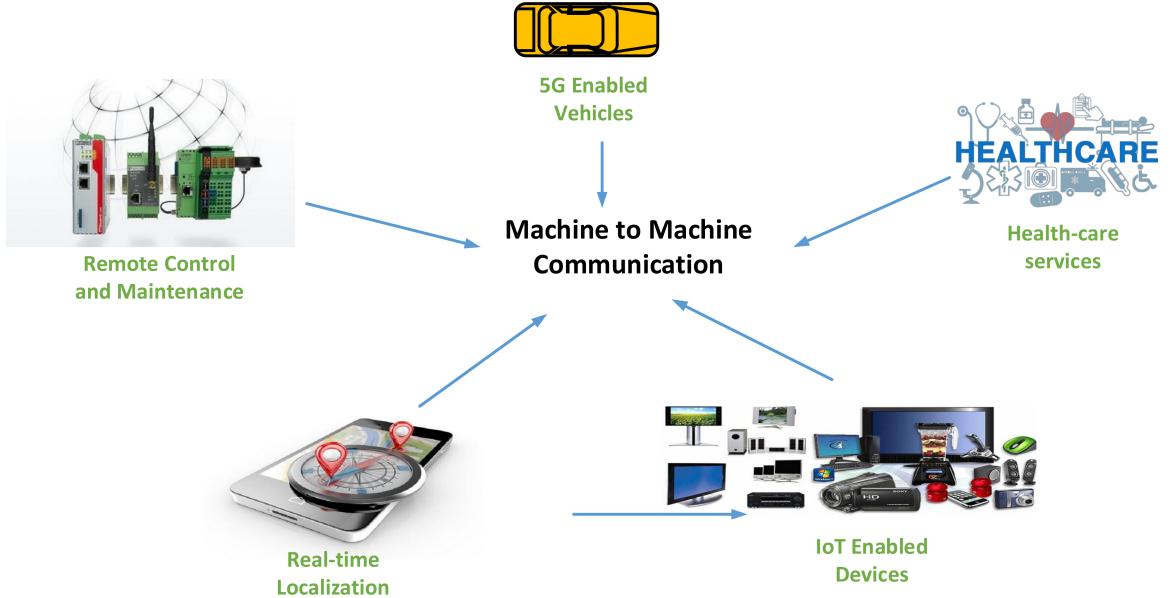


Figure 2.4: Machine-to-Machine communication paradigm in 5G Cellular System.

Implementing these requirements for 5G system, new communication paradigms needs to be developed at network and physical layer. Figure 2.4 shows the machine-to-machine communication and how it helps in integrating devices which can be used in health care, localization, vehicular communication, remote maintenance and consumer electronics. In the future standardization of the 5G networks, the M2M communication is recognized as one of the key technologies that will support the 5G architecture. As a consequence of this proliferation of M2M technology, large numbers of devices will produce huge amount of irregular data and significantly higher capacity of cellular networks will be needed, in terms of spectrum, power control as well as the number of devices serviced simultaneously. For next generation 5G system, it would be common to have situations where devices will be in close proximity to each other and would like to wirelessly share data like videos, images, etc. or would like to interact for gaming, social networking. If these communication scenarios are handled by simply connecting through the network it can lead to various inefficiencies at various levels. The reason is that cellular networks were primarily designed

to support mobile devices with a large amount of data to transmit, whereas most M2M communications are duty-cycling and delay-tolerant with small data packets between stationary terminals [34]. Thus, connecting a massive number of M2M devices directly with cellular networks will saturate uplink traffic easily [35]. In this context, non-cellular connections should serve as important supplement and extension to cellular networks. Some of the challenges of connecting the devices via cellular networks are:

- Multi-hopping is used to reach destination which otherwise requires fundamentally a single hop. This entails a multi-fold waste of signaling resources as well as higher latency.
- Huge amount of transmit power is consumed in uplink( fraction of a watt) and down-link( several watts) for data communication which can be easily achieved by milli-watts of power. This leads to unnecessary levels of battery drainage and interference to all other devices occupying the same signaling resources in near proximity.
- Propagation path-losses to the base stations are much stronger than direct link to the device, hence the corresponding spectral inefficiencies are also lower.

It is pretty evident that M2M has huge potential to handle single hop inter device communication more efficiently but these task could be offloaded to other radio access technologies such as Bluetooth, ultra-wide band (UWB) or Zigbee. The applications requiring a mixture of low-latency and high data rates for e.g. interaction between users via augmented reality could be more apt for 5G M2M communication system. In particular, machine-to-machine is as an important enabler for applications requiring low latency, especially in future network deployments utilizing baseband centralization and radio virtualization. There are still various research challenges which need to be taken care of before we can integrate this technology in 5G cellular communication system. The devices for M2M communication need to be designed from both hardware and protocol perspective by providing the needed flexibility at both the PHY and MAC layers. Research needs to be conducted for possible extra overheads for control and channel estimation and also accessing of true net gains associated with M2M mode. And finally, accurate simulation and experimental test-beds needs to be designed for testing M2M communication.

## 2.7 Leaky Coaxial Cable for LTE-R

Leaky Coaxial Cable (LCX) [36] is an antenna technology designed to deliver radio services in tunnel environment. It consists of small periodic slots to allow radio frequency (RF) signals to escape which act as extended antenna elements. LCX cables were invented to provide the uniform signal coverage in underground mine where radio coverage was very limited due to the geography of the mines. Recently, leaky coaxial cables have been widely used in the field of railway communication especially in tunnels. Leaky feeders are constructed from coaxial cable where the outer shield has a series of holes with different shapes and different distances among them. The coaxial cable is usually about hundreds of metres long and can be fixed through a building or a tunnel offering radio radiation in a way that would require many individual omni-directional antennae. So far, they have been only used to supplement the direct wireless communication system between a BS and trains, mostly transmitting voice signals. Nowadays, they are being used as an alternative solution to distributed antenna systems in indoor environments like commercial buildings [37, 38] and university buildings, high speed trains, cars, etc. The LCX radio system is almost noise free and has enough bandwidth to support multiple RF signals carrying voice and data simultaneously. Figure ?? shows the conventional leaky coaxial cable along x-axis with periodic radiating slots and wave propagation along z-axis. Generally, it consists of three parts: inner conductor, dielectric material and outer conductor. LCX has a dual functionality i.e. they can transmit and receive RF signals using their slots. The frequency range for a leaky cable is given by [39]:

$$\frac{c}{\sqrt{\varepsilon_r - 1})d} \geq f \leq \frac{c}{\sqrt{\varepsilon_r + 1})d} \quad (2.1)$$

A radio system based on LCX has been deployed in Japan for high speed railways "Series N700" [40] to connect the train to ground network. Wi-Fi access points are chosen for in-train communications with peak data rates of 2 Mbps for uplink and downlink. Although current technologies can provide wireless communication services in HSTs, the capacity of communication system is very low (1–4 Mbps). These data rates are insufficient for next generation wireless communication system where the peak data rates of 0.5 – 5 Gbps are expected. LTE-R communication system can be implemented for achieving high data rates

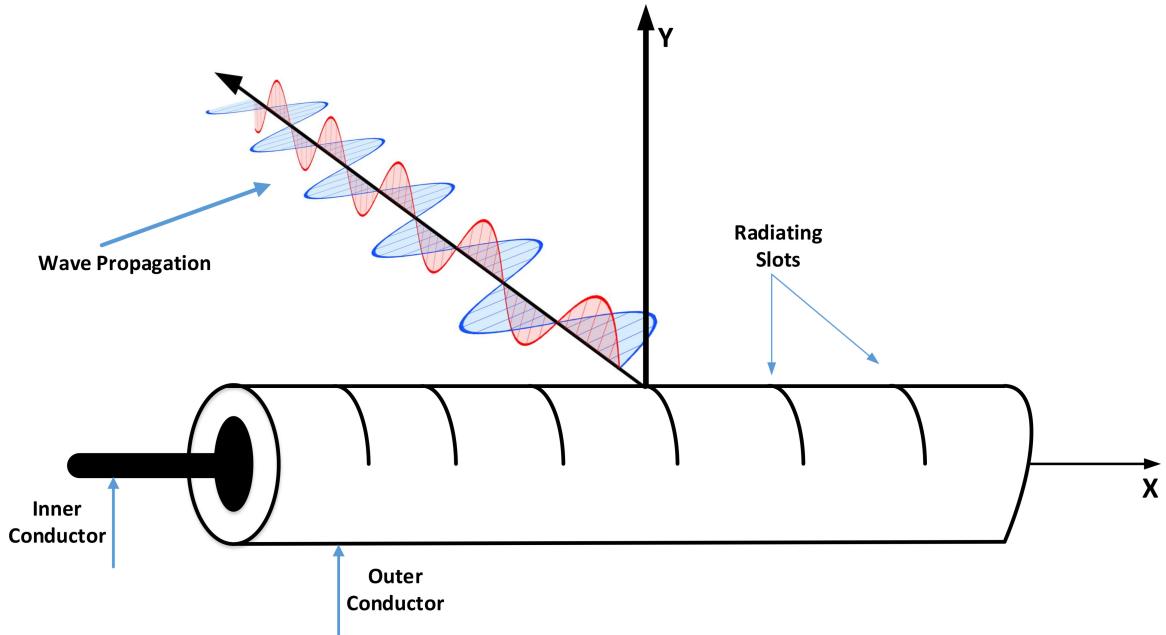


Figure 2.5: Leaky Coaxial Cable

but it cannot be achieved by using conventional cellular system. The penetration loss due to tunnel walls is very high and secondly, the fast moving trains causes large Doppler shifts leading to poor connectivity due to retransmissions. Hence, leaky coaxial cable is the best candidate for achieving extensive internet access inside a tunnel environment for high speed railways.

## Chapter 3

# Heterogeneous Cooperative Spectrum Sensing (CSS)

This chapter provides the background information needed to understand the chapters that follows. It examines the basic outlines of a heterogeneous networks and how cooperative spectrum sensing (CSS) can help in enhancing the accuracy of signal source estimation. The fusion center (FC) collects the data from the sensor node network and process it to make the reliable decision. Secondly, this chapter investigates various algorithms which can be used in heterogeneous network to estimate signal source. Finally, it also outlines the necessary hardware and software tools used in the implementation of heterogeneous CSS chapter.

### 3.1 Spectrum Sensing

Spectrum sensing plays a key role in the decision-making part of cognitive radio network. It is required by CR to detect the presence of spectrum white space and also to estimate accurately the presence of incumbent users (PUs). Since the primary users have prerogative on the spectrum band usage, so it is of paramount importance to avoid interfering with the PU. It is very challenging to get an accurate estimate under a practical fading environment based on conventional spectrum sensing techniques. Various non-idealities such as shadowing, multipath and fluctuating noise variance can make it difficult to detect the primary

user. In the case of fast varying channels over time, some works focus on improving the performance of spectrum sensing and signal identification [41–43]. To combat the Doppler shift caused due to high-speed environments algorithms [44] have been proposed for channel estimation and equalization. Cooperative spectrum sensing [45] can be used to tackle lot of the problems caused due to multipath, shadowing and high Doppler shift. In the paper we tested the performance of cooperative spectrum sensing using soft and hard data fusion schemes on software-defined radio test-bed. The experiment confirmed that cooperation among sensor nodes improved the spectrum sensing performance as a result of increased spatial-temporal diversity in received signal source. The various types of spectrum sensing schemes are discussed below in the following subsections.

### 3.1.1 Energy Detection

In energy detection (ED) we use the energy spectra of the received signal and compare it against a predefined threshold level to estimate the presence of the signal. In ED scheme we only rely on the energy of the signal in the frequency channel, and no phase information is required. The key advantage of the ED scheme is that it doesn't require any prior information of the signal, i.e. type of modulation scheme, phase information or any other signal parameter. Energy detection can be considered as binary hypothesis testing scheme and is given by:

$$y(n) = \begin{cases} w(n), & \mathbf{H}_0 \\ s(n) + w(n), & \mathbf{H}_1 \end{cases} \quad (3.1)$$

where  $y(n)$  represents the received signal,  $s(n)$  represents the signal source (PU), and  $w(n)$  is the white Gaussian noise  $w(n) \sim N(0, \sigma_n^2)$ .  $\mathbf{H}_0$  describes the hypothesis when there is no signal present, while the hypothesis  $\mathbf{H}_1$  is the presence of signal.

The decision whether the signal is present or absent is decided by evaluating a local test statistic  $L$  to see whether it is above or below certain fixed threshold  $\tau$ . The local test statistic  $L$ , which is the complex-magnitude squared of the FFT samples, is compared with  $\tau$  using equation:

$$L = \sum_{n=1}^M |y(n)|^2 = \begin{cases} < \tau, & \mathbf{H}_0 \\ > \tau, & \mathbf{H}_1 \end{cases} \quad (3.2)$$

where  $|y(n)|^2$  is the energy of a specific FFT bin and  $n=1,2,3\dots M$  are the number of samples received.

The probability of false alarm  $P_{fa}$  and probability of detection  $P_d$  are given by:

$$P_f = Q\left(\frac{\tau - M(2\sigma_n^2)}{\sqrt{M(2\sigma_n^2)}}\right), \quad (3.3)$$

$$P_d = Q\left(\frac{\tau - M(2\sigma_n^2)(1 + \gamma)}{\sqrt{M(1 + 2\gamma)(2\sigma_n^2)}}\right). \quad (3.4)$$

In cooperative spectrum sensing, each sensor node transmits the local sensing data to the fusion center for signal source detection. The local sensing data has to be quantized, thus yielding quantization errors. To minimize the quantization error in local test statistic  $L$  and to reduce the effect of noise variance, the energy of the received signal  $y(n)$  is normalized [?].

The local test statistic  $L$  for the  $r^{th}$  sensor node is given as:

$$L_r = \frac{1}{M_r \sigma_{n,r}^2} \sum_{r=1}^{M_r} |y(n)|^2 \quad (3.5)$$

where  $M_r$  is the number of samples used to estimate the power of the signal source in the node,  $\sigma_{n,r}$  is the noise power variance.

In Eq (3.8)  $s(n)$  is considered as a deterministic signal and  $w(n)$  is a Gaussian random variable with a variance of  $\sigma_n^2$ . Based on CLT,  $L_r$  will have a following distribution [7]:

$$L_r = \begin{cases} N(1, \frac{1}{M_r}), & \mathbf{H}_0 \\ N(\gamma_r + 1, \frac{1 + 2\gamma_r}{M_r}), & \mathbf{H}_1 \end{cases} \quad (3.6)$$

where  $\gamma_r$  is the received SNR of the  $r^{th}$  SU. The local decision statistic  $L_r$  is quantized before transmission due to the bandwidth constraint, and this can lead to quantization errors. The values of  $L_r$  received by FC can be modeled as:

$$\beta_r = L_r + w_{q,r}, \quad (3.7)$$

where  $\beta_r$  is the decision statistic received by the FC and  $w_q$  is the noise added to the signal due to fading and quantization error. In [46], the  $w_q$  is modeled as a Gaussian noise with zero mean and  $\sigma_q^2$  variance. Figure explains the normalized energy detection scheme in the block diagram form.

### 3.1.2 Cyclostationary Method

### 3.1.3 Matched Filter Detection

### 3.1.4 Wavelet Detection

## 3.2 Cognitive Radios

Cognitive radio (CR) [47] is a communication systems paradigm that focuses on employing highly agile, environmentally aware, intelligent wireless platforms in order to autonomously select and configure device operating parameters based on the prevailing radio and network environmental conditions [3]. In general the cognitive radio may be expected to look at parameters such as channel occupancy rate, available channels, bandwidth required for data transmission and the modulation types that may be used. It must also look at the regulatory requirements set by the Federal Communications Commission. In some instances a knowledge of geography and this may alter what it may be allowed to do. Software-defined radios (SDR) are mainly responsible for making cognitive radios used in wireless communications system a reality. Software radios provide a vast untapped potential to personalize services, and they make the process of modifying the radio characteristics extremely simple.

The work is underway to determine the best possible methods of developing the cognitive radio communications system that can fulfill all its requirements. To facilitate the intelligent decision making capabilities in these cognitive radio systems, machine learning algorithms have been proposed in the literature [48–51] to automate the reconfiguration process. The Figure 3.1 describes the various building blocks of a cognitive radio system. The spectrum sensing is performed to estimate the spectrum holes in the band and after the analysis the decision strategy is prepared. The radio is configured with the new parameters based on

the radio environment and the spectrum decision made.

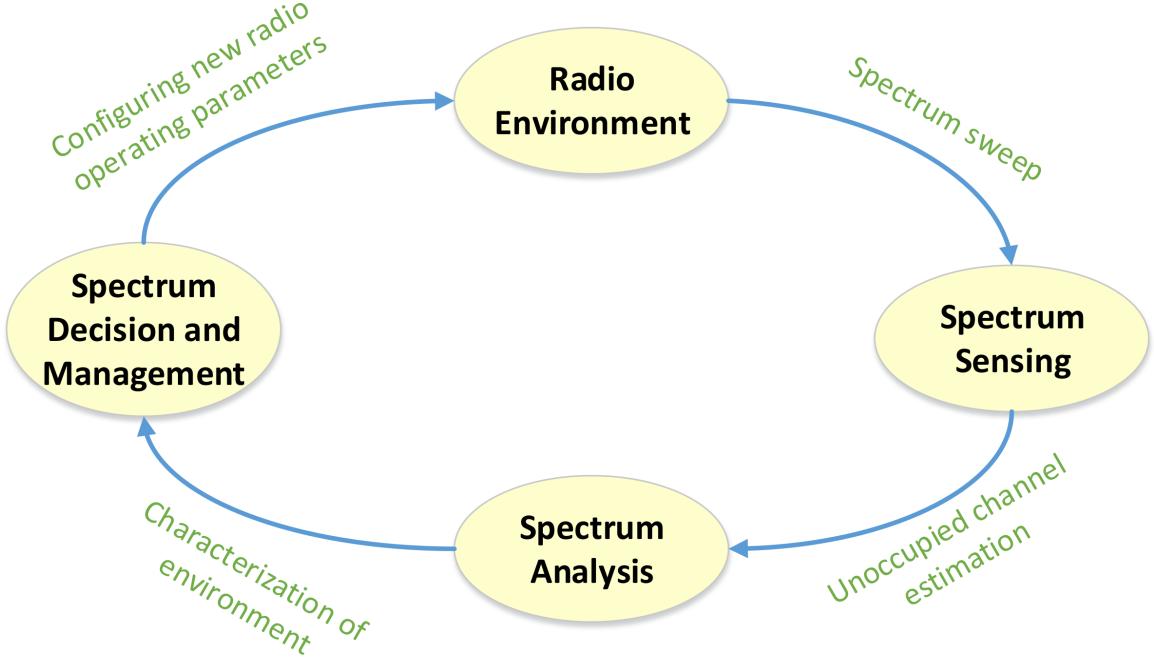


Figure 3.1: The block diagram explaining the basic parts of Cognitive Radio system. The operating parameters are configured based on the characterization of the wireless environment.

### 3.3 CSS in Heterogeneous Networks

In Heterogeneous CR network, each radio is equipped with different numbers of antennas, sampling rates and RF characteristics. In addition, each sensor node may experience distinct channel fading and suffer from different noise levels due to their respective locations and device performances, such as amplifier and ADC. As a result, each node may have different sensing capabilities and reliability values. This is a universal and fundamental characteristic of a heterogeneous CR network which requires robust algorithms to achieve high accuracy in signal source detection for estimating the presence of a primary user [52]. In this thesis, we investigate the cooperative spectrum sensing in heterogeneous networks with a centralized FC, a transmitter acting as a signal source and four sensor nodes. As explained earlier we are using energy detection as one of the spectrum sensing technique since it possesses a

very low implementation complexity [4]. The energy detection scheme detects the presence or absence of a signal source based on its intercepted energy signature. If the energy of the signal is higher than a certain threshold, this indicates that the channel is occupied. The ED can be modeled by the equation:

$$y(n) = \begin{cases} w(n), & \mathbf{H}_0 \\ s(n) + w(n), & \mathbf{H}_1 \end{cases} \quad (3.8)$$

where  $y(n)$  represents the received signal,  $s(n)$  represents the signal source PU, and  $w(n)$  is the white Gaussian noise  $w(n) \sim N(0, \sigma_n^2)$ .  $\mathbf{H}_0$  describes the hypothesis when there is no signal present, while the hypothesis  $\mathbf{H}_1$  is the presence of signal.

The decision whether the signal is present or absent is decided by evaluating a local test statistic  $L$  to see whether it is above or below certain fixed threshold  $\tau$ . The local test statistic  $L$ , which is the complex-magnitude squared of the FFT samples, is compared with  $\tau$  using equation:

$$L = \sum_{n=1}^M |y(n)|^2 = \begin{cases} < \tau, & \mathbf{H}_0 \\ > \tau, & \mathbf{H}_1 \end{cases} \quad (3.9)$$

where  $|y(n)|^2$  is the energy of a specific FFT bin and  $n=1,2,3\dots M$  are the number of samples received.

The probability of false alarm  $P_{fa}$  and probability of detection  $P_d$  are given by:

$$P_f = Q\left(\frac{\tau - M(2\sigma_n^2)}{\sqrt{M(2\sigma_n^2)}}\right), \quad (3.10)$$

$$P_d = Q\left(\frac{\tau - M(2\sigma_n^2)(1 + \gamma)}{\sqrt{M(1 + 2\gamma)(2\sigma_n^2)}}\right). \quad (3.11)$$

In cooperative spectrum sensing, each sensor node transmits the local sensing data to the fusion center for signal source detection. The local sensing data has to be quantized, thus yielding quantization errors. To minimize the quantization error in local test statistic  $L$  and to reduce the effect of noise variance, the energy of the received signal  $y(n)$  is normalized [52].

The local test statistic  $L$  for the  $r^{th}$  sensor node is given as:

$$L_r = \frac{1}{M_r \sigma_{n,r}^2} \sum_{r=1}^{M_r} |y(n)|^2 \quad (3.12)$$

where  $M_r$  is the number of samples used to estimate the power of the signal source in the node,  $\sigma_{n,r}$  is the noise power variance.

In Eq (3.8)  $s(n)$  is considered as a deterministic signal and  $w(n)$  is a Gaussian random variable with a variance of  $\sigma_n^2$ . Based on CLT,  $L_r$  will have a following distribution [7]:

$$L_r = \begin{cases} N(1, \frac{1}{M_r}), & \mathbf{H}_0 \\ N(\gamma_r + 1, \frac{1 + 2\gamma_r}{M_r}), & \mathbf{H}_1 \end{cases} \quad (3.13)$$

where  $\gamma_r$  is the received SNR of the  $r^{th}$  SU. The local decision statistic  $L_r$  is quantized before transmission due to the bandwidth constraint, and this can lead to quantization errors. The values of  $L_r$  received by FC can be modeled as:

$$\beta_r = L_r + w_{q,r}, \quad (3.14)$$

where  $\beta_r$  is the decision statistic received by the FC and  $w_q$  is the noise added to the signal due to fading and quantization error. In [46], the  $w_q$  is modeled as a Gaussian noise with zero mean and  $\sigma_q^2$  variance.

### 3.4 Software Defined Radios

We have already explained about the cooperative spectrum sensing in heterogeneous networks, now in this section we will look at the platform for testing the CSS algorithms. The platform which we use is a new hardware frontier called Software-Defined Radio (SDR) and is discussed in details.

There has been a huge shift in the definition of a Software-Defined Radio and it has a lot to do with the question of where the hardware ends and software begins. Dr. J. Mitola III coined the Software Defined Radio which he described as a of digital signal processing (DSP) primitives, a meta-level system for combining the primitives into communication system (Tx, channel model, Rx, etc.), functions and a set of target processors on which the software radio is hosted for real-time communications [53]. Dr. J. Mitola explains in the thesis how software provides the flexibility which using hardware alone can never be

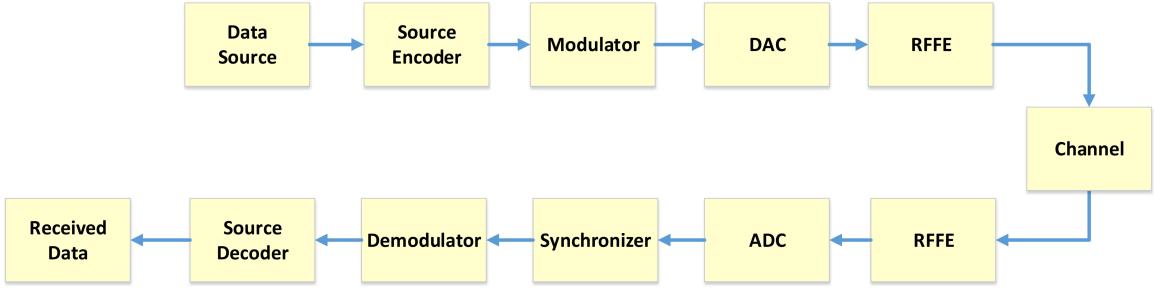


Figure 3.2: Software defined radio pushes all the adaptive elements and data manipulation operation into software. The goal of SDR is to provide or define all of the radio operation in software.

achieved. And as time progresses SDR would become more dominant and close to ideal Cognitive Radio system.

The SDR technology existed since the 1970s but the key milestone in the advancement of SDR technology took place in early 1990s with the U.S. military initiative called SpeakEasy I/II. The SpeakEasy project was implemented to use programmable processing to emulate more than ten existing military radios, operating in frequency bands between 2 MHz and 2 GHz [54]. With SpeakEasy the operator could talk to ten radios operating under different standards with any hardware modifications. With all these features, unfortunately there were some shortcomings which left much to be desired. The device was large enough to fit on the back of a pickup truck [54] which is good for ground station but not if the mobility is an important factor. And in 1992 the field programmable gate arrays (FPGA) were not computationally efficient, hence required large time to change their operating characteristics. The two software-defined radios we used in the thesis are Universal Software Radio Peripheral (USRP N210) and RTL-SDR R2832U. In the subsequent subsections we discuss the two SDRs in detail.

### 3.4.1 USRP N210 and RTL-SDR

The Universal Software Radio Peripheral (USRP) N210 has a very different RF characteristics compared to RTL-SDR hence modeling an ideal heterogeneous network. The USRP N210 provides a very high bandwidth, dynamic range processing capability. The

product architecture includes a Xilinx Spartan 3A-DSP 3400 FPGA [55], 100 MS/s dual ADC, 400 MS/s dual DAC and gigabit ethernet connectivity to stream data to and from host processors. A modular design allows the USRP N210 to operate from DC to 6 GHz, while an expansion port allows multiple USRP N210 series devices to be synchronized and used in a MIMO configuration. An optional GPSDO module can also be used to discipline the USRP N210 reference clock to within 0.01 ppm of the worldwide GPS standard. The USRP N210 can stream up to 50 MS/s to and from host applications. Users can implement custom functions in the FPGA fabric, or in the on-board 32-bit RISC softcore. The USRP N210 provides a larger FPGA than the USRP N200 for applications demanding additional logic, memory and DSP resources. The FPGA also offers the potential to process up to 100 MS/s in both the transmit and receive directions. The FPGA firmware can be reloaded through the Gigabit Ethernet interface [22].

RTL-SDR is a very cheap software defined radio that uses a DVB-T TV tuner dongle based on the RTL2832U chipset. With the combined efforts of Antti Palosaari, Eric Fry and Osmocom it was found that the signal I/Q data could be accessed directly, which allowed the DVB-T TV tuner to be converted into a wideband software defined radio via a new software driver. Essentially, this means that a cheap \$20 TV tuner USB dongle with the RTL2832U chip can be used as a computer based radio scanner. This sort of scanner capability would have cost hundreds or even thousands of dollars just a few years ago. The RTL-SDR is also often referred to as RTL2832U, DVB-T SDR, RTL dongle or the \$20 Software Defined Radio.

### 3.4.2 GNU Radio and Software-Defined Radio

We have used GNU Radio [56] and MATLAB software in the thesis to implement the cooperative spectrum sensing for hard decision and soft decision schemes. GNU Radio is an open source development toolkit that provides re-configurable signal processing blocks to implement and test out software-defined radios and signal processing systems. GNU Radio allows for SDR developers to develop unique signal processing blocks and SDR systems. GNU Radio was started in 2001, originally forked from the SpectrumWare project developed at the Massachusetts Institute of Technology. Since 2001, the code base has undergone

massive changes, containing almost no code from the original SpectrumWare project [57]. Physically the code consist of three languages Python, C++, and SWIG. Python provides the overarching control of the system or program, while C++ provides the actual signal processing blocks and mathematics. SWIG is a wrapper for C++ which allows Python to dynamically wrap around C++ and control or compile with it. A diagram below better illustrates this architecture. It is also important to mention that there as significant paradigm shifts in the community, pushing more and more code to Python rather than C++, due to its easier programming syntax and structure [58].

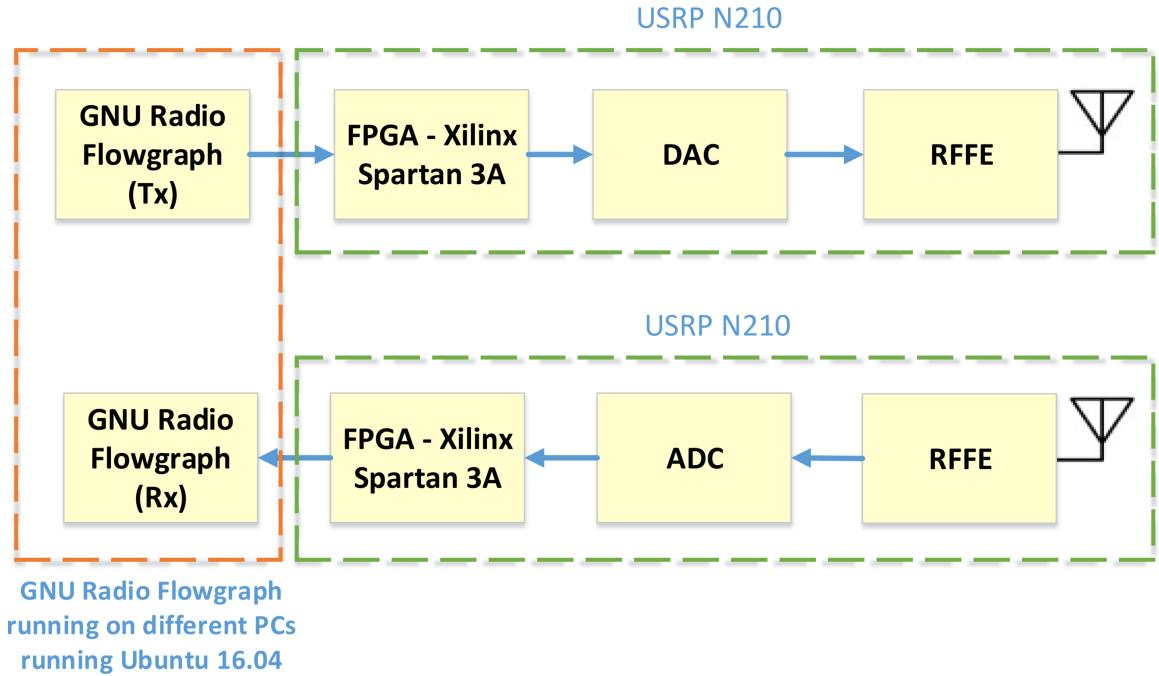


Figure 3.3: GNURadio and Software-Defined Radio

The GNU Radio software provides the framework and tools to build and run software radio or just general signal-processing applications. The GNU Radio applications themselves are generally known as "flowgraphs", which are a series of signal processing blocks connected together, thus describing a data flow. GNU Radio provides a very structured framework of flow design. Data processing segments are extremely self contained to minimize error propagation during system debugging. Since the software is open-source full access to all code is provided, giving low-level access to all operation within GNU Radio. Much of the

actions have been abstracted to the limited knowledge of the lower layers, but if specific actions are required for an application then serious depth or knowledge is needed about the overall projects structure, which is quite overloading.

### 3.4.3 MATLAB

MATLAB [59] is an extremely well known engineering, mathematical, biological, and financial software suite. MATLAB provide massive data leverage and advanced communication system models and algorithm for significant data processing.

## 3.5 Summary

This chapter outlined and examined the topics of jamming and anti-jamming techniques, and provided a foundation in communication system theory and advanced equalizer design. Secondly it setup an understanding of Software-Defined Radio, the power of such an architecture, and examples of implementations and existing software for future designs. Next, this thesis will consider a new anti-jamming technique and design an implementation of such a system. After the implementation is investigated, the result of specific experiments on such an implementation will be analyzed.

## Chapter 4

# Proposed LTE-R Test-bed Implementation and Results

This chapter provides background information on LTE proposed for high-speed railway communication system. It outlines the architecture for LTE-R leveraging the existing LTE network for high speed railway networks. For uniform coverage inside a tunnel environment, leaky coaxial cables (LCX) has been proposed for efficient communication. We explain LCX cables in details in the following chapter especially for tunnel environment. The severe channel impairments inside a tunnel are also discussed with special attention to high Doppler shift cause due to high velocities of trains and harsh multipath fading environment. Finally, the proposed two-ray propagation channel model is discussed and dynamic K-factor is derived.

### 4.1 LTE-R Testbed in Matlab

We have implemented the simulation testbed in MATLAB, consisting of a transmitter, a channel and a receiver. The K-factor values for the channel model are obtained from Eq.(4.10) and are used to generate a time-series BER curve for different modulation schemes used in LTE-R. The values used for the electrical material properties for tunnel walls [60] and its specifications are given in Table 4.1. The relative permittivity for the tunnel walls is

Table 4.1: Tunnel and Tx/Rx Characteristics

	Dimensions	Simulation Parameters
Tunnel	Width = 8.6 m, Height = 7.3 m	$\varepsilon_r = 5, \sigma = 0.1 \text{ Sm}^{-1}$
Leaky Feeder Cable (Tx)	Height = 6.1 m	$f_c (\text{GHz}) = 2, 3, 5$
Train (Rx)	Height = 4.2 m	$v (\text{km/h}) = 300, 400, 500$

taken as  $\varepsilon_r = 5$  and  $\sigma$  is set to 0.1. The simulation is conducted for velocity of  $v(\text{Km/h}) = 300, 400$  and  $500 \text{ Km/h}$ . Since the frequency band allocation for LTE-R will most probably be from 2–6 GHz, hence the  $f_c$  values chose are 2, 3 and 5 GHz. The height of the receiver is assumed to be around the length of the train and height of tunnel is chosen as the size of the leaky coaxial cable.

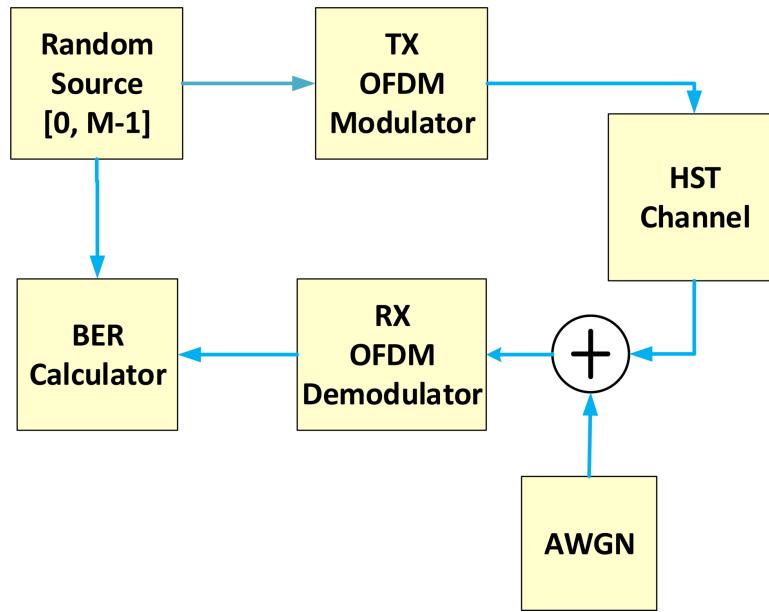


Figure 4.1: Block diagram of a communication system through a HST channel using QPSK, 16-QAM and 64-QAM.

Figure ?? shows the block diagram of the simulation test-bed used for the performance analysis of LTE-R. The random source block generate the symbol between 0 and M-1, where  $M = 4, 16, 64$  for QPSK, 16QAM and 64QAM respectively. The data is then modulated

with specific modulation type and then pass through the proposed HST channel model. The additive white gaussian noise is added after applying the channel coefficients to the signal data. We then demodulate the data packets and pass it to the ber calculator object which then computes the bit-error rate. The simulations are run for SNR values ranging from 0–20 dB and plots are generated for all three modulation schemes.

## 4.2 HST LTE-R in a Tunnel

Using LTE System Toolbox provided by MATLAB we generated Figure ?? which shows the received resource grid without equalization. The frame worth of data was modulated with QPSK, 16QAM and 64QAM for equal number of subcarriers and mapped to symbol in a subframe. We generate ten subframes individually and create one frame after merging all subframes. The frame is passed through our proposed high speed train channel model, with additive white gaussian noise added. We can see that without equalization the received resource grid has lot of errors and will lead to numerous retransmissions.

**Received OFDM Signal Under HST Fading Channel**

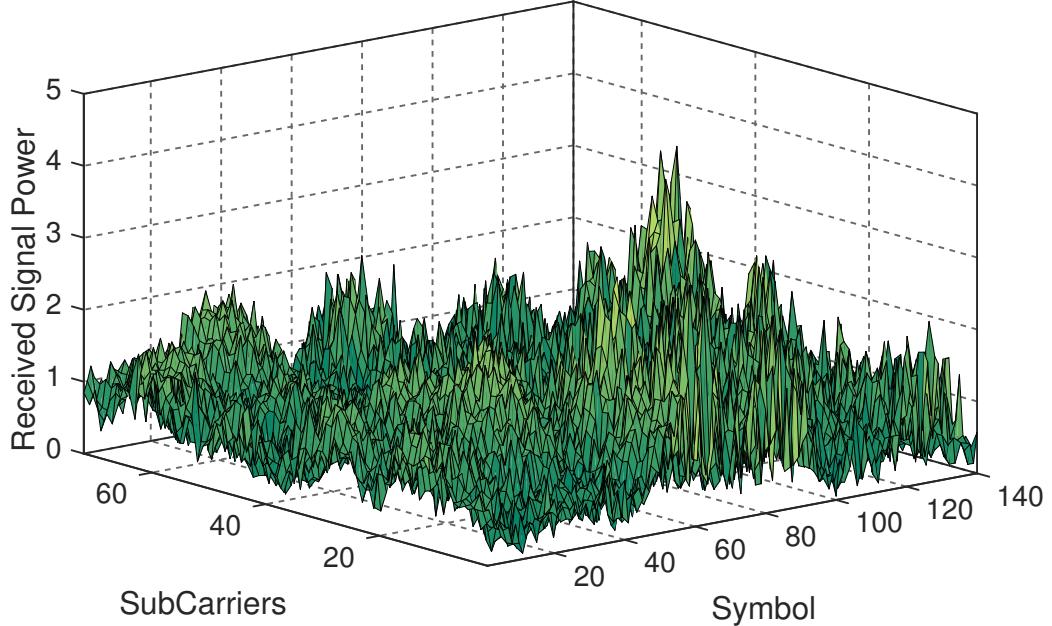


Figure 4.2: Received LTE-R OFDM signal under HST Ricean Fading Environment.

#### 4.2.1 K-factor in a Tunnel

In Figure ??, we calculated the K-factor for the HST inside a tunnel with velocity  $v = 500$  km/h for different center frequencies. It shows the variation of the Rician K-factor with the distance between the transmitter and receiver as the train is moving along the tunnel. We computed the K-factor for a leaky cable with periodic slots separated by distance  $d$  in fixed time-steps. The plot shows that the K-factor varies significantly over a short distances. Therefore, assuming a single K-factor for the channel model is not accurate, we use time-series K-factor to do our channel modeling.

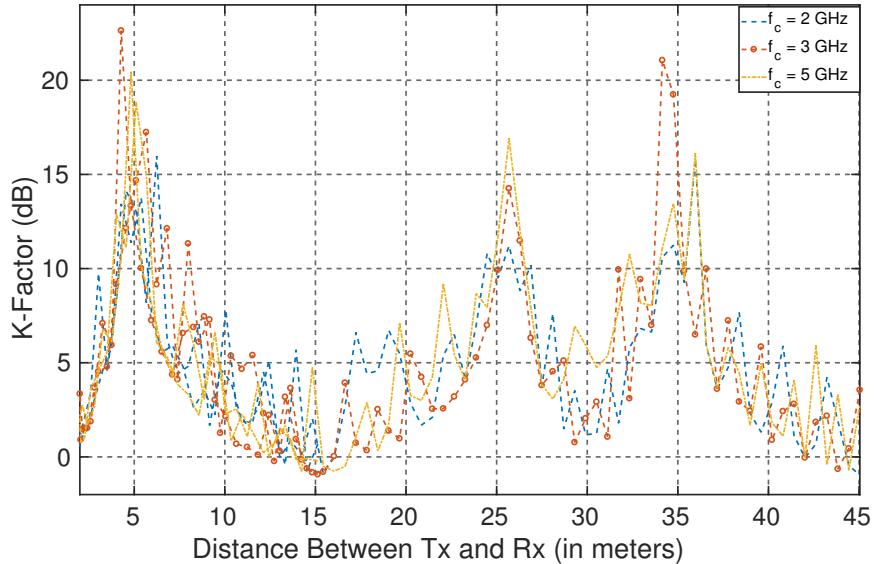


Figure 4.3: K-factor versus  $D_{LOS}$  for different center frequencies  $f_c = 2, 3$  and  $5$  GHz.

#### 4.2.2 BER Performance

To show the impact of varying K-factor on the channel, we computed the BER curve for different modulation schemes of LTE-R with different K-factors. Fig. 4.4a shows the BER versus SNR performance for QPSK modulation for different K-factors of the tunnel channel model. The figure clearly demonstrates for higher K-factor we have a better performance while performance degrades as K-factor goes low. Fig.4.4b shows the  $E_b/N_0$  versus BER for 16-QAM and as we can see the BER is higher compared to QPSK. Fig. 4.4c shows the

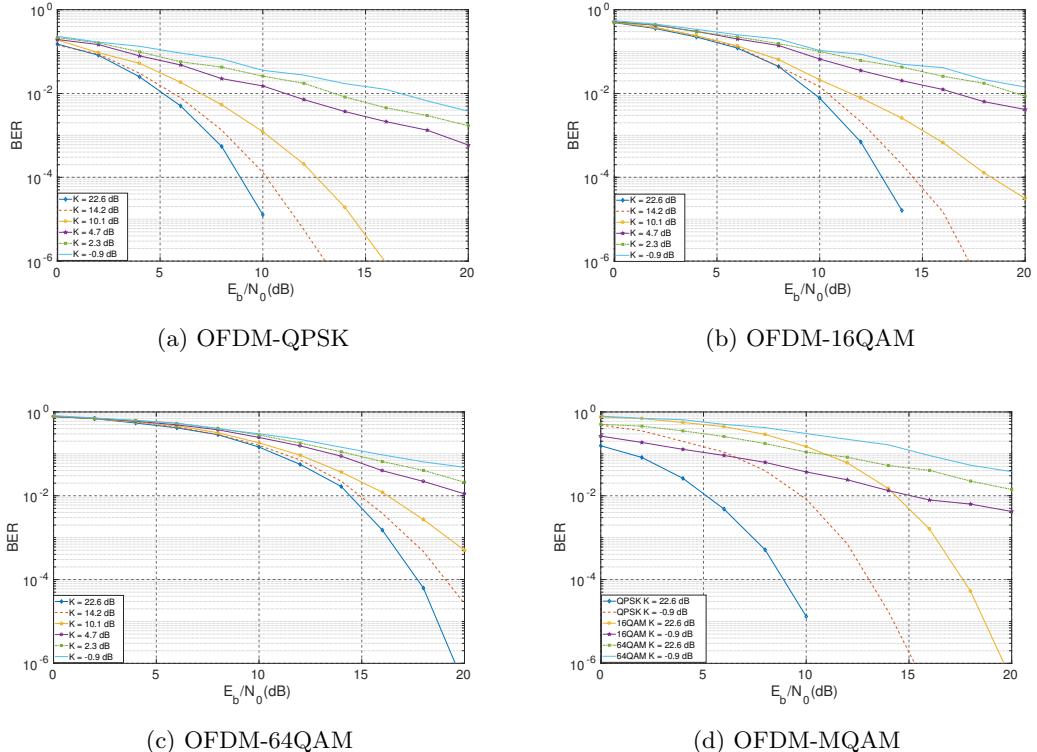


Figure 4.4: Comparison of  $E_b/N_0$  verus BER for LTE-R OFDM modulation with different  $K$ -factors. The first three sub-figures shows the  $E_b/N_0$  versus BER for individual modulation schemes employed in LTE-R and in last plot we compare all the modulation schemes for different  $K$ -factors.

$E_b/N_0$  versus BER for 64-QAM for different K-factors. And finally we compare all the modulation schemes for the best and worst K-factor in Fig. 4.4d.

In Fig. ?? we calculate the BER performance for a high speed train in discrete time-steps. As the train moves towards the LCX slot the SNR goes high and the SNR decreases as the train moves away. This trend can be seen in the plot, as we move towards the slot the BER curve decreases and it starts increases once we move. It is important to consider here that due to the varying nature of K-factor the BER curve also varies significantly. Hence, by considering the time-varying nature of K-factor we can have a better performance analysis.

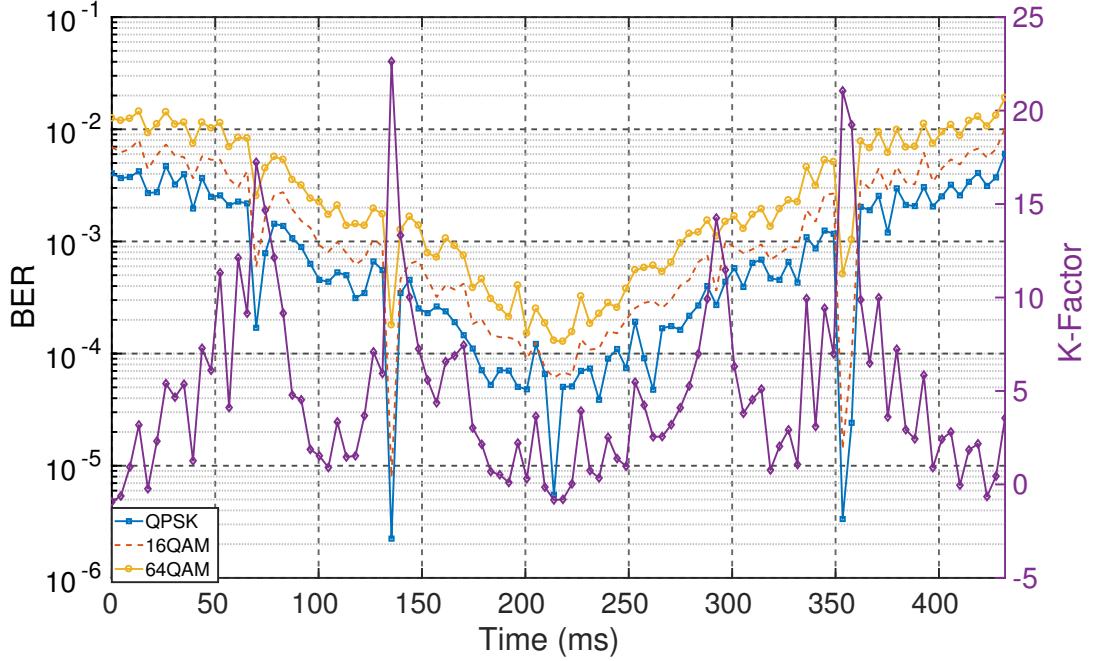


Figure 4.5: BER variation with time for HST with different modulation schemes of LTE-R. As the train moves towards the antenna the general trend of BER goes down with small-scale fluctuations due to varying K-factor.

### 4.3 Channel Impairments Inside a Tunnel

The tunnel environment is affected by multipath and diffraction effects due to multiple reflections from the tunnel walls, which leads to a substantial fading environment. By deploying LCX cables, we can eliminate the large penetration loss due to tunnel walls. However, small-scale fading can still cause a large amount of errors and decrease the QoS for the communication link. High velocity trains experience very high Doppler shifts and a fast fading channel. These problems can lead to significant BER degradation of the LTE system. The frequency shifts caused by the Doppler phenomenon can lead to shifts in the sub-carrier frequencies for OFDM, which leads to synchronization errors. The maximum Doppler shifts for a train traveling at 500 km/h is 2.314 kHz for a 5 GHz carrier frequency. This large Doppler shift can also lead to significant drops in the quality of wireless signals and increase the bit error rate. Thus, to develop an efficient and reliable communication link

inside tunnels, we need to properly model this channel impairment and build our proposed channel model by taking into account these tunnel phenomena. These impairments are described in detail in the following subsections.

### 4.3.1 Multipath Fading

The following time-varying multipath channel impulse response considers the effects of Doppler shift and scattering [61]:

$$h(\tau, t) = \sum_{k=0}^L h_k(t) e^{-j2\pi f_c \tau_k(t)} \delta[\tau - \tau_k(t)], \quad (4.1)$$

where  $\tau$  is the path delay,  $t$  is time in seconds,  $\delta[\tau - \tau_k(t)]$  is the impulse response,  $f_c$  is the carrier frequency,  $h_k(t)$  is the envelope of the time-varying channel and consists of both large and small-scale fading components. Since the structure of LCX is almost the same as a leaky waveguide, the large scale fading of channel can be modeled linearly [62]. There is also no signal shadowing and the line-of-sight (LOS) signal component is always present along the tunnel. This type of channel fading can be best described by a Ricean fading model. The probability density function  $p(\alpha)$  of a Rician fading model is given by [63]:

$$p(\alpha) = \frac{2\alpha(1+K)}{\Omega} I_0\left(2\alpha\sqrt{\frac{K+K^2}{\Omega}}\right) e^{\frac{-K-\alpha^2(1+K)}{\Omega}}, \quad (4.2)$$

where  $K$  is the Rician factor and  $\alpha$  is the complex amplitude of the channel response function that has a unity second moment, *i.e.*,  $\Omega \equiv E[\alpha^2] = 1$ .

### 4.3.2 Doppler Shift

The 3GPP channel model [64] is used for its Doppler shift profile in high speed railway environment. The measurements obtained for the Doppler frequency shift are implemented for two scenarios. The first scenario is for an open space while the second scenario is for high speed trains. Doppler shift is not taken into consideration. There exists a third scenario for tunnels using multiple antennas. Since the slots of LCX can be modeled as multiple antenna system, we use this Doppler shift profile for our proposed channel. The Doppler shift variation  $f_s(t)$  is described by:

$$f_s(t) = f_d \cos \theta(t), \quad (4.3)$$

where  $f_d$  is the maximum Doppler shift,  $\theta$  is the elevation angle and  $\cos \theta(t)$  is given by:

$$\cos \theta(t) = \begin{cases} \frac{D_s/2 - vt}{\sqrt{D_{\min}^2 + ((D_s/2) - vt)^2}}, & 0 \leq t \leq \frac{D_s}{v} \\ \frac{-1.5D_s + vt}{\sqrt{D_{\min}^2 + ((-1.5D_s) - vt)^2}}, & \frac{D_s}{v} \leq t \leq \frac{2D_s}{v} \end{cases} \quad (4.4)$$

where  $D_s/2$  is the initial distance of the train from base-station, and  $D_{\min}$  is base-station (BS)-Railway track distance, both in meters;  $v$  is the velocity of the train in m/s,  $t$  is time in seconds.

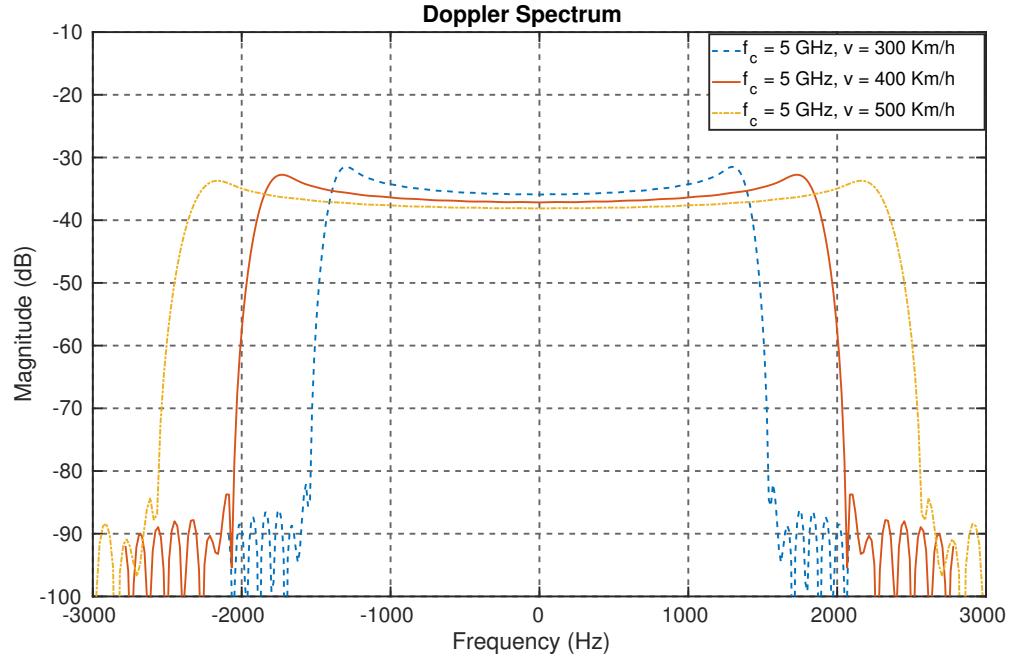


Figure 4.6: Doppler spectrum for LTE-R at different train velocities  $v$  (km/h) = 300, 400 and 500 and  $f_c = 5$  GHz.

Figure ?? shows the Doppler spectrum for  $f_c = 5$  GHz and  $v = 300 \text{ Km/h}, 400 \text{ Km/h}$  and  $500 \text{ Km/h}$ , and as we can see in the figure the maximum Doppler shift range is from -

2.314 kHz to +2.314 kHz. These range of Doppler shift values can lead to very high bit-error rate and poor connectivity in communication system. In the following section we discuss our proposed channel model which consists of Doppler shift profile for high speed train and dynamic K-factor for tunnel environment.

#### 4.4 Proposed Channel Model

The tunnel measurement campaign conducted in [20] shows that the amplitude variation inside tunnel follows Rician distribution. In the thesis, we apply the approach used in [65] for single elevation angle  $\theta$  and expand it to a time-varying case. In this thesis, we model  $\theta$  as a function of time and derive time-series  $K$ -factor for the tunnel environment. Figure ?? describes our proposed channel model which is implemented using dynamic K-factor and Doppler shift profile derived using Eq.(4.3). In the following section we discuss the classical two-ray propagation model and mathematical derivation of dynamic K-factor for our proposed channel model.

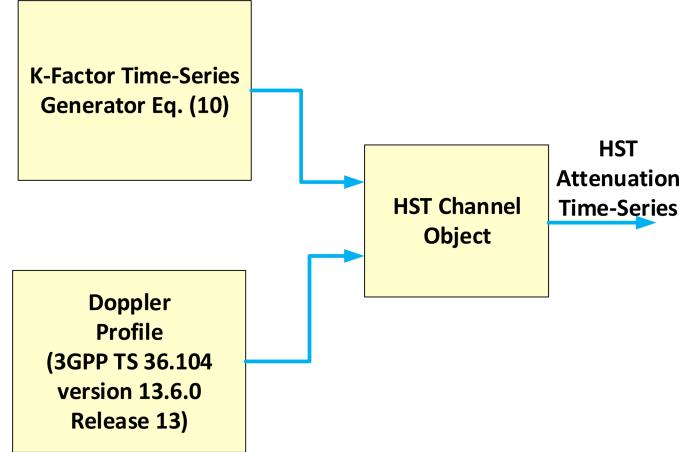


Figure 4.7: HST channel model consisting of time-series K-factor and Doppler shift caused due to velocity of the train.

The reflection coefficient  $\Gamma$  [66] as a function of time  $t$  is given by:

$$\Gamma(t) = \frac{C \sin \theta(t) - \sqrt{(\varepsilon_r - j\chi(t)) - (\cos \theta(t))^2}}{C \sin \theta(t) + \sqrt{(\varepsilon_r - j\chi(t)) - (\cos \theta(t))^2}}, \quad (4.5)$$

where  $C = 1$  is for horizontal polarization, and  $C = \varepsilon_r - j\chi(t)$  for vertical polarization. Furthermore,  $\chi(t)$  is given by:

$$\chi(t) = \frac{\sigma}{\omega(t)\varepsilon_0} = \frac{\sigma}{2\pi f_r(t)\varepsilon_0} = \frac{1.8 \times 10^{10}\sigma}{f_r(t)}. \quad (4.6)$$

with  $\varepsilon_0 = 8.854 \times 10^{-12}$  F/m, and  $\sigma$  is conductivity of the tunnel. The frequency  $f_r(t)$  is the resultant frequency caused by the Doppler shift and is given by:

$$f_r(t) = f_c(t) - f_s(t) \quad (4.7)$$

where  $f_c(t)$  is the sub-carrier frequency, and  $f_s(t)$  is the Doppler shift given by Eq. (4.3).

The phase difference function of  $t$ ,  $\Delta\phi(t)$  between the two reflected paths is given by [24]:

$$\Delta\phi(t) = \frac{2\pi}{\lambda(t)} \left( \sqrt{D_{\text{LOS}}^2 + (h_t + h_r)^2} - \sqrt{D_{\text{LOS}}^2 + (h_t - h_r)^2} \right), \quad (4.8)$$

where  $\lambda(t)$  is the resultant time-varying wavelength at the receiver,  $D_{\text{LOS}}$  is the distance between the transmitter and receiver antennas which is changing dynamically with  $t$ , and both  $h_t$  and  $h_r$  are the heights of the transmitter and receiver antennas, respectively.

The resultant received power  $p_r(t)$  is given by the sum of the LOS received power plus the received multipath power, resulting in:

$$p_r(t) = p_t(t) \left( \frac{\lambda}{4\pi d} \right)^2 G_t G_r \left[ 1 + |\Gamma(t)|^2 + 2|\Gamma(t)| \cos(\angle\Gamma(t) - \angle\Delta\phi(t)) \right] \quad (4.9)$$

which is a function of the transmitter power  $p_t(t)$  and the reflection coefficient  $\Gamma(t)$ , where  $G_t$  and  $G_r$  are the transmitter and receiver antenna gains, respectively. The  $K$ -factor is defined as the ratio of the direct path power and the power in the scattered paths, and is given as:

$$K(t) = \frac{1}{|\Gamma(t)|^2 + 2|\Gamma(t)| \cos(\angle\Gamma(t) - \angle\Delta\Phi(t))} \quad (4.10)$$

## 4.5 Summary

We analyzed the BER performance of a LTE-R system for high speed trains inside tunnel environments using our proposed channel model. For the implementation of our channel,

we first derived the time-series K-factor function using the classical two-ray propagation model. We then analyzed the LTE-R performance under our channel model for different modulation schemes for various K-factors. Finally, we compared all the modulation schemes under worst and best K-factor, and we observed that for low  $E_b/N_0$  sub-carriers must be modulated with QPSK for maintaining reliable communication link.

## Chapter 5

# CSS Implementation and Results

### 5.1 Overview

This chapter outlines the implementation of heterogeneous test-bed for cooperative spectrum sensing using software-defined radios N210, and RTL-SDR and we also describe the simulation test-bed for performance analysis of LTE-R communication system in a tunnel environment. We start with heterogeneous cooperative spectrum sensing where we first describe the experimental setup which is implemented using soft and hard fusion schemes. The sensor nodes which consists of three RTL-SDRs and one USRP N210 are placed in a controlled indoor laboratory environment around 8-10 meters apart. The signal source is being simulated by another USRP N210 which is transmitting a DQPSK signal at 450 MHz. The post-processing is done on a Fusion Center which makes the decision based on global test statistic using both soft and hard decision schemes. The performance of both the schemes are evaluated in a real fading scenario on a hardware test-bed.

We then discuss our LTE-R test-bed implementation in MATLAB and compare the performance for QPSK, 16QAM and 64QAM modulation schemes in a tunnel environment. First we show the results of K-factor variation in a tunnel environment for a high speed train in a tunnel. Using the K-factor we then implement our channel model which also takes into effect the high Doppler shift due to mobility of the train. The bit-error rate curves are generated for the LTE-R modulation schemes for the proposed channel model for high

speed train in a tunnel. Finally, we show a time-varying BER curve for high speed train moving with the velocity  $v$  at discrete timesteps.

## 5.2 Experimental Setup for Heterogeneous CSS

The measurements are performed using software-defined radios (SDRs) and the post processing is conducted on desktop computers. The desktop computer consists of i7 Intel processor with eight cores and 3.41 GHz clock cycle running Ubuntu 16.04. The sensor node network is implemented using RTL-SDR dongles and Ettus Research USRP N210 on GNU Radio Software platform. The measurements are analyzed in MATLAB and measurement plots are generated. Figure 5.1 consists of three RTL-SDRs and two USRP N210s. One USRP N210 in the middle acts as a primary user and other SDRs are sensor nodes. All the SDRs were placed in the lab atleast 5-6 meters away from the primary user.

These sensor nodes collect the spectral data, normalize it and then transmit it to the FC for the detection. For soft data fusion, the data is quantized in the local sensor nodes before it is transmitted to FC due to the limited bandwidth of the overhead channel. The delays caused by different sensor nodes is ignored, as it would require extra computational complexity and it is out of the scope of this thesis. The USRP N210 transmits a DQPSK modulated signal with 4 samples per symbol with the alpha factor of the root raised cosine filter set to 0.35. The transmitter gain and amplitude are varied to get different SNR values for each node. The sensor nodes collect the data for the equivalent of 300,000 energy samples, and each measurement is conducted three times to eliminate any irregularities. The noise variance  $\sigma_n^2$  for each SU is estimated by running each sensor node without any transmission at 450 MHz. The flow-graph is executed multiple times to get a better estimate of noise variance. Once the data is received from all the sensor nodes, Probability of detection is calculated for different received SNR values for all the sensor nodes. To properly evaluate the performance of each of the cooperative spectrum sensing techniques, the average  $P_d$  is calculated for each scheme.

All four sensor nodes have different sampling rates to truly model the heterogeneous environment. The USRP N210 which is also used as a 4<sup>th</sup> sensor node has a very low noise

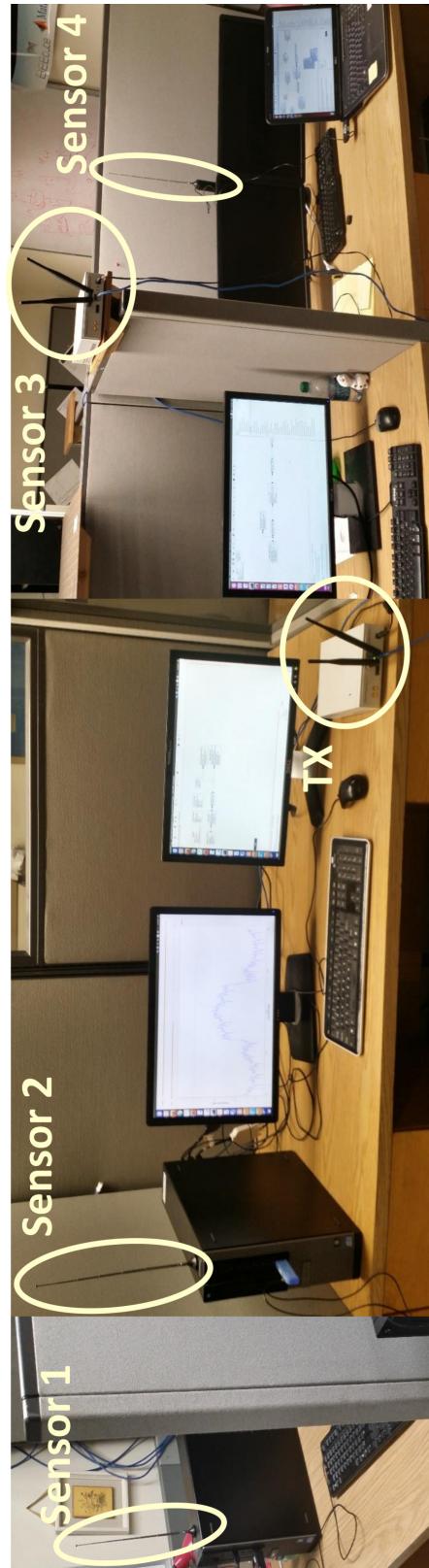


Figure 5.1: Experimental Test-Bed For Cooperative Sensing in Heterogeneous Network. Sensors 1, 2 and 4 are RTL-SDR units, sensor 3 is USRP N210 and TX is another USRP N210 unit which is used as a signal source for this work.

floor compared to the three other and hence can detect a signal with very low SNR. This is a challenging factor for data fusion when the nodes have different operating parameters and FC has to make optimal decisions by combining this varying data. And due to their spatial diversity the node closest to the transmitter will have different SNR compared to the other nodes. All these factors impact the data combining at FC. The GNURadio flow-graph for the sensor nodes is shown in the Figure 5.3. The same flow-graph is used for all sensor nodes with different operating parameters. For USRP N210 node, we replaced the RTL-SDR source with UHD:USRP Source gnuradio block. The central frequency is kept at 450 MHz and the operating parameters of each sensor node is provided in the table. The values of the transmitter amplitude and gain are varied to get the different sets of SNR values which are used for computing  $P_d$  values for each node. The plots are generated in MATLAB by using the data files from gnuradio platform.

### 5.2.1 Sensor Node and Signal Source Setup

To evaluate the performance of the cooperative spectrum sensing, the USRP N210 is used as a transmitter where its gain and amplitude are varied. Figure 5.2 shows the flowgraph used for the transmitter. The flowgraph starts with the random source block which generates the random data between 0 and three, and pass it to the differential phase shift keying (DPSK) modulator block. The DPSK block modulates the signal with differential quadrature phase shift keying (DQPSK) scheme, apply root raise cosine filter with excess bandwidth value of 0.35 and pass it to multiply const block. It is used to control the SNR value and finally the data is dumped into USRP N210 sink which transmits the data over the air where other sensor nodes can estimate the signal presence using soft and hard decision schemes.

The sensor nodes are placed inside the laboratory and are connected to distributed system. Figure 5.3 shows the flowgraph running on the receiver, for flowgraph running on USRP we use UHD:USRP source instead of RTL-SDR source. The frequency around 450 MHz is swepted in regular intervals and the continuous data stream is passed to FFT block which does the forward FFT operation with blackmann harris window. The data is first converted into parallel stream of the FFT size using stream to vector gnu radio

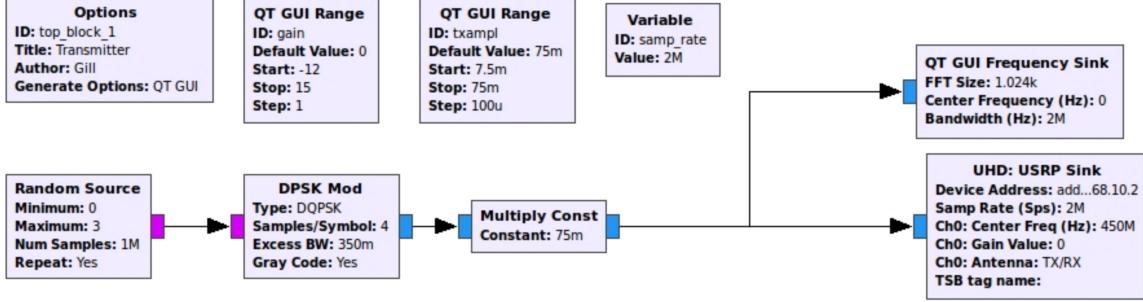


Figure 5.2: GNU Radio Flowgraph For Transmitter Running on USRP N210.

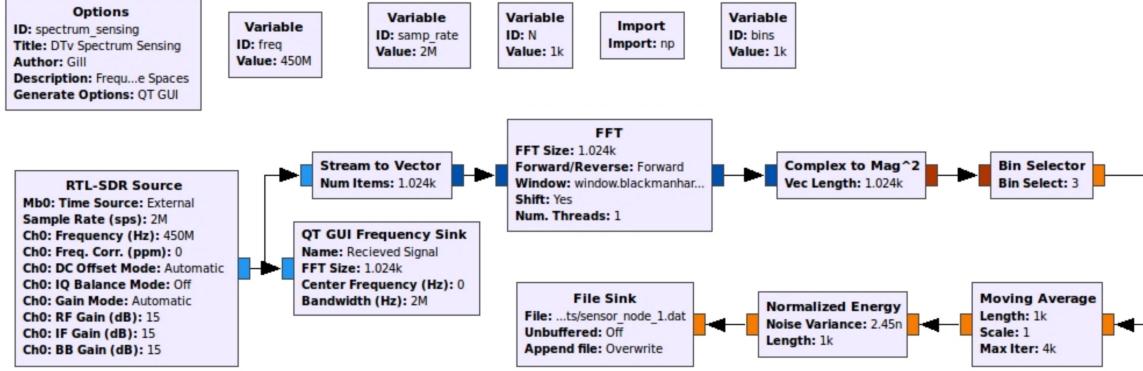


Figure 5.3: GNURadio Flowgraph For USRP and RTL-SDR sensor nodes.

Table 5.1: Operating Characteristics of Sensor Nodes

Nodes	$F_s$	Gain	FFT Size	Bin Size
RTL-SDR-1	1.1 Msps	10 dB	512	2.148 KHz
RTL-SDR-2	1.8 Msps	10 dB	512	3.515 KHz
RTL-SDR-3	2.4 Msps	10 dB	512	4.687 KHz
USRP N210	7.2 Msps	10 dB	512	14.062 KHz

block. We complex to magnitude block converts the complex values into float and take their magnitude. The bin selector block is used to select the bin where the narrow-band signal is being transmitted by the transmitter. We then take the moving average of the values, normalize them and then store it to the file sink. The normalized energy values are collected for each sensor at different SNR values and then post processing is performed in

MATLAB. The operating parameters of each sensor node is provided in the table 5.1.

### 5.3 Hard-Data Fusion Scheme

Cooperative spectrum sensing using hard-data fusion is a proven method for improving the detection performance. In this scheme, all sensor nodes sense the signal source individually and send their sensing decision in the form of 1-bit binary data. For hard data fusion, the noise  $w_q$  can be neglected since the sensor nodes can just transmit their decision statistic in an efficient way, where the floating values are not required. For example, the SUs can just transmit "1" and "0" depending on whether the primary user is present or absent. Furthermore, in the Fusion Center the decision can be made by using the OR, AND, or majority rule algorithms. For the AND decision rule, the FC performs the logical AND operation for all the local decisions and conducts the detection. Similarly, for the OR rule, the logical OR operation is used to decide whether the signal is present or not. Finally, the majority rule conducts majority vote and decides based on it [67, 68]. The  $P_d$  for AND, OR and Majority Rule for R=4 sensor nodes is given by [69]:

$$\begin{aligned} P_{d,AND} &= (P_d)^4 \\ P_{d,OR} &= 1 - (1 - P_d)^4 \\ P_{d,MJR} &= 6P_{davg}^2(1 - P_{davg})^2 + 4P_{davg}^3(1 - P_{davg}) + P_{davg}^4 \end{aligned} \quad (5.1)$$

where  $P_{davg}$  is the average probability of detection of the sensor units. Similarly, we can calculate the  $P_{fa}$  for all three schemes by replacing  $P_{davg}$  by  $P_{faavg}$  in Eq (5.4).

### 5.4 Soft-Data Fusion Scheme

In soft-data fusion based cooperative spectrum sensing, information from different CR users is combined to make a decision on the presence or absence of the primary user. In 5.3 we discuss the conventional hard combination where each CR user feedbacks one-bit message regarding whether observed energy is above a certain threshold . In this section, we discuss soft combination of the local test statistic of each sensor nodes and how it is combined to make the decision in the FC. Since in the soft combination accurate energy values from different CR users are utilized to make a decision, this scheme is more accurate and complex

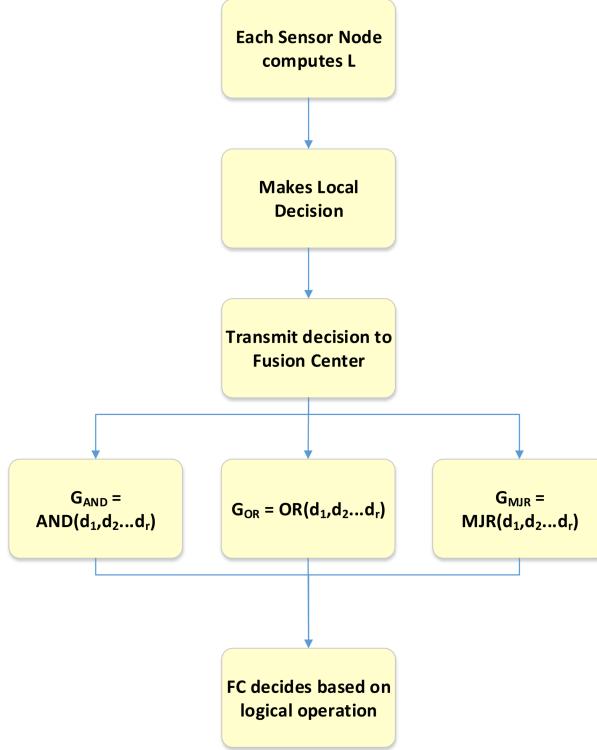


Figure 5.4: Flowchart showing AND, OR and Majority Rule Fusion schemes.

to implement. In this thesis, we discuss two popular soft decision fusion schemes: Maximum Normalized Energy (MNE) scheme and Equal Gain Combining (EGC) scheme.

#### 5.4.1 Maximum Normalized Energy Scheme

The local test statistic in each sensor node is computed and then transmitted to FC after quantization. In this thesis, we are using four sensor nodes equipped with different sensing abilities such as the sampling rates and noise floor. Therefore, the global test statistic  $G$  can be modeled by:

$$G_{MNE} = \max\{\beta_r\}, \quad (5.2)$$

The  $P_{fa}$  and  $P_d$  values for the MNE-CS is given by: [?],

$$P_{fa} = 1 - \prod_{r=1}^R \left( 1 - Q\left(\frac{\tau - 1}{\sqrt{\frac{1}{M_r} + \sigma_{q,r}}}\right) \right), \quad (5.3)$$

$$P_d = 1 - \prod_{r=1}^R \left( 1 - Q\left( \frac{\tau - 1 - \gamma_r}{\sqrt{\frac{1+2\gamma_r}{M_r} + \sigma_{q,r}}} \right) \right). \quad (5.4)$$

where  $\tau$  is the global threshold for MNE,  $M_r$  is the number of samples for  $r^{th}$  sensor node, and  $\sigma_{q,r}$  is the noise variance for the received local test statistic. The algorithm for MNE-CS is illustrated by the flowchart in Figure 5.5.

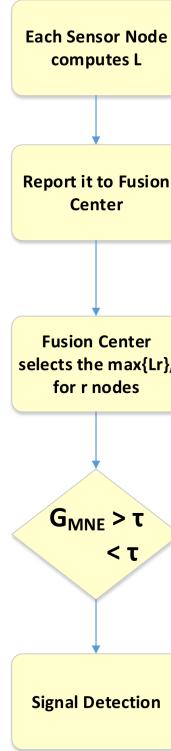


Figure 5.5: Flowchart describing Maximum Normalized Energy Scheme in detail.

#### 5.4.2 Equal Gain Combining Scheme

For EGC, the global decision statistic is the mean of the  $\beta$  values for all the sensor nodes. It has been shown in [19] that the EGC scheme performs better than the MNE scheme in a noisy channel. The EGC scheme can be modeled by:

$$G_{EGC} = \frac{1}{M} \sum_{r=1}^M \beta_r, \quad (5.5)$$

where  $G_{EGC}$  is global test statistic of EGC scheme. The  $P_{fa}$  and  $P_d$  values for the EGC-CS scheme are given by:

$$P_{fa} = Q\left(\frac{\tau - 1}{\sqrt{\frac{1}{R^2} \sum_{r=1}^R \left(\frac{1}{M_r} + \sigma_{q,r}^2\right)}}\right), \quad (5.6)$$

$$P_d = Q\left(\frac{\tau - \frac{1}{R} \sum_{r=1}^R (1 + \gamma_r)}{\sqrt{\frac{1}{R^2} \sum_{r=1}^R \left(\frac{1 + 2\gamma_r}{M_r} + \sigma_{q,r}^2\right)}}\right). \quad (5.7)$$

The algorithm for EGC-CS is also illustrated by the flowchart in Figure ??.

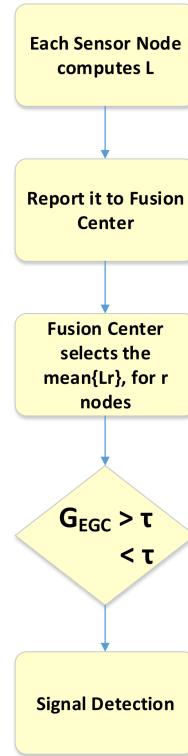


Figure 5.6: Flowchart describing Equal Gain Combining Scheme in detail.

## 5.5 Heterogeneous CSS Results

In this thesis, we implemented the heterogeneous cooperative spectrum sensing (CSS) using both hard and soft data fusion schemes. We start by collecting the data across 450

MHz band for all sensor nodes in a distributed manner. The spectrum sensing data is normalized for both soft and hard data fusion schemes using the same operational parameters to compare their performance accurately. The measurements are performed using software-defined radios (SDRs) and the post processing is conducted on desktop computers. The desktop computer consists of i7 Intel processor with eight cores and 3.41 GHz clock cycle running Ubuntu 16.04. The sensor node network is implemented using RTL-SDR dongles and Ettus Research USRP N210 on GNU Radio Software platform. These sensor nodes collect the spectral data, normalize it and then transmit it to the FC for the detection. For soft data fusion, the data is quantized in the local sensor nodes before it is transmitted to FC due to the limited bandwidth of the overhead channel.

### 5.5.1 Hard Decision Combining

The Figure 5.7 shows the  $P_{davg}$  versus  $SNR_{avg}$  for all four sensor nodes. It can be seen that OR performs the best while AND performs the worst in a fading channel. The SN R avg was computed by taking the mean of all the SNRs for the sensor nodes. The SNR was varied for each sensor node by varying the transmitter amplitude and gain in the gnuradio flowgraph.

In Figure 5.8 the ROC characteristics for the hard decision combining at two different  $SNR_{avg}$  for all three hard data fusion schemes are provided. It is pretty evident from the plot that OR scheme performs better than both AND and majority rule scheme. The AND scheme performs the worst because it depends on all sensor nodes to have same decision which is very difficult in a real fading environment. For lower SNR values OR outperform majority rule by a large margin but as we go to higher SNR values their performance converges.

### 5.5.2 Soft Decision Combining

The Figure 5.9 shows the  $P_{davg}$  vs  $SNR_{avg}$  for both soft and hard data fusion schemes. MNE and OR scheme overlap on the plot because in MNE scheme we take the maximum normalized energy and compare it the global test statistic, whereas in OR scheme we estimate the signal source by either of sensor node decision. This makes both the scheme almost

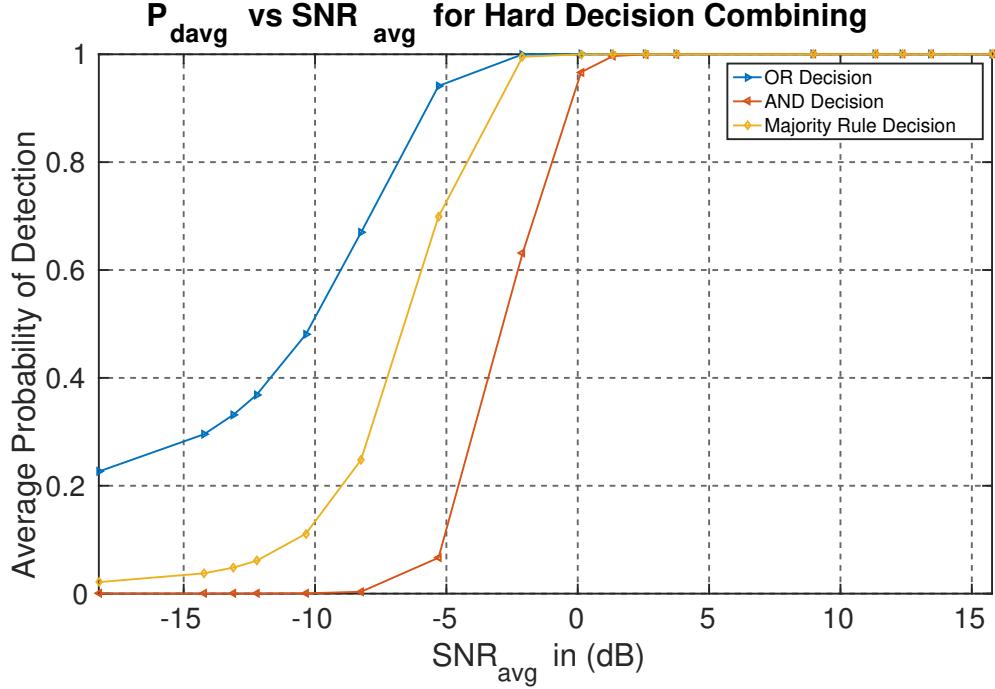


Figure 5.7: Probability of Detection versus  $SNR_{avg}$  For Hard Decision Combining.

same and this is visible in the results. The EGC scheme performs the best because it takes into consideration all the sensor nodes and its global test statistic gives equal weightage to all sensor nodes. AND scheme performs the worst as it was expected. It is very important to understand that at higher SNR values,  $SNR_{avg} \geq 2$  dB, we see all schemes converging to the same decisions. This tells us that in noiseless environment we can choose hard fusion schemes because of their implementation complexity and we can select soft fusion in severe fading environment as they tend to be more accurate in these scenarios.

## 5.6 Summary

This chapter outlined and examined the topics of jamming and anti-jamming techniques, and provided a foundation in communication system theory and advanced equalizer design. Secondly it setup an understanding of Software-Defined Radio, the power of such an architecture, and examples of implementations and existing software for future designs. Next, this thesis will consider a new anti-jamming technique and design an implementation of

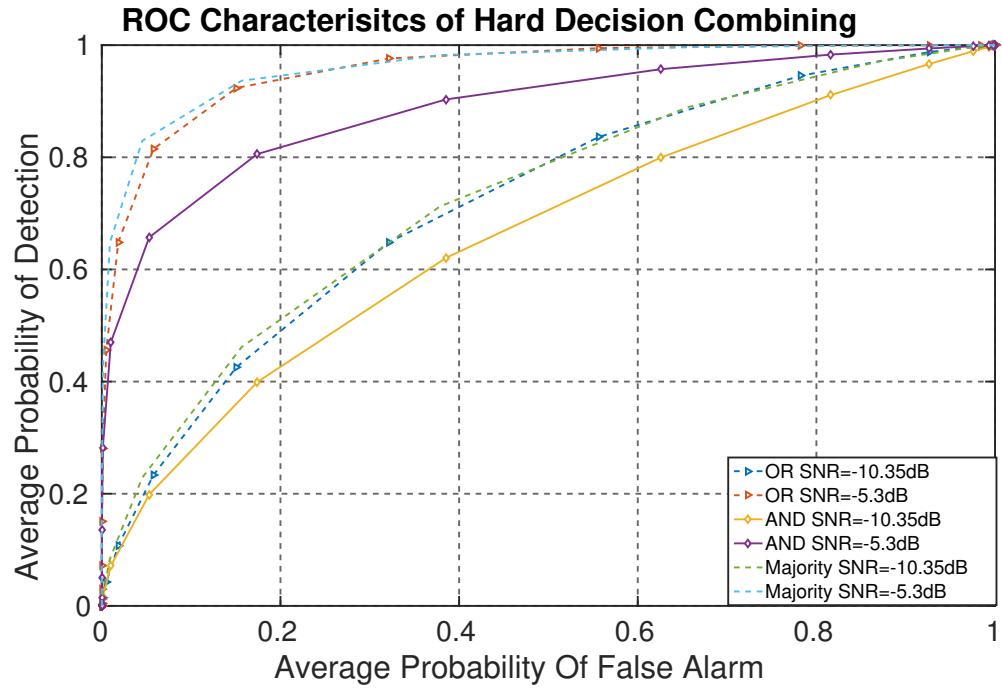


Figure 5.8: ROC Characteristics for Hard Decision Combining with Different SNRs.

such a system. After the implementation is investigated, the result of specific experiments on such an implementation will be analyzed.

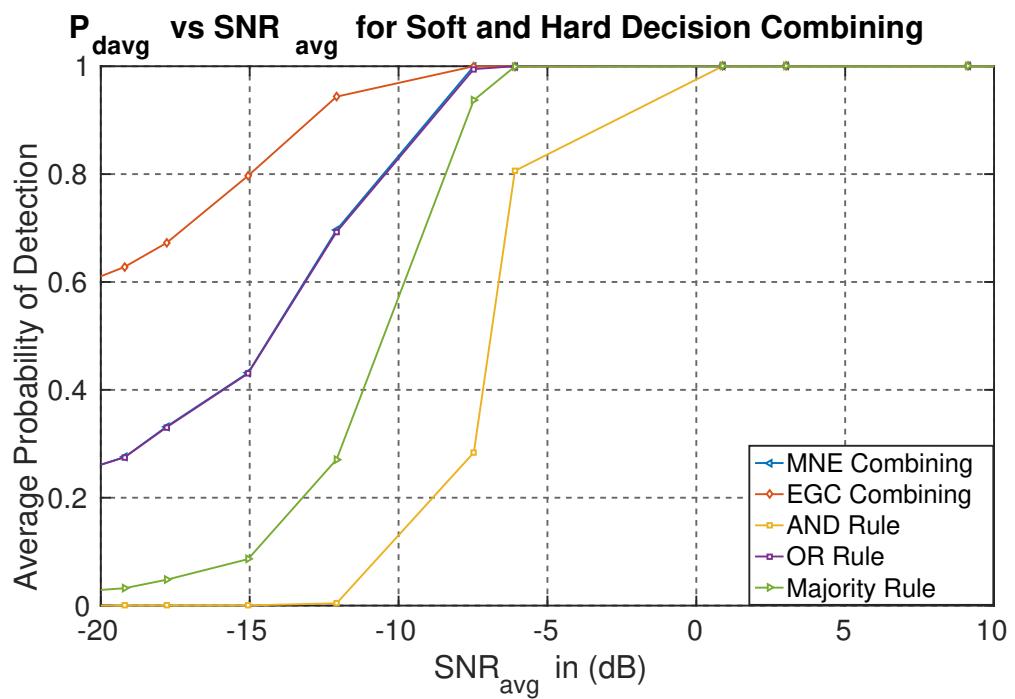


Figure 5.9: Probability of Detection versus  $SNR_{avg}$  For Soft and Hard Decision Combining.

# Chapter 6

## Conclusions

This chapter summarizes the work as part of this project and then suggests related research that can be performed in the future. The research achievements includes implementation of test-bed for cooperative spectrum sensing in heterogeneous network, the performance measure of both soft and hard data fusion schemes in a real fading scenario. The simulation test-bed is implemented in MATLAB to test the performance of LTE-R in our proposed channel. The proposed channel is built on two-ray propagation model with time-series K-factor which we have derived mathematically and also uses Doppler shift profile for high speed trains. The future work section describes how we can take the mobility effect into the cooperative spectrum sensing and then test the performance of soft and hard fusion schemes in a mobile scenario. For LTE-R for future work we will use LTE antenna toolbox provided by MATLAB to implement the simulation test-bed which is more realistic and close to the actual system.

### 6.1 Research Outcomes

- We analyzed the BER performance of a LTE-R system for high speed trains inside tunnel environments using our proposed channel model. For the implementation of our channel, we first derived the time-series K-factor function using the classical two-ray propagation model.

- We then analyzed the LTE-R performance under our channel model for different modulation schemes for various K-factors. Finally, we compared all the modulation schemes under worst and best K-factor, and we observed that for low  $E_b/N_0$  sub-carriers must be modulated with QPSK for maintaining reliable communication link.
- The last plot shows the BER curve for discrete time-step when the train is moving with a velocity of 500 Km/h and carrier frequency for all modulation scheme is set to 3 GHz. The plot is also overlayed with continuous K-factor variation with the propagation of the train. It can be observed from the plot that as the K-factor goes high the BER drops, which represents the train moving towards the LCX slot.
- As the train move away from the slot the BER starts increasing. For reliable and efficient communication links the sub-carriers have to be modulated with QPSK for low K-factor values, or more LTE repeaters are required inside the tunnel to get good connectivity. However, the most important factor that has to be taken into consideration is the real-time channel equalization to reduce the BER rate.

## 6.2 Future Work

- In this paper, we conducted an experimental study for cooperative spectrum sensing using normalized energy detection for both soft and hard decision combining techniques. It was found that the soft fusion schemes works better than hard decision for real fading environment with low SNR values.
- For higher values, all schemes converged to the same decision which led us to conclude that hard fusion schemes pays better when the environment is less noisy due to their low complexity as compared to soft fusion. For future work, it is worth exploring an increase in the number of nodes and adding mobility for the testing the performance of heterogeneous networks in a time-variant channel.

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## Appendix A

# Heterogeneous Cooperative Spectrum Sensing Code

### A.1 harddecisionpdoc.m

```
% Hard-Decision Combining Results For Sensor Nodes
clc;
close all;
clear all;
%% Parameter Initialization
N = 32;
k=4;%sensor nodes..
variance = 24.32e-9;
pfa = 0.05;
threshold = (qfuncinv(pfa)+sqrt(N)).*sqrt(N)*2*variance;
snrthreotical = -18:0.5:20;
snrlinear = 10.^ (snrthreotical/10);
%% SNR values from USRP and RTL-SDR
snrpracticalavg = [-18.23,-14.22,-13.1,-12.22,-10.35,-8.25,-5.3,-2.1,0.13,...1.34,2.58,3.76,8.98,11.33,12.35,13.45,15.77];
snrlinearprac = 10.^ (snrpracticalavg/10);
%% Computing Detection Probability and ROC Characteristics
```

```

pdprac = qfunc((threshold-2*N*variance.*(1+snrlinearprac))./...
    (sqrt(N.*(1+2*snrlinearprac))*(2*variance)));
pdpracor = 1-(1-pdprac).^4;
pdpracand = pdprac.^4;
tmp1 = (1-pdprac).^2;
tmp2 = (1-pdprac);
pdpracmjr = (6*pdprac.^2).*tmp1+(4*pdprac.^3).*tmp2+pdprac.^4;

pfapracor = 1-(1-pfa).^4;
pfapracand = pfa.^4;
tmp1 = (1-pfa).^2;
tmp2 = (1-pfa);
pfapracmjr = (6*pfa.^2).*tmp1+(4*pfa.^3).*tmp2+pfa.^4;

%% ROC Characteristics...
figure(1)
hold on;
grid on;
plot(pfapracor,pdpracor(:,5), '-->', 'LineWidth', 2, 'MarkerFaceColor', 'auto');
plot(pfapracor,pdpracor(:,7), '-->', 'LineWidth', 2, 'MarkerFaceColor', 'auto');
plot(pfapracand,pdpracand(:,5), '-d', 'LineWidth', 2, 'MarkerFaceColor', 'auto');
plot(pfapracand,pdpracand(:,7), '-d', 'LineWidth', 2, 'MarkerFaceColor', 'auto');
plot(pfapracmjr,pdpracmjr(:,5), '--', 'LineWidth', 2, 'MarkerFaceColor', 'auto');
plot(pfapracmjr,pdpracmjr(:,7), '--', 'LineWidth', 2, 'MarkerFaceColor', 'auto');
xlabel('Average Probability Of False Alarm');
ylabel('Average Probability of Detection');
title('ROC Characterisitcs of Hard Decision Combining');
hold off;
set(gca, 'fontsize', 30, 'box', 'on', 'LineWidth', 2, 'GridLineStyle', '--', 'GridAlpha'
    , 0.7);
lgd = legend('OR SNR=-10.35dB', 'OR SNR=-5.3dB', 'AND SNR=-10.35dB', ...
    'AND SNR=-5.3dB', 'Majority SNR=-10.35dB', 'Majority SNR=-5.3dB');
lgd.FontSize=20;

%% Probability of detection
figure(2)
hold on;

```

```

grid on;
plot(snrpracticalavg,pdpracor,'->','LineWidth',2);
plot(snrpracticalavg,pdpracand,'-<','LineWidth',2);
plot(snrpracticalavg,pdpracmjr,'-d','LineWidth',2);
xlabel('SNR_{avg} in (dB)');
ylabel('Average Probability of Detection');
title('P_{davg} vs SNR_{avg} for Hard Decision Combining');
hold off;
set(gca,'fontsize',30,'box','on','LineWidth',2,'GridLineStyle','--','GridAlpha'
,0.7);
lgd = legend('OR Decision','AND Decision','Majority Rule Decision');
lgd.FontSize=20;
axis([-18.23 15.77 0 1])

```

## A.2 softharddecisionpd.m

```

% Soft Decision Combining for sensor nodes...
%% Initializing parameters..
close all;
clear all;
N = [100,200,300,400];% Different sum factor
k=4;% Number of Sensor Nodes
variance = [24.025e-9,23.695e-9,25.678e-9,0.0323e-9];
pfa = 0.05;%Probability of false alarm
for i=1:4
    threshold(i) = (qfuncinv(pfa)+sqrt(N(i)))*sqrt(N(i))*2*variance(i);
end
% SNR Values from four sensor nodes
snrpractical = [-21.45,-18.23,-15.45,-13.3,-12.67,-9.35,-2.23,-4.32,2.98,6.95,13
.57,21.78;...
-22.23,-20.22,-17.34,-15.32,-13.45,-11.27,-7.75,-5.67,2.53,4.78
,12.67,20.32;...
-25.34,-23.34,-21.67,-20.33,-16.76,-13.38,-8.56,-6.53,0.38,1.34
,7.89,16.54;...
-27.32,-24.97,-22.34,-22.23,-17.34,-14.32,-11.35,-7.85,-2.35,-0

```

```

.98,2.38,5.98];

for i=1:4
    snrlinearprac(i,:) = 10.^ (snrpractical(i,:)/10);
end
for i=1:4
    pdprac(i,:) = qfunc((threshold(i)-2*N(i)*variance(i).* (1+snrlinearprac(i,:)))./
    ...
    (sqrt(N(i)*(1+2*snrlinearprac(i,:)))*(2*variance(i))));
end
for i=1:12
    snravg(i) = mean(snrpractical(:,i));
end
%% MNE based CS..
pdpracmne = 1-(1-pdprac(1,:)).*(1-pdprac(2,:)).*(1-pdprac(3,:)).*(1-pdprac(4,:))
;
pdpracand = mean(pdprac).^4;
pdpracm = mean(pdprac);
pdpracor = 1-(1-pdpracm).^4;
tmp1 = (1-pdpracm).^ (k-2);
tmp2 = (1-pdpracm);
pdpracmjr = (6*pdpracm.^ (k-2)).*tmp1+(4*pdpracm.^ (k-1)).*tmp2+pdpracm.^ k;
figure(1)
hold on;
grid on;
plot(snravg,pdpracmne,'-<','LineWidth',2,'MarkerFaceColor','auto');
axis([-20 10 0 1])
%% EGC based CS..
snrlinearmean = 10.^ (snravg/10);
snrlinear = 10.^ (snrpractical/10);
pfa = 0.01;
M= mean(N);
threshold = mean(threshold);
for i=1:length(snravg)
    num(i) = threshold-snrlinearmean(i);
    den(i) = (1/16)*((1+2*snrlinear(1,i))/N(1)+variance(1)+(1+2*snrlinear(2,i))/N(2)+variance(2)+(1+2*snrlinear(3,i))/N(3)+variance(3)+(1+2*snrlinear(4,i))/N(4)+variance(4));
end

```

```

pdegc(i) = qfunc(num(i)/sqrt(den(i)));
end

%% Plotting the data..
plot(snravg,pdegc,'-d','LineWidth',2,'MarkerFaceColor','auto');
plot(snravg,pdpracand,'-s','LineWidth',2,'MarkerFaceColor','auto');
plot(snravg,pdpracor,'-s','LineWidth',2,'MarkerFaceColor','auto');
plot(snravg,pdpracmjr,'->','LineWidth',2,'MarkerFaceColor','auto');
title('P_{avg} vs SNR_{avg} for Soft and Hard Decision Combining');
xlabel('SNR_{avg} in (dB)');
ylabel('Average Probability of Detection');
set(gca,'fontsize',30,'box','on','LineWidth',2,'GridLineStyle','--','GridAlpha'
,0.7);
legend('MNE Combining','EGC Combining','AND Rule','OR Rule','Majority Rule');

```

### A.3 spectrumsenseusrp.py

```

#!/usr/bin/env python
#
# Copyright 2005,2007,2011 Free Software Foundation, Inc.
#
# This file is part of GNU Radio
#
# GNU Radio is free software; you can redistribute it and/or modify
# it under the terms of the GNU General Public License as published by
# the Free Software Foundation; either version 3, or (at your option)
# any later version.
#
# GNU Radio is distributed in the hope that it will be useful,
# but WITHOUT ANY WARRANTY; without even the implied warranty of
# MERCHANTABILITY or FITNESS FOR A PARTICULAR PURPOSE. See the
# GNU General Public License for more details.
#
# You should have received a copy of the GNU General Public License
# along with GNU Radio; see the file COPYING. If not, write to

```

```
# the Free Software Foundation, Inc., 51 Franklin Street,
# Boston, MA 02110-1301, USA.
#
from gnuradio import gr, eng_notation
from gnuradio import blocks
from gnuradio import audio
from gnuradio import filter
from gnuradio import fft
from gnuradio import uhd
from gnuradio.eng_option import eng_option
from optparse import OptionParser
import sys
import math
import struct
import threading
from datetime import datetime
import time
from gnuradio.wxgui import stdgui2, fftsink2, form
import wx

sys.stderr.write("Warning: this may have issues on some machines+Python version
combinations to seg fault due to the callback in bin_statistics.\n\n")

class ThreadClass(threading.Thread):
    def run(self):
        return

class tune(gr.feval_dd):
    """
    This class allows C++ code to callback into python.
    """
    def __init__(self, tb):
        gr.feval_dd.__init__(self)
        self.tb = tb

    def eval(self, ignore):
```

```

"""
This method is called from blocks.bin_statistics_f when it wants
to change the center frequency. This method tunes the front
end to the new center frequency, and returns the new frequency
as its result.

"""

try:
    new_freq = self.tb.set_next_freq()
    while(self.tb.msgq.full_p()):
        time.sleep(0.1)
    return new_freq

except Exception, e:
    print "tune: Exception: ", e


class parse_msg(object):
    def __init__(self, msg):
        self.center_freq = msg.arg1()
        self.vlen = int(msg.arg2())
        assert(msg.length() == self.vlen * gr.sizeof_float)
        t = msg.to_string()
        self.raw_data = t
        self.data = struct.unpack('%df' % (self.vlen,), t)

class my_top_block(gr.top_block):

    def __init__(self):
        gr.top_block.__init__(self)

        usage = "usage: %prog [options] min_freq max_freq"
        parser = OptionParser(option_class=eng_option, usage=usage)
        parser.add_option("-a", "--args", type="string", default="",
                          help="UHD device device address args [default=%default"
                               "]")

```

```

parser.add_option("", "--spec", type="string", default=None,
                  help="Subdevice of UHD device where appropriate")
parser.add_option("-A", "--antenna", type="string", default=None,
                  help="select Rx Antenna where appropriate")
parser.add_option("-s", "--samp-rate", type="eng_float", default=1e6,
                  help="set sample rate [default=%default]")
parser.add_option("-g", "--gain", type="eng_float", default=None,
                  help="set gain in dB (default is midpoint)")
parser.add_option("", "--tune-delay", type="eng_float",
                  default=0.25, metavar="SECS",
                  help="time to delay (in seconds) after changing
                        frequency [default=%default]")
parser.add_option("", "--dwell-delay", type="eng_float",
                  default=0.25, metavar="SECS",
                  help="time to dwell (in seconds) at a given frequency
                        [default=%default]")
parser.add_option("-b", "--channel-bandwidth", type="eng_float",
                  default=6.25e3, metavar="Hz",
                  help="channel bandwidth of fft bins in Hz [default=%
                        default]")
parser.add_option("-l", "--lo-offset", type="eng_float",
                  default=0, metavar="Hz",
                  help="lo_offset in Hz [default=%default]")
parser.add_option("-q", "--squelch-threshold", type="eng_float",
                  default=None, metavar="dB",
                  help="squelch threshold in dB [default=%default]")
parser.add_option("-F", "--fft-size", type="int", default=None,
                  help="specify number of FFT bins [default=samp_rate/
                        channel_bw]")
parser.add_option("", "--real-time", action="store_true", default=False,
                  help="Attempt to enable real-time scheduling")

(options, args) = parser.parse_args()
if len(args) != 2:
    parser.print_help()
    sys.exit(1)

```

```
    self.channel_bandwidth = options.channel_bandwidth

    self.min_freq = eng_notation.str_to_num(args[0])
    self.max_freq = eng_notation.str_to_num(args[1])

    if self.min_freq > self.max_freq:
        # swap them
        self.min_freq, self.max_freq = self.max_freq, self.min_freq

    if not options.real_time:
        realtime = False
    else:
        # Attempt to enable realtime scheduling
        r = gr.enable_realtime_scheduling()
        if r == gr.RT_OK:
            realtime = True
        else:
            realtime = False
            print "Note: failed to enable realtime scheduling"

    # build graph
    self.u = uhd.usrp_source(device_addr=options.args,
                             stream_args=uhd.stream_args('fc32'))

    # Set the subdevice spec
    if(options.spec):
        self.u.set_subdev_spec(options.spec, 0)

    # Set the antenna
    if(options.antenna):
        self.u.set_antenna(options.antenna, 0)

    self.u.set_samp_rate(options.samp_rate)
    self.usrp_rate = usrp_rate = self.u.get_samp_rate()

    self.lo_offset = options.lo_offset
```

```

    if options.fft_size is None:
        self.fft_size = int(self.usrp_rate/self.channel_bandwidth)
    else:
        self.fft_size = options.fft_size

    self.squelch_threshold = options.squelch_threshold

    s2v = blocks.stream_to_vector(gr.sizeof_gr_complex, self.fft_size)

    mywindow = filter.window.blackmanharris(self.fft_size)
    fftr = fft.fft_vcc(self.fft_size, True, mywindow, True)
    power = 0
    for tap in mywindow:
        power += tap*tap
    c2mag = blocks.complex_to_mag_squared(self.fft_size)
    self.freq_step = self.nearest_freq((0.75 * self.usrp_rate),
                                       self.channel_bandwidth)
    self.min_center_freq = self.min_freq + (self.freq_step/2)
    nsteps = math.ceil((self.max_freq - self.min_freq) / self.freq_step)
    self.max_center_freq = self.min_center_freq + (nsteps * self.freq_step)
    self.next_freq = self.min_center_freq
    tune_delay = max(0, int(round(options.tune_delay * usrp_rate /
                                   self.fft_size))) # in fft_frames
    dwell_delay = max(1, int(round(options.dwell_delay * usrp_rate /
                                   self.fft_size))) # in fft_frames
    self.msgq = gr.msg_queue(1)
    self._tune_callback = tune(self) # hang on to this to keep it
                                    from being GC'd
    stats = blocks.bin_statistics_f(self.fft_size, self.msgq,
                                    self._tune_callback, tune_delay,
                                    dwell_delay)
    self.connect(self.u, s2v, fftr, c2mag, stats)

    if options.gain is None:

        g = self.u.get_gain_range()
        options.gain = float(g.start()+g.stop())/2.0

```

```
    self.set_gain(options.gain)
    print "gain =", options.gain

def set_next_freq(self):
    target_freq = self.next_freq
    self.next_freq = self.next_freq + self.freq_step
    if self.next_freq >= self.max_center_freq:
        self.next_freq = self.min_center_freq

    if not self.set_freq(target_freq):
        print "Failed to set frequency to", target_freq
        sys.exit(1)

    return target_freq

def set_freq(self, target_freq):
    """
    Set the center frequency we're interested in.

    Args:
        target_freq: frequency in Hz
        @rypte: bool
    """

    r = self.u.set_center_freq(uhd.tune_request(target_freq, rf_freq=
        target_freq + self.lo_offset), rf_freq_policy=
        uhd.tune_request.POLICY_MANUAL)
    if r:
        return True

    return False

def set_gain(self, gain):
    self.u.set_gain(gain)
```

```

def nearest_freq(self, freq, channel_bandwidth):
    freq = round(freq / channel_bandwidth, 0) * channel_bandwidth
    return freq

def main_loop(tb):

    def bin_freq(i_bin, center_freq):
        freq = center_freq - (tb.usrp_rate / 2) + (tb.channel_bandwidth * i_bin)
        return freq

    bin_start = int(tb.fft_size * ((1 - 0.25) / 2))
    bin_stop = int(tb.fft_size - bin_start)
    fid = open("./usrp.dat", "wb")
    while 1:
        m = parse_msg(tb.msgq.delete_head())
        for i_bin in range(bin_start, bin_stop):
            center_freq = m.center_freq
            freq = bin_freq(i_bin, center_freq)
            power_db = 10 * math.log10(m.data[i_bin] / tb.usrp_rate)
            signal = m.data[i_bin] / (tb.usrp_rate)

            if (power_db > tb.squelch_threshold) and (freq >= tb.min_freq) and (
                freq <= tb.max_freq):
                print freq, signal, power_db
                fid.write(struct.pack('<f', signal))
    fid.close() #closing the file

if __name__ == '__main__':
    t = ThreadClass()
    t.start()

    tb = my_top_block()
    try:
        tb.start()
        main_loop(tb)

    except KeyboardInterrupt:

```

```
    pass
```

## A.4 gnuradiortlsdrsense.py

```
#!/usr/bin/env python2
# -*- coding: utf-8 -*-

#####
# GNU Radio Python Flow Graph
# Title: DTV Spectrum Sensing
# Author: Gill
# Description: Frequency Sweep for UHF White Spaces
# Generated: Fri Mar 10 14:30:20 2017
#####

if __name__ == '__main__':
    import ctypes
    import sys
    if sys.platform.startswith('linux'):
        try:
            x11 = ctypes.cdll.LoadLibrary('libX11.so')
            x11.XInitThreads()
        except:
            print "Warning: failed to XInitThreads()"

from PyQt4 import Qt
from gnuradio import blocks
from gnuradio import eng_notation
from gnuradio import fft
from gnuradio import gr
from gnuradio import qtgui
from gnuradio.eng_option import eng_option
from gnuradio.fft import window
from gnuradio.filter import firdes
from optparse import OptionParser
import numpy as np
```

```
import osmosdr
import sip
import sys
import time

class spectrum_sensing(gr.top_block, Qt.QWidget):

    def __init__(self):
        gr.top_block.__init__(self, "DTv Spectrum Sensing")
        Qt.QWidget.__init__(self)
        self.setWindowTitle("DTv Spectrum Sensing")
        try:
            self.setWindowIcon(Qt.QIcon.fromTheme('gnuradio-grc'))
        except:
            pass
        self.top_scroll_layout = Qt.QVBoxLayout()
        self.setLayout(self.top_scroll_layout)
        self.top_scroll = Qt.QScrollArea()
        self.top_scroll.setFrameStyle(Qt.QFrame.NoFrame)
        self.top_scroll_layout.addWidget(self.top_scroll)
        self.top_scroll.setWidgetResizable(True)
        self.top_widget = Qt.QWidget()
        self.top_scroll.setWidget(self.top_widget)
        self.top_layout = Qt.QVBoxLayout(self.top_widget)
        self.top_grid_layout = Qt.QGridLayout()
        self.top_layout.addLayout(self.top_grid_layout)

        self.settings = Qt.QSettings("GNU Radio", "spectrum_sensing")
        self.restoreGeometry(self.settings.value("geometry").toByteArray())

#####
# Variables
#####
self.samp_rate = samp_rate = int(2e6)
self.freq = freq = 450e6
self.N = N = 1000
```

```

#####
# Blocks
#####
self.rtlsdr_source_0 = osmosdr.source( args="numchan=" + str(1) + " " +
    '' )
self.rtlsdr_source_0.set_time_source('external', 0)
self.rtlsdr_source_0.set_sample_rate(samp_rate)
self.rtlsdr_source_0.set_center_freq(freq, 0)
self.rtlsdr_source_0.set_freq_corr(0, 0)
self.rtlsdr_source_0.set_dc_offset_mode(2, 0)
self.rtlsdr_source_0.set_iq_balance_mode(0, 0)
self.rtlsdr_source_0.set_gain_mode(True, 0)
self.rtlsdr_source_0.set_gain(15, 0)
self.rtlsdr_source_0.set_if_gain(15, 0)
self.rtlsdr_source_0.set_bb_gain(15, 0)
self.rtlsdr_source_0.set_antenna('', 0)
self.rtlsdr_source_0.set_bandwidth(0, 0)

self.qtgui.freq_sink_x_0 = qtgui.freq_sink_c(
    1024, #size
    firdes.WIN_BLACKMAN_hARRIS, #wintype
    0, #fc
    samp_rate, #bw
    "Recieved Signal", #name
    1 #number of inputs
)
self.qtgui.freq_sink_x_0.set_update_time(0.10)
self.qtgui.freq_sink_x_0.set_y_axis(-120, 0)
self.qtgui.freq_sink_x_0.set_y_label('Relative Gain', 'dB')
self.qtgui.freq_sink_x_0.set_trigger_mode(qtgui.TRIG_MODE_FREE, 0.0, 0,
    '')
self.qtgui.freq_sink_x_0.enable_autoscale(True)
self.qtgui.freq_sink_x_0.enable_grid(True)
self.qtgui.freq_sink_x_0.set_fft_average(1.0)
self.qtgui.freq_sink_x_0.enable_axis_labels(True)
self.qtgui.freq_sink_x_0.enable_control_panel(False)

```

```

if not True:
    self.qtgui.freq_sink_x_0.disable_legend()

if "complex" == "float" or "complex" == "msg_float":
    self.qtgui.freq_sink_x_0.set_plot_pos_half(not True)

labels = ['', '', '', '', '',
          '', '', '', '', '']
widths = [2, 1, 1, 1, 1,
          1, 1, 1, 1, 1]
colors = ["blue", "red", "green", "black", "cyan",
          "magenta", "yellow", "dark red", "dark green", "dark blue"]
alphas = [1.0, 1.0, 1.0, 1.0, 1.0,
          1.0, 1.0, 1.0, 1.0, 1.0]

for i in xrange(1):
    if len(labels[i]) == 0:
        self.qtgui.freq_sink_x_0.set_line_label(i, "Data {0}".format(i))
    else:
        self.qtgui.freq_sink_x_0.set_line_label(i, labels[i])
        self.qtgui.freq_sink_x_0.set_line_width(i, widths[i])
        self.qtgui.freq_sink_x_0.set_line_color(i, colors[i])
        self.qtgui.freq_sink_x_0.set_line_alpha(i, alphas[i])

self._qtgui.freq_sink_x_0.win = sip.wrapinstance(
    self.qtgui.freq_sink_x_0.pyqwidget(), Qt.QWidget)
self.top_layout.addWidget(self._qtgui.freq_sink_x_0.win)
self.fft_vxx_0 = fft.fft_vcc(1024, True, (window.blackmanharris(1024)),
                            True, 1)
self.blocks_vector_to_stream_0 = blocks.vector_to_stream(gr.sizeof_float
                                                       *1, 1024)
self.blocks_stream_to_vector_0 = blocks.stream_to_vector(
    gr.sizeof_gr_complex*1, 1024)
self.blocks_moving_average_xx_0 = blocks.moving_average_ff(N, 1, 4000)
self.blocks_file_sink_2_0 = blocks.file_sink(gr.sizeof_float*1, '/home/
gill/Desktop/ms-thesis/gr-spectrumsensing/grc rtl-sdr_sensing/
Results/snr_check.dat', False)

```

```

        self.blocks_file_sink_2_0.set_unbuffered(False)
        self.blocks_complex_to_mag_squared_0 = blocks.complex_to_mag_squared
        (1024)

#####
# Connections
#####
self.connect((self.blocks_complex_to_mag_squared_0, 0), (
    self.blocks_vector_to_stream_0, 0))
self.connect((self.blocks_moving_average_xx_0, 0), (
    self.blocks_file_sink_2_0, 0))
self.connect((self.blocks_stream_to_vector_0, 0), (self.fft_vxx_0, 0))
self.connect((self.blocks_vector_to_stream_0, 0), (
    self.blocks_moving_average_xx_0, 0))
self.connect((self.fft_vxx_0, 0), (self.blocks_complex_to_mag_squared_0,
    0))
self.connect((self.rtlsdr_source_0, 0), (self.blocks_stream_to_vector_0,
    0))
self.connect((self.rtlsdr_source_0, 0), (self.qtgui_freq_sink_x_0, 0))

def closeEvent(self, event):
    self.settings = Qt.QSettings("GNU Radio", "spectrum_sensing")
    self.settings.setValue("geometry", self.saveGeometry())
    event.accept()

def get_samp_rate(self):
    return self.samp_rate

def set_samp_rate(self, samp_rate):
    self.samp_rate = samp_rate
    self.rtlsdr_source_0.set_sample_rate(self.samp_rate)
    self.qtgui_freq_sink_x_0.set_frequency_range(0, self.samp_rate)

def get_freq(self):
    return self.freq

def set_freq(self, freq):

```

```
    self.freq = freq
    self.rtlsdr_source_0.set_center_freq(self.freq, 0)

def get_N(self):
    return self.N

def set_N(self, N):
    self.N = N
    self.blocks_moving_average_xx_0.set_length_and_scale(self.N, 1)

def main(top_block_cls=spectrum_sensing, options=None):

    from distutils.version import StrictVersion
    if StrictVersion(Qt.qVersion()) >= StrictVersion("4.5.0"):
        style = gr.prefs().get_string('qtdesigner', 'style', 'raster')
        Qt.QApplication.setGraphicsSystem(style)
    qapp = Qt.QApplication(sys.argv)

    tb = top_block_cls()
    tb.start()
    tb.show()

    def quitting():
        tb.stop()
        tb.wait()
    qapp.connect(qapp, Qt.SIGNAL("aboutToQuit()"), quitting)
    qapp.exec_()

if __name__ == '__main__':
    main()
```

## Appendix B

# LTE-R Analysis Code

### B.1 kfactordist.m

```
% Calculating K-factor for the tunnel environment for HST
clear all;
close all;
clc;

%% Creating a doppler profile for high speed raiway scenario..
Ds = 30;%Initial Distance between tx and rx times 2..
Dmin = 2;% Distance between raiway tracks and leaky feeder cables...
Kf = [];
fc = 3e9;%center frequency..
c = 3e8;
v = 138.9;%300;
t = linspace(0, (2*Ds)/v(1),100);
fd = (v*fc)/3e8;%maximum doppler frequency...
costheta = zeros(size(t));%angle between BS and MS
d1 = [];
for i=1:length(t)
    d1(i) = sqrt(2^2+(Ds/2-v(1)*t(i))^2);%distance between tx and rx..
    if t(i) >=0 && t(i)<= (Ds/v)
        costheta(i) = ((Ds/2)-v*t(i))./sqrt(Dmin^2+(Ds/2-v*t(i))^2);
    end
end
```

```

elseif t(i) > (Ds/v) && t(i)<=(2*Ds)/v
costheta(i) = (-1.5*Ds+v*t(i))./sqrt(Dmin^2+(-1.5*Ds+v*t(i))^2);
end
end
fs = fd*costheta;
thetadeg = acosd(costheta);
fc_wds = fc-fs;
lambda = c./fc_wds;
Cin = 5-((0.1*1.8e10)./fc_wds)*1j;
C = Cin;
gammanum = C.*sind(thetadeg)-sqrt(Cin-(cosd(thetadeg)).^2);
gammaden = C.*sind(thetadeg)+sqrt(Cin-(cosd(thetadeg)).^2);
gamma = gammanum./gammaden;
ht = 6.1;%height of feeder cable
hr = 4.2;%height of the train
var1 = sqrt(d1.^2+(ht+hr)^2);
var2 = sqrt(d1.^2+(ht-hr)^2);
phase = (((2*pi)./lambda).* (var1-var2))*180/pi;
gammad = atan2d(imag(gamma),real(gamma));
phasegamma = abs(cosd(gammad-phase));
K = abs(gamma).^2+2*abs(gamma).*phasegamma;
Kf = 10*log10(1./K);

```

## B.2 bercalculation.m

```

% Demonstration of Eb/N0 Vs SER for M-QAM modulation scheme
clc;
load Kf;
load t;
%-----Input Fields-----
%% QPSk
bitsperframe=1e3; %Number of input symbols
EbN0dB = [linspace(0,20,50) fliplr(linspace(0,20,50))]; %Define EbN0dB range for
simulation
M=4; %for QPSk modulation.

```

```

hMod = comm.RectangularQAMModulator('ModulationOrder',M);
const = step(hMod,(0:3)');
%-----
refArray = 1/sqrt(2)*const';
k=log2(M);
totPower=15; %Total power of LOS path & scattered paths

EsN0dB = EbN0dB + 10*log10(k);
biterrsim = zeros(size(EsN0dB));
%---Generating a uniformly distributed random numbers in the set [0,1,2,...,M-1]
data=ceil(M.*rand(bitsperframe,1))-1;
s=refArray(data+1); %QPSK Constellation mapping with Gray coding
%--- Reference Constellation for demodulation and Error rate computation--
refI = real(refArray);
refQ = imag(refArray);
%---Place holder for Symbol Error values for each Es/N0 for particular M value--
index=1;
u=1;
% Kf = 4.9;
K = 10.^ (Kf/10);
for x=EsN0dB
    sn=sqrt(K(u)/(K(u)+1)*totPower); %Non-Centrality Parameter
    sigma=totPower/sqrt(2*(K(u)+1));
    h=((sigma*randn(1,bitsperframe)+sn)+li*(randn(1,bitsperframe)*sigma+0));
    numerr = 0;
    numBits = 0;
    while numerr < 100 && numBits < 1e7
        %-----
        %Channel Noise for various Es/N0
        %-----
        %Adding noise with variance according to the required Es/N0
        noiseVariance = 1/(10.^ (x/10));%Standard deviation for AWGN Noise
        noiseSigma = sqrt(noiseVariance/2);
        %Creating a complex noise for adding with M-QAM modulated signal
        %Noise is complex since M-QAM is in complex representation
        noise = noiseSigma*(randn(size(s))+li*randn(size(s)));
        received = s.*h + noise;

```

```

%-----I-Q Branching-----
received = received./h;
r_i = real(received);
r_q = imag(received);
%---Decision Maker-Compute  $(r_i - s_i)^2 + (r_q - s_q)^2$  and choose the
smallest
r_i_repmat = repmat(r_i,M,1);
r_q_repmat = repmat(r_q,M,1);
distance = zeros(M,bitsperframe); %place holder for distance metric
minDistIndex=zeros(bitsperframe,1);
for j=1:bitsperframe
    %---Distance computation -  $(r_i - s_i)^2 + (r_q - s_q)^2$  -----
    distance(:,j) = (r_i_repmat(:,j)-refI').^2+(r_q_repmat(:,j)-refQ').
        ^2;
    %---capture the index in the array where the minimum distance occurs
    [dummy,minDistIndex(j)]=min(distance(:,j));
end
y = minDistIndex - 1;
%-----Symbol Error Rate Calculation
-----
dataCap = y;
numerr = sum(dataCap~=data)+numerr;
numBits = numBits+bitsperframe;
disp(numerr);
end
symErrSimulatedqpsk(1,index) = numerr/numBits;
biterrsime(1,index) = symErrSimulatedqpsk(1,index)/k;
index=index+1;
% u=u+1;
end

%% 16 QAM
bitsperframe=1e3; %Number of input symbols
EbN0dB = [linspace(0,10,50) fliplr(linspace(0,10,50))]; %Define EbN0dB range for
simulation
M=16; %for QPSK modulation.
hMod = comm.RectangularQAMModulator('ModulationOrder',M);

```

```

const = step(hMod, (0:M-1)');
%-----
refArray =1/sqrt(10)*const';
k=log2(M);
totPower=10; %Total power of LOS path & scattered paths

EsN0dB = EBN0dB + 10*log10(k);
biterrsim = zeros(size(EsN0dB));
%---Generating a uniformly distributed random numbers in the set [0,1,2,...,M-1]
data=ceil(M.*rand(bitsperframe,1))-1;
s=refArray(data+1); %QPSK Constellation mapping with Gray coding
%--- Reference Constellation for demodulation and Error rate computation--
refI = real(refArray);
refQ = imag(refArray);
%---Place holder for Symbol Error values for each Es/N0 for particular M value--
index=1;
u=1;
K = 10.^ (Kf/10);
for x=EsN0dB
    numerr = 0;
    numBits = 0;
    while numerr < 100 && numBits < 1e7
        sn=sqrt(K(u) / (K(u)+1)*totPower); %Non-Centrality Parameter
        sigma=totPower/sqrt(2*(K(u)+1));
        h=((sigma*randn(1,bitsperframe)+sn)+li*(randn(1,bitsperframe)*sigma+0));
        %-----
        %Channel Noise for various Es/N0
        %-----
        %Adding noise with variance according to the required Es/N0
        noiseVariance = 1/(10.^ (x/10));%Standard deviation for AWGN Noise
        noiseSigma = sqrt(noiseVariance/2);
        %Creating a complex noise for adding with M-QAM modulated signal
        %Noise is complex since M-QAM is in complex representation
        noise = noiseSigma*(randn(size(s))+li*randn(size(s)));
        received = s.*h + noise;
        %-----I-Q Branching-----

```

```

received = received./h;
r_i = real(received);
r_q = imag(received);
%---Decision Maker-Compute  $(r_i - s_i)^2 + (r_q - s_q)^2$  and choose the
smallest
r_i_repmat = repmat(r_i,M,1);
r_q_repmat = repmat(r_q,M,1);
distance = zeros(M,bitsperframe); %place holder for distance metric
minDistIndex=zeros(bitsperframe,1);
for j=1:bitsperframe
%---Distance computation -  $(r_i - s_i)^2 + (r_q - s_q)^2$  -----
distance(:,j) = (r_i_repmat(:,j)-refI').^2+(r_q_repmat(:,j)-refQ').
^2;
%---capture the index in the array where the minimum distance occurs
[dummy,minDistIndex(j)]=min(distance(:,j));
end
y = minDistIndex - 1;
%-----Symbol Error Rate Calculation
-----
dataCap = y;
numerr = sum(dataCap~=data)+numerr;
numBits = numBits+bitsperframe;
disp(numerr);
end
symErrSimulatedqam(l,index) = numerr/numBits;
biterrsime(l,index) = symErrSimulatedqam(l,index)/k;
index=index+1;
u=u+1;
end

%% 64 QAM Modulation...
bitsperframe=1e3; %Number of input symbols
EbN0dB = [linspace(0,10,50) fliplr(linspace(0,10,50))]; %Define EbN0dB range for
simulation
M=64; %for QPSK modulation.
hMod = comm.RectangularQAMModulator('ModulationOrder',M);
const = step(hMod, (0:M-1)');

```

```

%-----
refArray =1/sqrt(42)*const';
k=log2(M);
totPower=10; %Total power of LOS path & scattered paths
EsN0dB = EbN0dB + 10*log10(k);
biterrsime = zeros(size(EsN0dB));
%---Generating a uniformly distributed random numbers in the set [0,1,2,...,M-1]
data=ceil(M.*rand(bitsperframe,1))-1;
s=refArray(data+1); %QPSK Constellation mapping with Gray coding
%--- Reference Constellation for demodulation and Error rate computation--
refI = real(refArray);
refQ = imag(refArray);
%---Place holder for Symbol Error values for each Es/N0 for particular M value--
index=1;
u=1;
K = 10.^ (Kf/10);
for x=EsN0dB
    sn=sqrt(K(u) / (K(u)+1)*totPower); %Non-Centrality Parameter
    sigma=totPower/sqrt(2*(K(u)+1));
    h=((sigma*randn(1,bitsperframe)+sn)+1i*(randn(1,bitsperframe)*sigma+0));
    numerr = 0;
    numBits = 0;
    while numerr < 100 && numBits < 1e7
        %-----
        %Channel Noise for various Es/N0
        %-----
        %Adding noise with variance according to the required Es/N0
        noiseVariance = 1/(10.^(x/10));%Standard deviation for AWGN Noise
        noiseSigma = sqrt(noiseVariance/2);
        %Creating a complex noise for adding with M-QAM modulated signal
        %Noise is complex since M-QAM is in complex representation
        noise = noiseSigma*(randn(size(s))+1i*randn(size(s)));
        received = s.*h + noise;
        %-----I-Q Branching-----
        received = received./h;
        r_i = real(received);
        r_q = imag(received);

```

```

%---Decision Maker-Compute  $(r_i-s_i)^2+(r_q-s_q)^2$  and choose the
smallest
r_i_repmat = repmat(r_i,M,1);
r_q_repmat = repmat(r_q,M,1);
distance = zeros(M,bitsperframe); %place holder for distance metric
minDistIndex=zeros(bitsperframe,1);
for j=1:bitsperframe
    %---Distance computation -  $(r_i-s_i)^2+(r_q-s_q)^2$  -----
    distance(:,j) = (r_i_repmat(:,j)-refI').^2+(r_q_repmat(:,j)-refQ').
        ^2;
    %---capture the index in the array where the minimum distance occurs
    [dummy,minDistIndex(j)]=min(distance(:,j));
end
y = minDistIndex - 1;
%-----Symbol Error Rate Calculation
-----
dataCap = y;
numerr = sum(dataCap~=data)+numerr;
numBits = numBits+bitsperframe;
disp(numerr);
end
symErrSimulatedqam64(1,index) = numerr/numBits;
biterrsime(1,index) = symErrSimulatedqam64(1,index)/k;
index=index+1;
u=u+1;
end

%%
fig = figure;
semilogy(t*1e3,symErrSimulatedqpsk(1,:),'-d','LineWidth',2);
hold on;
grid on;
semilogy(t*1e3,symErrSimulatedqam(1,:),'-d','LineWidth',2);
semilogy(t*1e3,symErrSimulatedqam64(1,:),'-d','LineWidth',2);
xlabel('Time (ms)');
ylabel('Bit Error Rate (Pb)');
title(['BER For OFDM Under Rician Fading Environment Inside Tunnel']);

```

```
set(gca,'fontsize',30,'box','on','LineWidth',2,'GridLineStyle','--','GridAlpha'  
,0.7);  
axis([0 max(t)*1e3 10e-7 0])  
lgd = legend('QPSK','16QAM','64QAM');  
lgd.FontSize=20;
```