



Contents lists available at ScienceDirect

Ad Hoc Networks

journal homepage: www.elsevier.com/locate/adhoc

Flexible channel selection mechanism for cognitive radio based last mile smart grid communications

Saud Althunibat^{a,*}, Qi Wang^b, Fabrizio Granelli^b

^a Al-Hussein Bin Talal University, Ma'an, Jordan

^b University of Trento, Trento, Italy

ARTICLE INFO

Article history:

Received 22 July 2015

Revised 23 October 2015

Accepted 25 October 2015

Available online xxx

Keywords:

Smart grid

Smart grid communications

Cognitive radio

Channel selection

Last mile smart grid communications

Spectrum sensing

ABSTRACT

Smart grid (SG) operation requires a reliable, accurate and effective communication link between the distributed meters and the control center. However, dedicating a portion of the spectrum is difficult due to the spectrum scarcity problem. Cognitive radio (CR) technology has been nominated as a good candidate for SG communications due to its efficiency and flexibility. Indeed, channel selection in CR-based SG systems is still an open issue, and it is investigated in this paper. The paper proposes a novel channel selection mechanism that is able to adapt the selection criteria based on the type of transmitted data. The proposed mechanism is proven to provide high performance compared to the non-adaptable mechanisms.

© 2015 Published by Elsevier B.V.

1. Cognitive radios for smart grid communication

The communications infrastructure for smart grid (SG) has recently received an increasing attention. Given the heterogeneity of the SG, it is necessary to clarify that the considered scenario in the paper consists of the communication links used to interconnect the distributed meters at the customers' side and the local control center. To some extent, from a communications perspective, this could be considered the last mile of the SG system. Many wireless technologies have been nominated in the literature as candidate infrastructures for SG communications in such scenario [1,2]. However, two main properties that may distinguish SG nature from other wireless systems: first, both the transmitter and receiver are always fixed, which alleviate many problems generated by mobile transmitters, such as fast channel variation, frequent handovers, etc.; second, a considerable part of the transmitted data has a regular (fixed) generation rate, i.e.,

both transmitter and receiver know who will send these data and when. Moreover, such types of data usually tolerate a relatively long delay (in communication terms). These properties should be taken into account in order to identify suitable, efficient and reliable communication infrastructures for the SG.

Cognitive radio (CR) technology is one of the candidate technologies to serve for last mile SG communications [3–5]. CR is an intelligent software-defined radio technology that facilitates efficient, reliable and dynamic use of the unused portions of the radio spectrum by adapting its configuration according to the environment radio conditions [6]. The IEEE 802.22 is the first standardization activity on CR networks based on opportunistic utilization of TV spectrum bands [7].

CR based on IEEE 802.22 represents a good candidate for last mile SG communications for several reasons, (i) Unlike other wireless technologies, CRs operate in unused TV frequencies, which, in view of spectrum scarcity, represents an extremely useful feature. (ii) high data rates up to tens of Mbps can be achieved by CR. (iii) Due to the long-range propagation characteristics of the TV bands, the coverage area can reach up to 100 km. (iv) CR is adaptive, programmable

* Corresponding author. Tel.: +962799543913.

E-mail addresses: saud.althunibat@ahu.edu.jo (S. Althunibat), wang@disi.unitn.it (Q. Wang), granelli@disi.unitn.it (F. Granelli).

<http://dx.doi.org/10.1016/j.adhoc.2015.10.008>

1570-8705/© 2015 Published by Elsevier B.V.

40 and flexible, since it is built on a software-defined radio
41 platform.

42 Although the above mentioned reasons can make CR a
43 preferred choice for SG communication, three main issues
44 should be still addressed. First, identifying the unused TV
45 frequencies represents a key issue in CR implementation.
46 Usually, information about band occupancy is obtained by a
47 pre-process called spectrum sensing, which consumes time
48 and energy [8]. Second, errors can occur in identifying the
49 availability of the TV bands due to probable imperfect spec-
50 trum sensing, which is reflected on the reliability of cognitive
51 transmission [9]. Third, cognitive transmission is a threat-
52 ened transmission due to the possible and unpredictable ap-
53 pearance of the original licensed user during transmission.
54 However, the last issue becomes less significant in case of
55 CR based on IEEE 802.22 since a TV frequency that has been
56 identified as vacant cannot be reused during CR's usage in the
57 same area. As a consequence, the former two issues (the first
58 and the second) are taken into consideration throughout this
59 work.

60 Based on the dynamic smart grid communication envi-
61 ronment and comprehensive communication applications in
62 the smart grid architecture, researchers have explored dif-
63 ferent challenges of the CR-based SG communications in
64 recent years [10,11]. Ghassemi et al. [12] studied current
65 energy management requirements in the smart grid and em-
66 ployed CR as the solution for the communication require-
67 ments in the wide area networks (WANs) of SG, based on IEEE
68 802.22 standard. In particular, the additional bandwidth re-
69 quired for non-critical data is addressed by Ghassemi et al.
70 [12]. Stand-alone radio and secondary radio were proposed
71 for CR communication systems based on IEEE 802.22 stan-
72 dard. According to such two architectures, Ghassemi et al.
73 [12] employed dual-radio scheme for CR-based transmission,
74 which can offer an effective spectrum sensing process. Yu
75 et al. [13] proposed a new spectrum access scheme named
76 hybrid spectrum access (HSA) in which both licensed and un-
77 licensed spectrum bands can be scheduled to guarantee the
78 transmission services in SG. The HSA scheme is employed
79 to balance the spectrum utilization and the QoS provision-
80 ing in different environments. Based on the HSA strategy, Yu
81 et al. considered two types of spectrum sensing schemes,
82 periodic and on-demand sensing. The transmission QoS in
83 SG is improved via employing the various applications of
84 the spectrum sensing schemes. Deng et al. in [14] deliv-
85 ered cognitive radio into SG to improve the communication
86 quality in distribution network for balancing the real-time
87 power demand and to schedule the peak load. Especially,
88 in the data transmission process of smart meters, the deci-
89 sion of channel selection and the reduction the communi-
90 cation outage can be guaranteed by the spectrum sensing
91 and channel switching, in which the licensed and unlicensed
92 channels have equal opportunities to be selected. During the
93 best sensing-performance tradeoff, the primary user using
94 the licensed channel is protected. The sensing-performance
95 tradeoff problem is formulated to explore the probability of
96 lower communication cost. The collision of the communi-
97 cation outage is also analyzed for the power consumers.

98 The errors can occur during the process of spectrum
99 sensing, maybe caused by dynamic parameters in the harsh
100 SG environments, just as equipment noise, cross-tier inter-

ference and etc. According to this challenge, Shah et al. [15] 101
presented a CR-based cross-layer framework to meet the 102
QoS requirements of diverse SG applications. The proposed 103
mechanism employed the emerging cognitive radio tech- 104
nology to reduce the noisy and congested spectrum bands, 105
yielding reliable and high capacity wireless links for SG com- 106
munications. Meanwhile, a distributed control algorithm 107
(DCA) was also designed to support QoS via channel and 108
control, scheduling and routing decisions, which maximizes 109
the network utilization under the constraints of QoS. 110

111 Due to the features of IEEE 802.22 standard, CR based 112
communication technologies can serve the SG applications 113
with high reliability even in harsh environmental condi- 114
tions. The growing demands of multimedia applications 115
in SG communications require large bandwidth and net- 116
work resources. In particular, large-size and time-sensitive 117
multimedia delivery requires more reliability and stability 118
in CR based SG communications. In this case, Wang et al. 119
[16] and Huang et al. [17] explored related issues of CR 120
based multimedia communications and delivered their own 121
mechanisms. Considering the mutual interactions with the 122
enhanced reliability and efficiency of whole SG system, 123
Wang et al. [16] proposed a combined scheme of electrical 124
and CR based networks to effectively support the wireless 125
transmissions of large-sized multimedia data created by the 126
dynamic SG applications. In order to achieve the required 127
quality of experience (QoE) performance in the SG system, 128
a CR networking paradigm is also proposed in [16] to effi- 129
ciently manage the channel allocation for both primary users 130
(PUs) and secondary users (SUs).

131 As we discussed previously, the CR based communication 132
platform is essentially needed to support large-size multi- 133
media data delivery in the SG environment. According to the 134
various traffic types of SG communications, Huang et al. [17] 135
developed CR-based channel allocation and traffic schedul- 136
ing schemes to mitigate the risk in multimedia data trans- 137
mission via employing CR based communication networks. 138
In these schemes, two types of channel switch, including 139
periodic switching and triggered switching are introduced. 140
Meanwhile, the spectrum sensing errors are also considered 141
for supporting such mechanisms. Based on the proposed 142
schemes, the different traffic types in SG are prioritized for 143
traffic scheduling of the SUs and the system utility optimiza- 144
tion problem for SG communications system is solved as 145
well.

146 In this paper, we focus on the channel selection in CR- 147
based SG communications. Generally, several channel selec- 148
tion mechanisms in CR networks have been proposed. In [18], 149
Jiang et al. created a dynamic programming approach to ex- 150
plore the channel selection scheme for an optimal sensing 151
order with adaptive modulation. The optimal sensing order 152
problem is settled in the scenario of multi-channel cogni- 153
tive medium access control involved opportunistic transmis- 154
sions. In the SNR-based channel selection scheme, the SNR 155
is regulated in the time slot and changes stochastically in 156
the previous and next time interval. The SNR is also denoted 157
as the core parameter to create a common probability den- 158
sity function and the achievable transmission rate, in order 159
to deliver the optimal selection mechanism of sensing order 160
in even more complex scenarios. The dynamic transmis- 161
sion rate is settled to compare with expected rate on each

channel to decide which channel should be selected as the current transmission channel for secondary user. In [19], Zou et al. explored the adaptive cooperation diversity mechanism with best-relay selection in cognitive radio networks. In the proposed adaptive cooperation scheme, the SNR is introduced to establish the equations of the primary outage probability and secondary outage probability. Specifically, the performance of secondary outage probability has direct linear relationship with transmit SNR, when the primary outage probability is guaranteed. In the best-relay selection mechanism, the mathematical relationship of these two probabilities decides the result of adaptive cooperation. As for the reliability-based mechanisms, a reliability-based clustering cooperative sensing mechanism is addressed by Liu et al. in [20], considering the accuracy and throughput for the spectrum sensing between randomly allocated primary user and secondary user. The secondary users' reliability is one important step in the proposed reliability-based multiple sensors clustering cooperative scheme. The parameter represents the reliability of secondary user in settled time interval, can be amended to fit dynamic complicated environment. The system throughput has been enhanced in view of the situation of primary user. In order to reduce the communication outage in the data transmission, the spectrum sensing and channel switching are introduced in the proposed mechanism with the energy detector. Also in [21], in order to explore the channel selection problem with reliability guarantee in the cognitive radio networks, Zhang et al. addressed a novel scheme formulated the route reliability based on the link valid probability includes the channel usable probability and channel numbers. A link valid probability based reliability modeling (LVPR) is proposed to value the route reliability based on link valid probability. The use of link is valid as long as at least one channel can be employed. Then the channel selection is preferable on the stability consideration. Other proposed channel selection mechanisms aim at improving energy efficiency. For example, in [22], Abolarinwa et al. presented energy-efficient channel selection using a learning algorithm where a transmitter selects the most energy efficient channel based on its learned experience in the past. In [23], a channel selection mechanism for control channels in cognitive mesh networks has been proposed. In detail, a cooperative swarm intelligence-based algorithm is used to select the control channel based on its quality and the decisions of neighbors. Channel contention is used as metric for channel selection in [24]. Specifically, the sensing process should be able to provide information about the contention of all channels, and consequently, the less-contention channel will be used in order to improve throughput. Similarly, the energy efficiency has been considered together with the channel contention as a combined metric for channel selection in [25].

The above discussed works have not considered the type of the transmitted data. Such an issue should affect the overall performance of the system. In SG systems, transmitted data belong to different categories, namely, delay-sensitive and nondelay-sensitive data. Ignoring the data type in selecting the transmitting channel may lead to degraded performance. For example, selecting the channel based on their reliability will negatively influence the throughput. On the other hand, adopting the expected data rate as a selection metric will increase the re-transmission probability, which

delays the delivery of delay-sensitive data. In this work, we propose a mechanism for selecting the best channel to use for SG communications. In the proposed mechanism, the type of transmitted data plays a significant role in choosing the best channel, as the criteria for selecting the transmit channel change according to the transmitted data. In detail, data that are delay sensitive should be transmitted through a channel with high reliability, while other types of data that are not delay sensitive can be transmitted over channels with high data rates. Based on such principle, the transmit node will have more freedom to select the transmit channel based on the special requirement for their instantaneous transmitted data. Although our proposed channel selection mechanism might be applied to different communication systems, its high performance in CR-based SG systems is evident. This is due to that the proposed mechanism takes into consideration a special property of CR-based SG systems which is the delay sensitivity of the transmitted data. Nevertheless, the proposed mechanism might be able to achieve high performance in communication systems that share this special property with the SG systems.

The rest of the paper is organized as follows. Section 2 presents the system model considered in this paper. The proposed mechanism is discussed in detail in Section 3. The performance of the proposed mechanism is explored through simulation results in Section 4, and finally conclusions are drawn in Section 5.

2. System model

A typical SG system is considered, as shown in Fig. 1. In general, a SG consists of power plant, power utilities, transmission grid, distribution grid and terminal power consumers. There are two types of flows in the SG system, the power flow and the information flow. One of the important differences between the traditional power grid and the smart grid is that terminal power consumers do not only unilaterally receive the orders from utilities, but they can also report and send their demand information back to utilities via communication links. Considering the SG communications environmental conditions, safe, stable, reliable and efficient communications between terminal power consumers and utilities is recognized as a huge challenge.

Recently, the major changes are being promoted in the distribution networks of SG system. A simple architecture of the distribution grid is shown in Fig. 2. From Fig. 2, the distribution grid represents the last mile of the SG and it includes, primary substation, secondary substation, feed lines and terminal power consumers with residential renewable energy generation system. Huge amount of terminal power consumers would have dynamic and random power consumption behaviors that would generate an aggregate power demand and other potential requests to the utilities. The paper discusses the usage of the TV spectrum based CR communication technologies as a potential candidate to establish the wireless communication links in the distribution section of the SG system for the multimedia information delivery.

The considered TV spectrum is divided into L channels : $\{C_1, C_2, \dots, C_L\}$. The process of SG transmission based on CR technology is performed in three phases: spectrum sensing, channel selection, and data transmission. In the spectrum

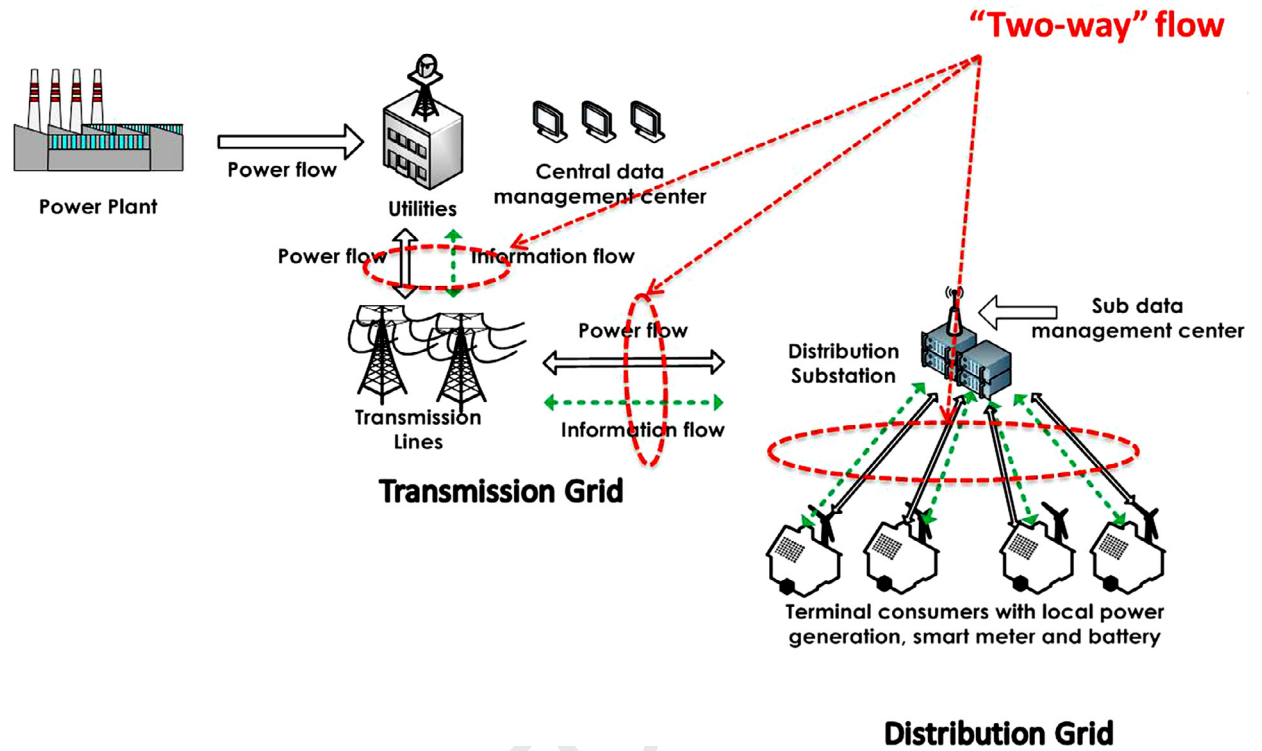


Fig. 1. A simplified view of a typical smart grid system.

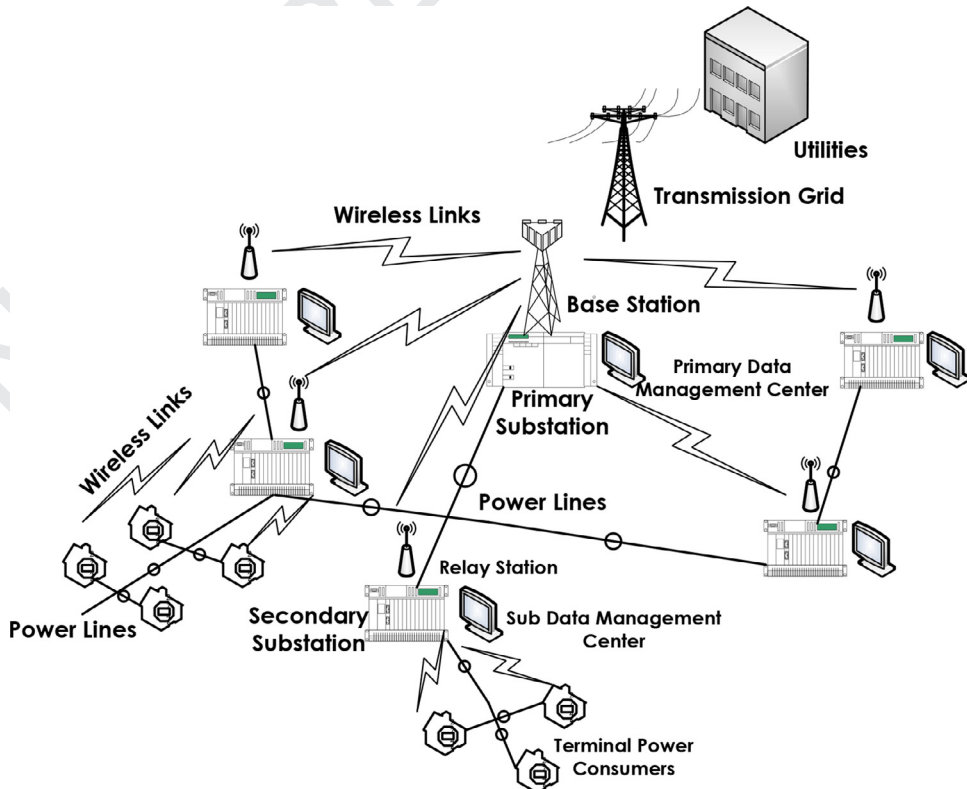


Fig. 2. The conceptual architecture of the distribution grid.

282 sensing phase, the idle channels are identified, and as a se-
 283 quence, a channel (or more) is selected in the channel selec-
 284 tion phase in order to transmit the data in the last phase. The
 285 following subsections provide additional details on each of
 286 the three phases.

287 2.1. Phase I: spectrum sensing

288 As mentioned earlier, the first phase is to identify the idle
 289 TV channel through the spectrum sensing process. The con-
 290 sidered spectrum sensing technique in this work is energy
 291 detection, due to its simplicity and low cost [26]. Energy de-
 292 tection is performed by collecting a set of energy samples
 293 from each channel, and then comparing the accumulated av-
 294 erage of the collected samples with a predefined threshold,
 295 denoted by λ . If the average is larger than λ , the correspond-
 296 ing channel will be declared as used, while otherwise it will
 297 be identified as idle. The following formula describes the
 298 function of the detection process [27]:

$$d_i = \begin{cases} \text{used,} & \text{if } Y_i \geq \lambda, \\ \text{idle,} & \text{if } Y_i < \lambda, \end{cases} \quad (1)$$

299 where d_i the detection decision of the i th channel, Y_i is the
 300 average of the collected samples from the i th channel.

301 The performance of spectrum sensing is usually measured
 302 by two probabilities, namely, the detection probability (P_d)
 303 and the false-alarm probability (P_f). The detection probability
 304 is defined as the probability that a channel is identified as
 305 used given that it is used, and is expressed as follows [27]:

$$P_{di} = Q\left(\frac{Y_i - \sigma_n^2 - \sigma_x^2}{\frac{\sigma_n^2 - \sigma_x^2}{\sqrt{S}}}\right), \quad (2)$$

306 where σ_n^2 is the noise variance, σ_x^2 is the variance of the
 307 transmitted signal, S is the number of the collected samples
 308 during the sensing, and $Q(x) = \frac{1}{\sqrt{2\pi}} \int_x^\infty \exp\left(-\frac{t^2}{2}\right) \cdot dt$.

309 The false-alarm probability is the probability that a chan-
 310 nel is identified as used given that it is idle, and is expressed
 311 as follows [27]:

$$P_{fi} = Q\left(\frac{Y_i - \sigma_n^2}{\frac{\sigma_n^2}{\sqrt{S}}}\right). \quad (3)$$

312 The reader should note that high performance can be
 313 achieved by high detection probability and low false-alarm
 314 probability [28]. Moreover, it should be underlined that the
 315 spectrum sensing is imperfect process, i.e. the availability de-
 316 cision can be a false decision [29]. The false decision occurs
 317 in two cases, the false-alarm case and the missed-detection
 318 case. While the former is described earlier, the latter occurs
 319 when a channel is identified as idle given that is used. The
 320 missed detection is the complementary probability of the de-
 321 tection probability, and can be expressed as follows:

$$P_{mdi} = 1 - Q\left(\frac{Y_i - \sigma_n^2 - \sigma_x^2}{\frac{\sigma_n^2 - \sigma_x^2}{\sqrt{S}}}\right). \quad (4)$$

322 2.2. Phase II: channel selection

323 As the idle channels have been identified according to the
 324 spectrum sensing process, the transmitter node will be able

to choose a channel for data transmission from the set of
 the available channels. The mechanism for selecting the used
 channel will be discussed later.

Generally, the achievable data rate of the i th channel (in
 case it is actually idle) can be expressed as follows (Shannon's
 law):

$$R_i = B \log_2 (1 + \gamma_i), \quad (5)$$

where B is the channel bandwidth and γ_i is the signal to
 noise ratio of the i th channel.

2.3. Phase III: data transmission

After identifying the transmit channel based on the
 adopted channel selection mechanism, the data transmission
 will start. The transmitted data can be classified into two cat-
 egories as follows [30–32]:

- Delay sensitive (DS): this type of data is very important
 and should be delivered as fast as possible to the control
 center. An example for this type is fault alarms.
- Non-delay sensitive (NDS): this type of data can tolerate a
 longer delay. For example, statistical data about the con-
 sumption, load variation, etc., as well as the regular meter
 readings.

3. The proposed channel selection mechanisms

In this section, we propose a channel selection mecha-
 nism for CR-base SG communications. The proposed mech-
 anism is designed to adapt the selection criteria according to
 the type of data to transmit. Thus, it should (as will be proved
 later) provide high performance compared to other channel
 selection mechanisms.

The proposed channel selection mechanism is described
 in Fig. 3. The figure describes the whole process of SG com-
 munication based on CR technology, starting from spec-
 trum sensing, channel estimation, channel selection, and
 data transmission.

As indicated in Fig. 3, the first step is to sense the channel
 set, which will result in identifying a subset of the sensed
 channels as idle. The idle subset can be different among
 nodes, since each node can experience different sensing per-
 formance and channel conditions. Therefore, the two com-
 municating nodes should exchange the sensing results which
 include: (i) the idle channel subset, and (ii) the missed-
 detection probability of each channel in the idle channel sub-
 set. Notice that only channels that have been jointly (i.e. by
 both nodes) identified as idle can be used. The two missed-
 detection probabilities will be used to compute the aver-
 age missed-detection probability of each channel in the joint
 idle channel subset. The average missed-detection probab-
 ility is used to compute the reliability of the channel (ρ_i) as
 follows:

$$\rho_i = \frac{1}{P_{mdi}} \quad (6)$$

The main aim of the next step, channel estimation, is to
 estimate the SNR of each channel in the joint idle channel
 subset.

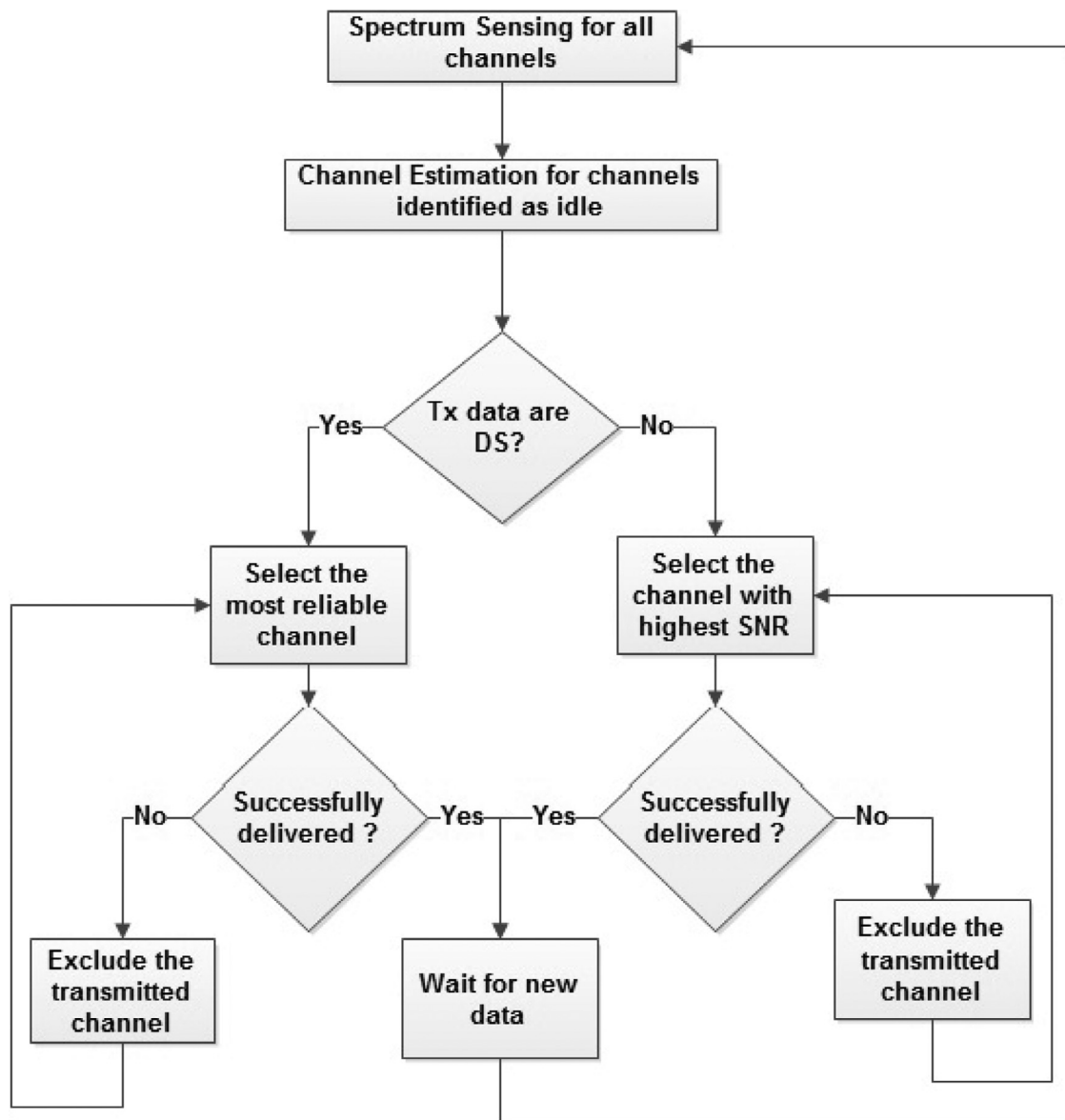


Fig. 3. The flow chart of the proposed channel selection mechanism for CR-based SG communications.

375 By performing the first two steps, the two communicat- 389
 376 ing nodes will have (i) a small subset of the channels that 390
 377 are commonly identified as idle, (ii) the reliability value of 391
 378 each channel in the obtained subset, and (iii) the SNR of each 392
 379 channel in the obtained subset. 393

380 Now, once the transmitter node has data ready for trans- 394
 381 mission, the transmitter should examine the type of the 395
 382 transmitted data: i.e. “is it DS or NDS?”. According to the data 396
 383 type, the transmitter should select the proper channel. For DS 397
 384 data, the channel should be the most reliable channel, since 398
 385 DS data are very important and should be successfully deliv- 399
 386 ered as soon as possible. On the other hand, for NDS data, the 400
 387 channel with highest SNR should be selected, since this type
 388 of data is large and requires high data rates.

The last step in the proposed mechanism is to ensure 389
 the delivery of the transmitted data, where we assume that 390
 the receiver node will inform the transmitter node if the 391
 data have been successfully delivered or not through an ac- 392
 knowledgment mechanism. If the data have not been suc- 393
 cessfully delivered, the transmitter node will re-transmit the 394
 data and exclude the previously used channel from the selec- 395
 tion mechanism. 396

4. Performance analysis and simulation results 397

In this section, the performance of the proposed channel 398
 selection mechanism is explored through simulation results. 399
 The results of other three channel selection mechanisms will 400

Table 1
Simulation parameters.

| | |
|-----------------------------------|---------------------------------------|
| Channel bandwidth (B) | 20 kHz |
| Probability of channel occupancy | 0.5 |
| Detection probability (P_d) | A uniform random variable on [0.5, 1] |
| False-alarm probability (P_f) | A uniform random variable on [0, 0.5] |
| Probability of DS data | 0.2 |
| Probability of NDS data | 0.8 |
| No. of channels | Indicated on each figure |
| Average SNR | Indicated on each figure. |
| No. of iterations | 10^4 |

401 be shown for comparison purpose. The three mechanisms are
402 as follows:

- 403 • **SNR-based channel selection mechanism:** It implies
404 that the transmit channel will be selected based on the
405 instantaneous SNR. Specifically, the channel whose SNR is
406 the maximum will be selected from those that have been
407 identified as idle by both the transmitter and the receiver.
- 408 • **Reliability-based channel selection mechanism:** The
409 most reliable channel is chosen for data transmission,
410 where the channel whose instantaneous missed-detection
411 probability is the minimum will be selected from channels
412 that have been identified as idle by both the transmitter
413 and the receiver.
- 414 • **Random channel selection mechanism:** Unlike the
415 above mechanisms, this mechanism randomly selects
416 the transmit channel from the channels that have been
417 identified as idle by both the transmitter and the
418 receiver.

419 The reader should notice that the proposed mechanism
420 can be considered as a combination of the first two mecha-
421 nisms. In the proposed mechanism, the type of the transmit-
422 ted data is used to identify the selection criteria (reliability
423 or data rate) most appropriate for the data to be delivered.
424 On the other hand, the three mechanisms described above
425 do not explicitly consider the data type for channel selection
426 process.

427 In the simulation setup, we consider the bandwidth of
428 each channel equals to 20 kHz. The probability that a channel
429 is occupied is identical for all channels and equals to 0.5,
430 which is a reasonable assumption, especially when no prior
431 statistics about channels' usage are available. Regarding
432 the sensing ability of the distributed nodes, the detection
433 probability and the false-alarm probability are modeled as
434 uniform random variables over the range [0.5, 1] and [0,
435 0.5], respectively. The probability that the transmitted data
436 is delay-sensitive is 0.2, while the probability of non-delay
437 sensitive data is 0.8. Regardless of the employed mechanism,
438 we consider that the transmitted data will not be delivered
439 in two cases: (i) if none of the channels has been identified
440 as idle by both transmitter and receiver nodes, and (ii) if
441 the transmit channel is actually occupied by a primary user.
442 The reader should notice that in case that the transmitted
443 data have not been delivered, the transmitter node should
444 retransmit the same data in the next transmission phase.
445 The shown results are taken from 10^4 transmission rounds.
446 **Table 1** summarizes the simulation parameters.

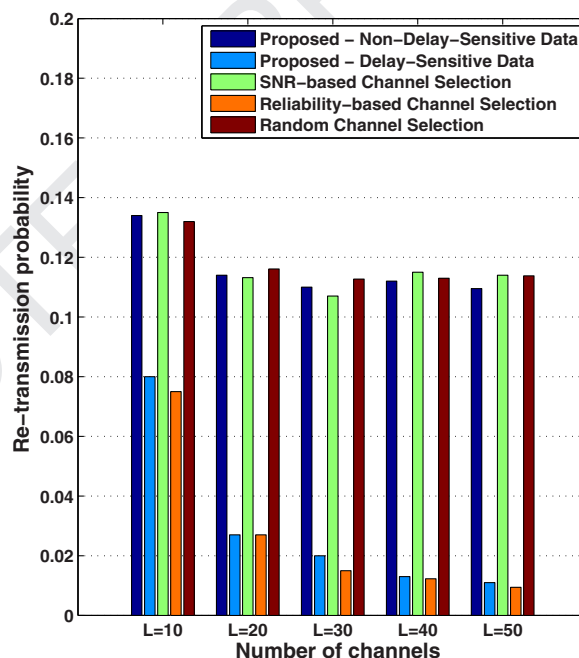


Fig. 4. The retransmission probability versus the total number of channels for the considered channel selection mechanisms. Average SNR = 5.

447 In Fig. 4, the retransmission probability versus the total
448 number of channels for the considered channel selection
449 mechanisms is shown. As the proposed mechanism is
450 the only one that distinguishes between the data type, it
451 is plotted in Fig. 4 separately for delay-sensitive and non-
452 delay-sensitive data types. The random and the SNR-based
453 mechanisms show a poor performance whatever the chan-
454 nels' number due to not including the reliability of the trans-
455 mit channel in the selection process. On the other hand, the
456 reliability-based mechanism has the best performance, as it
457 achieves the minimum retransmission probability. Regarding
458 our proposed mechanism, it achieves the minimum retrans-
459 mission probability (as in reliability-based mechanism) for
460 only delay-sensitive data since this type of data cannot tol-
461 erate long delays. For non-delay-sensitive data, our proposal
462 allows for higher retransmission probability (equals to the
463 SNR-based mechanism).

464 The total achievable data rate versus the number of chan-
465 nels for all the considered mechanisms is shown in Fig. 5.

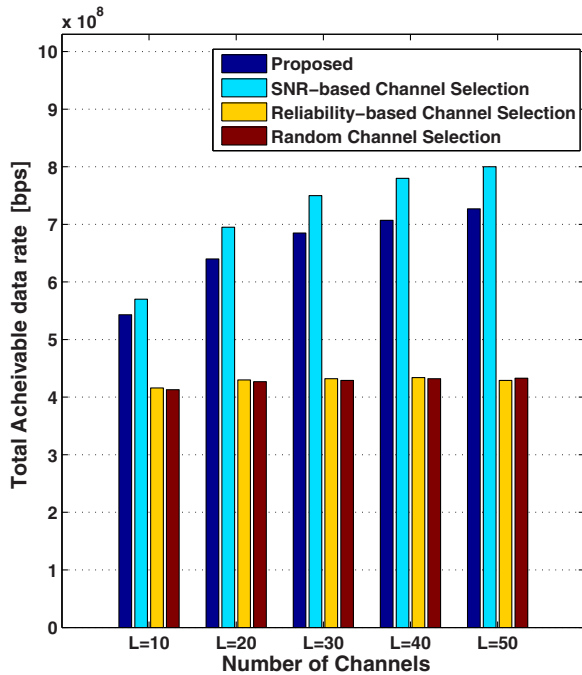


Fig. 5. The total achievable data rate versus the number of channels for all the considered mechanisms. Average SNR = 5.

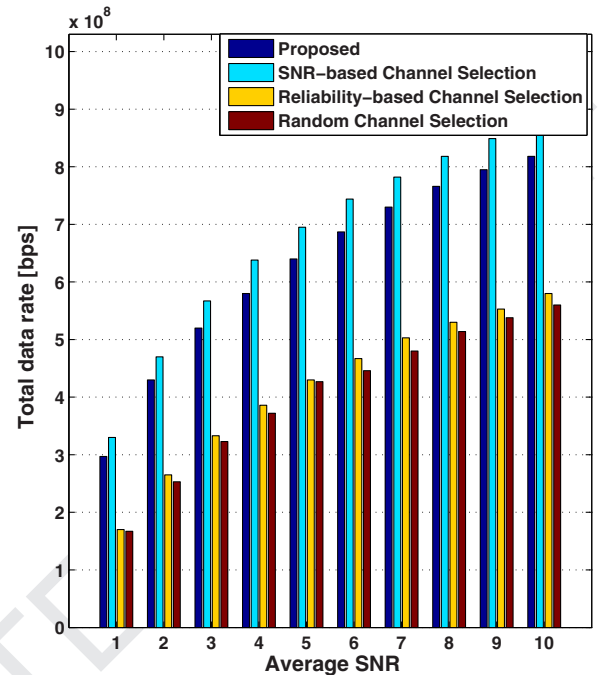


Fig. 6. The total data rate for all mechanisms versus the average SNR. Number of channels = 20.

466 Generally, for mechanisms that adopt the SNR in the selection (i.e. SNR-based and the proposed mechanism), the total data rate increases as the number of channels increases, while the other mechanisms (reliability-based and random mechanisms) achieve approximately constant total data rate although the number of channels is increasing. Intuitively, the SNR-based mechanism attains the maximum total data rate as it always chooses the maximum SNR for data transmission. However, the proposed mechanism is able to achieve high total data rate that is near to that achieved by the SNR-based mechanism. This is because the proposed mechanism partially (only for non-delay-sensitive data) depends on the channel SNR.

479 Comparing the results in Figs. 4 and 5, we can see that the reliability-based mechanism achieves the minimum retransmission probability, while it achieves low data rate. On the other hand, the SNR-based mechanism provides the maximum total data rate, while its achievable retransmission probability is high compared to the other mechanisms. In the light of this trade-off between reliability and data rate, the proposed mechanism achieves the balance for the trade-off, where it is able to achieve low retransmission probability for delay-sensitive data, and high data rate for non-delay-sensitive data.

490 The total data rate for all mechanisms versus the average SNR is shown in Fig. 6. Clearly, the total data rate increases for all mechanisms as the average SNR increases. The proposed mechanism achieves data rate that is near to the maximum value (achieved by the SNR-based mechanism).

495 In Fig. 7, the retransmission probability versus the occupancy probability of the channels for the proposed mechanism is depicted. The occupancy probability is the

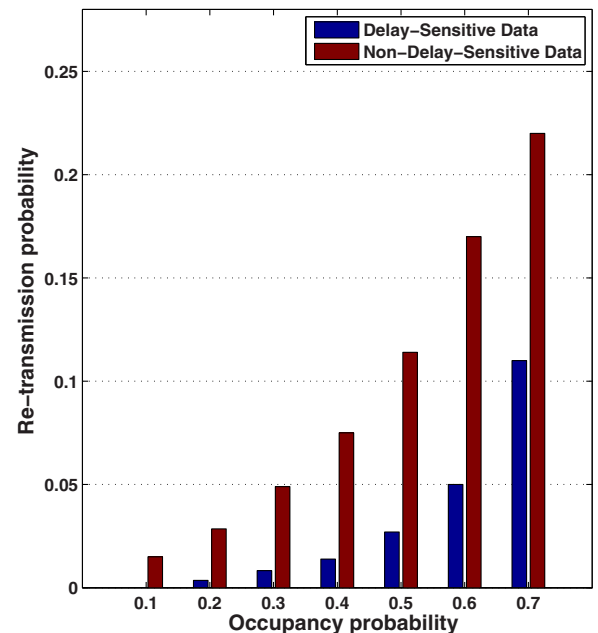


Fig. 7. The retransmission probability versus the occupancy probability of the channels for the proposed mechanism. Average SNR = 5, Number of channels = 20.

probability that the transmit channel is used by a primary user. It should be underlined that due to the fact that the proposed mechanism changes the selection criteria according to the data type, the delay-sensitive data can experience lower

retransmission probability than the non-delay sensitive data.

5. Conclusions

This paper investigates the channel selection problem of cognitive radio based smart grid communications in the distribution section. SNR-based mechanisms can offer high data rate, while the cost is in longer delays due to not considering the reliability of the transmit channels. On the other hand, reliability-based mechanisms are able to shorten the transmission delay, and the corresponding data rate loss is high. The proposed mechanism is able to achieve balance in this trade-off as it adapts the selection criteria based on the data type. The proposed mechanism selects the most reliable channel for delay-sensitive data, and the maximum SNR channel for non-delay-sensitive data. Simulation results have proved the high performance of the proposed mechanism compared to the other mechanisms. As a potential future work, the proposed channel selection mechanism can be evaluated in real environments taking into account the different sensing abilities of the transmitting nodes, the variable activity of licensed users and the potential congestion in case of large number of nodes.

References

- [1] Y. Yan, et al., A survey on smart grid communication infrastructures: motivations, requirements and challenges, *IEEE Commun. Surv. Tutorials* 15 (1) (2013) 5–20.
- [2] A.A. Khan, M.H. Rehmani, M. Reisslein, Cognitive radio for smart grids: survey of architectures, spectrum sensing mechanisms, and networking protocols, *IEEE Commun. Surv. Tutorials* (2015).in press
- [3] V.C. Gungor, D. ahin, Cognitive radio networks for smart grid applications: a promising technology to overcome spectrum inefficiency, *IEEE Vehicular Technol. Mag.* 7 (2) (2012) 41–46.
- [4] V. Kouhदारagh, D. Tarchi, A.V. Coralli, G.E. Corazza, Cognitive radio based smart grid networks, in: 2013 24th Tyrrhenian International Workshop on Digital Communications-Green ICT (TIWDC), IEEE, 2013, pp. 1–6.
- [5] M. Carlesso, A. Antonopoulos, F. Granelli, C. Verikoukis, Uplink scheduling for smart metering and real-time traffic coexistence in LTE networks, in: 2015 IEEE International Conference on Communications (ICC), IEEE, 2015, pp. 820–825.
- [6] J. Mitola III, G.Q. Maguire Jr., Cognitive radio: making software radios more personal, *IEEE Personal Commun.* 6 (4) (1999) 13–18.
- [7] C. Cordeiro, et al., IEEE 802.22: the first worldwide wireless standard based on cognitive radios, in: First IEEE International Symposium on New Frontiers in Dynamic Spectrum Access Networks, DySPAN, 2005.
- [8] T. Ycek, H. Arslan, A survey of spectrum sensing algorithms for cognitive radio applications, *IEEE Commun. Surv. Tutorials*, 11 (1) (2009) 116–130.
- [9] D. Cabric, S.S. Mishra, R.W. Brodersen, Implementation issues in spectrum sensing for cognitive radios, *IEEE Conference Record of the Thirty-Eighth Asilomar Conference on Signals, Systems and Computers* (2004).
- [10] M. Erol-Kantarci, H.T. Mouftah, Energy-efficient information and communication infrastructures in the smart grid: a survey on interactions and open issues, *IEEE Commun. Surv. Tutorials*, 17.1 (2015) 179–197.
- [11] W. Meng, R. Ma, H.H. Chen, Smart grid neighborhood area networks: a survey, *IEEE Netw.* 28.1 (2014) 24–32.
- [12] A. Ghassemi, S. Bavarian, L. Lampe, Cognitive radio for smart grid communications, in: *Smart Grid Communications (SmartGridComm)*, IEEE, 2010.
- [13] R. Yu, C.R. Rui, X. Zhang, L. Zhou, K. Yang, Hybrid spectrum access in cognitive-radio-based-smart-grid communications system, *IEEE Syst. J.* 8 (2) (2014).
- [14] R.L. Deng, J.m. Chen, X.H. Cao, Y.Z. Zhang, S. Maharjan, S. Gjessing, Sensing-performance tradeoff in cognitive radio enabled smart grid, *IEEE Trans. Smart Grid* 4 (1) (2013).

- [15] G.A. Shah, V.C. Gungor, O.B. Akan, A cross-layer QoS-aware communication framework in cognitive radio sensor networks for smart grid applications, *IEEE Trans. Ind. Inform.* 9 (3) (2013) 1477–1485.
- [16] H.G. Wang, Y. Qian, H. Sharif, Multimedia communications over cognitive radio networks for smart grid applications, *IEEE Wireless Commun.* 20 (4) (2013) 125–132.
- [17] J.F. Huang, H.G. Wang, Y. Qian, C.G. Wang, Priority-based traffic scheduling and utility optimization for cognitive radio communication infrastructure-based smart grid, *IEEE Trans. Smart Grid* 4 (1) (2013) 78–86.
- [18] H. Jiang, L.F. Lai, R.F. Fan, H.V. Poor, Optimal Selection of Channel Sensing Order in Cognitive Radio, *IEEE Trans. Wireless Commun.* 8 (January (1)) (2009).
- [19] Y.L. Zou, J. Zhu, B.Y. Zheng, Y.D. Yao, An adaptive cooperation diversity scheme with best-relay selection in cognitive radio networks, *IEEE Trans. Signal Process.* 58 (October (10)) (2010).
- [20] S. Liu, I. Ahmad, Y. Bai, Z.Y. Feng, Q.X. Zhang, Y.F. Zhang, A novel cooperative sensing based on spatial distance and reliability clustering scheme in cognitive radio system, in: *IEEE Vehicular Technology Conference (VTC Fall)*, 2013.
- [21] J.Z. Zhang, F.Q. Yao, Y.X. Liu, L. Cao, Robust route and channel selection in cognitive radio networks, in: *IEEE International Conference on Communication Technology (ICCT)*, 2012.
- [22] J.A. Abolarinwa, N.M.A. Latiff, S.K.S. Yusof, N. Faisal, Learning-based algorithm for energy-efficient channel decision in cognitive radio-based wireless sensor networks, in: *International Conference on Frontiers of Communications, Networks and Applications (ICFCNA 2014)*, Malaysia, IET, 2014, pp. 1–6.
- [23] T. Chen, H. Zhang, M.D. Katz, Z. Zhou, Swarm intelligence based dynamic control channel assignment in CogMesh, in: *IEEE International Conference on Communications Workshops, 2008 (ICC Workshops'08)*, IEEE, 2008, pp. 123–128.
- [24] A. Mesodiakaki, et al., Performance analysis of a cognitive radio contention-aware channel selection algorithm, *IEEE Trans. Vehicular Technol.* 64 (May (5)) (2015) 1958–1972.
- [25] A. Mesodiakaki, et al., Energy-efficient contention-aware channel selection in cognitive radio ad-hoc networks, *CAMAD 2012* (September 2012).
- [26] F.F. Digham, M.-S. Alouini, M.K. Simon, On the energy detection of unknown signals over fading channels, *IEEE Trans. Commun.* 55 (1) (2007) 21–24.
- [27] S. Althunibat, F. Granelli, On the reduction of power loss caused by imperfect spectrum sensing in OFDMA-based cognitive radio access, *IEEE Global Communications Conference (GLOBECOM)* (2012).
- [28] Y. Gao, Y. Jiang, Performance analysis of a cognitive radio network with imperfect spectrum sensing, in: *IEEE Conference on Computer Communications Workshops INFOCOM*, 2010.
- [29] S. Althunibat, M.D. Renzo, F. Granelli, Towards energy-efficient cooperative spectrum sensing for cognitive radio networks: an overview, *Telecommun. Syst.* 59 (1) (2015) 77–91.
- [30] W.P. Luan, D. Sharp, S. Lancashire, Smart grid communication network capacity planning for power utilities, in: *Transmission and Distribution Conference and Exposition, IEEE PES*, 2010.
- [31] Y. Yan, Y. Qian, H. Sharif, A survey on smart grid communication infrastructures: motivations, requirements and challenges, *IEEE Commun. Surv. Tutorials* 15 (1) (2012) 5–20.
- [32] A. Aggarwal, S. Kunta, P.K. Verma, A proposed communications infrastructure for the smart grid, in: *Innovative Smart Grid Technologies (ISGT)*, 2010.



Saud Althunibat is an Assistant Professor at the Department of Communications Engineering of Al-Hussein Bin Talal University, Jordan. He received the B.Sc. in Electrical Engineering/Communications in 2004 from Mutah University, Jordan, the M.Sc. Degree in Electrical Engineering/Communications in 2010 from the University of Jordan, Jordan, and the Ph.D. degree in Telecommunications in 2014 from the University of Trento, Italy. From 2011 to 2014, he has been a Marie-Curie Early-stage researcher working within the GREENET project at University of Trento. He is a reviewer in many international journals and a TPC member in many international conferences. He is the recipient of the best-paper award in IEEE CAMAD 2012, and was selected as exemplary reviewer in IEEE Communication Letters 2013. His research interests include Cognitive Radio Networks, Physical-Layer Security, Resource Allocation and Heterogeneous Networks.

645
646
647
648
649
650
651
652
653
654
655
656



Qi Wang received his M.Sc. degree in Electrical and Computer Engineering from Sungkyunkwan University, S. Korea in 2011, and Ph.D. degree in Information and Communications Technology from University of Trento, Italy in 2015. He was a Visiting Scholar at North Carolina State University from November 2013 to April 2014. His research interests include WiMAX, Wireless Sensor Network, and Wireless Communications in Smart Grid, Architecture Modelling of Smart Grid, Renewable Energy System, and Fast Power Charging Station for Electric Vehicles.



Fabrizio Granelli is IEEE ComSoc Distinguished Lecturer for 2012–2015, and Associate Professor at the Department of Information Engineering and Computer Science (DISI) of the University of Trento (Italy). He received the “Laurea” (M.Sc.) degree in Electronic Engineering and the Ph.D. in Telecommunications Engineering from the University of Genoa, Italy, in 1997 and 2001, respectively. In August 2004, August 2010 and April 2013, he was visiting professor at the State University of Campinas (Brasil). He is author or co-author of more than 140 papers with topics related to networking. He was guest-editor of ACM Journal on Mobile Networks and Applications, ACM Transactions on Modeling and Computer Simulation, and Hindawi Journal of Computer Systems, Networks and Communications. He was TPC Co-Chair of IEEE GLOBECOM Symposium on “Communications QoS, Reliability and Performance Modeling” in the years 2007, 2008, 2009 and 2012.

657
658
659
660
661
662
663
664
665
666
667
668
669
670
671
672
673
674

UNCORRECTED PROOF