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Mobility robustness optimization in self-organizing LTE femtocell networks

Wei Zheng^{1*}, Haijun Zhang^{1*}, Xiaoli Chu² and Xiangming Wen¹

Abstract

Femtocell is a promising solution for enhancing the indoor coverage and capacity in wireless networks. However, for the small size of femtocell and potentially frequent power on/off, existing handover schemes may not be reliable enough for femtocell networks. Moreover, improper handover parameters settings may lead to handover failures and unnecessary handovers, which make it necessary to enhance the mobility robustness for femtocells. In this article, we propose a gradient method and cost function-based mobility robustness optimization scheme for long term evolution (LTE) femtocell self-organizing networks. Moreover, signalling overhead of the scheme is analyzed. Simulation results show that the proposed scheme has a better performance than the fixed parameters method in terms of reduced the number of handover failures and unnecessary handovers with limited signalling modifications.

1 Introduction

As macrocells have limited indoor coverage, while above 50% of the voice services and 70% of the data traffics happen indoors [1], offloading the traffics from macrocells is badly needed. Femto base station (FBS), which is also known as HeNB [2], is a low power wireless access point that can meet the need. A femto user equipment (FUE) can save its power consumption for the smaller distance between transmitter and receiver in femtocell systems. Femtocell is more harmony to the environment due to the lower carbon emission. Furthermore, femtocells use the cable and DSL as the backhaul to access the service provider's network. As a result, it can improve the utility of the networks. Moreover, femtocell reduces the mobile operator's OPEX because of the user's self-deployment and self-configuration of FBS. However, the deployment of femtocell can also introduce technical challenges such as interference management and mobility management to long term evolution (LTE) and LTE-Advanced (LTE-A) networks.

In traditional mobility management schemes, the handover parameters are set by the mobile operator, which is inefficiency and inaccurate. Recently, mobility robustness optimization (MRO) as one of the usecases in the

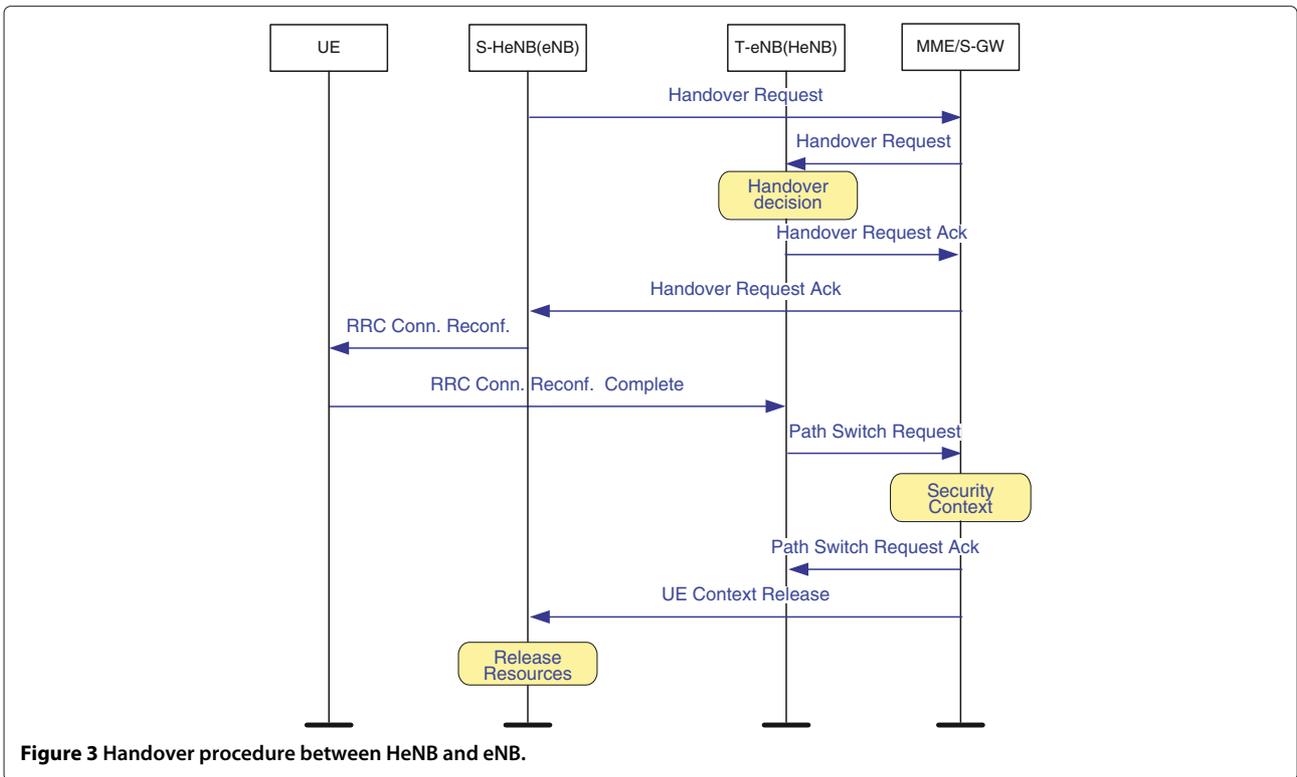
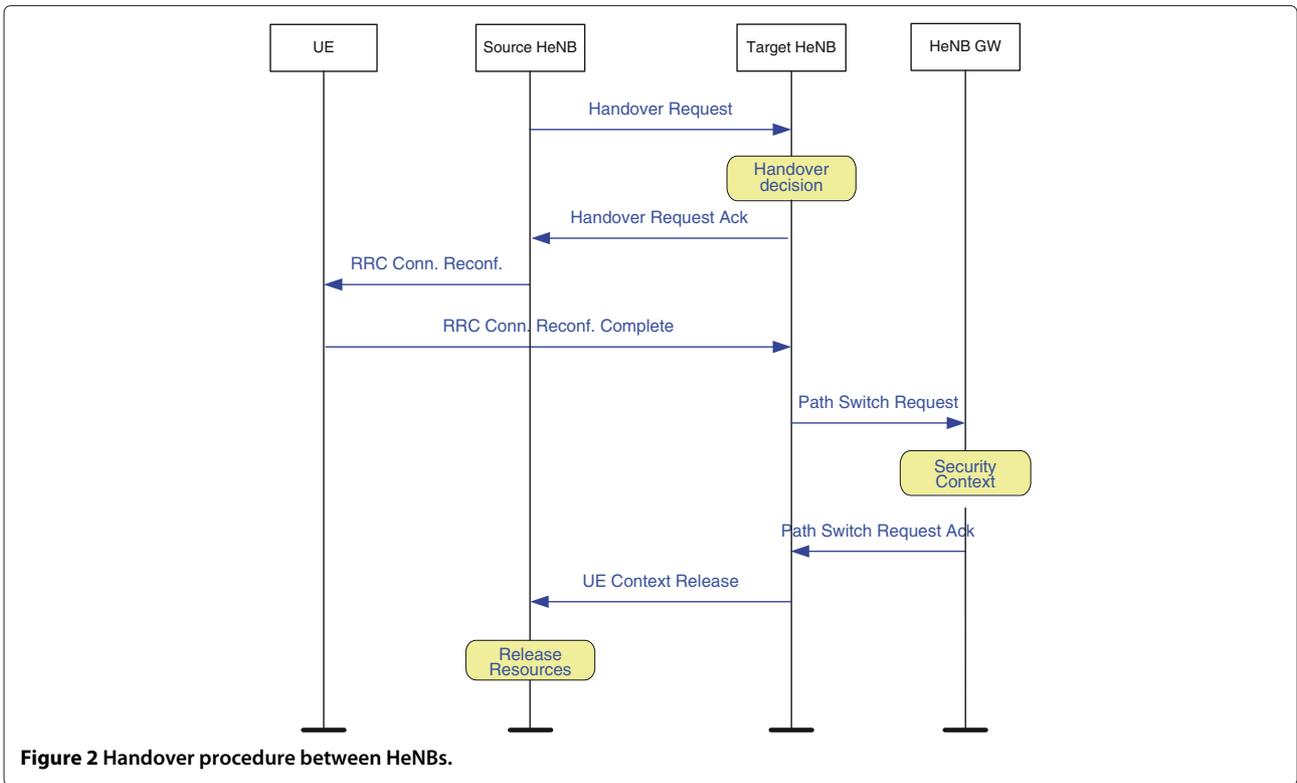
self-organization networks has been studied to reduce radio link failures (RLFs) and unnecessary handovers due to improper handover parameter settings [2]. However, traditional handover optimization is inappropriate for different characteristics in macrocell-femtocell hybrid deployment compared with those in macrocell due to the dynamic channel conditions and different mobility patterns. On one hand, the large number of femtocells makes it difficult to configure and maintains handover parameter optimization using existing schemes; on the other hand, since femtocell could be frequently turned on/off, channel conditions and neighboring cell list change frequently. To the best of authors' knowledge, there are few studies focusing on mobility robustness optimization in femtocell networks.

There are a lot of researches concerning the handover optimization in LTE macrocells. In [3], a cost function based handover parameter optimization scheme incorporating cell load, UE's velocity and service type, is proposed for 3GPP LTE macrocells. In [4], admission control strategy and handover self-optimization are considered to optimize the handover performance. Joint hysteresis and time to trigger (TTT) optimization scheme is investigated in [5] to reduce handover failures. Most of these studies optimize the handover parameters, such as hysteresis and TTT, based on self-optimization techniques.

MRO in macrocells has been extensively studied. In [6], hysteresis adapting based MRO scheme is proposed

*Correspondence: zhengweius@gmail.com; zhj@bupt.edu.cn

¹ Beijing Key Laboratory of Network System Architecture and Convergence, Beijing University of Posts and Telecommunications, Beijing 100876, China
Full list of author information is available at the end of the article



2.2 Handover procedures and parameters

Since X2 interface is introduced in Figure 1, which is different from the early version of E-UTRAN network architecture [21], HeNB related handover procedure should be revised in both HeNB-HeNB handover and eNB-HeNB handover. Here we will present handover call flow based on the current E-UTRAN architecture [20] as illustrated in Figures 2 and 3. Figure 2 shows the handover procedures between HeNBs and Figure 3 presents the handover flow between eNB and HeNB [16].

Handover in LTE/LTE-A femtocells is assumed as network-controlled hard handover with the assistance of UE [7,22]. Handover procedure in LTE femtocells can be divided into four phases: handover measurement, measurement report, handover decision, handover execution. FUE periodically measures the reference signal received powers (RSRPs) of the serving cell and neighbor cells according to the measurement configuration and neighbor cell list sent by FBS. According to [23], if the entering condition of A3 event (neighbor cell is better than serving cell) in (1) is satisfied and lasts for a duration of TTT, then

the A3 event will be triggered. After receiving the report from FUE, the serving cell will handover the FUE to the target cell. Otherwise, the FUE will leave A3 event if the leaving condition in (2) is satisfied.

$$RSRP_t + CIO_t - Hys > RSRP_s + CIO_s + Off \quad (1)$$

$$RSRP_t + CIO_t + Hys < RSRP_s + CIO_s + Off \quad (2)$$

where $RSRP_s$ and $RSRP_t$ are the signal strengths of the serving cell and the target cell measured by UE, respectively, CIO_s and CIO_t are the cell individual offset (CIO) of the serving cell and the target cell, respectively, Hys is the hysteresis value of A3 event, and Off is the offset parameter of A3 event. Given the value of Hys and Off , UE tends to handover to the cell whose CIO plus RSRP is higher.

2.3 Definition and detection of MRO events

Inappropriate handover parameters can lead to handover failures. Three types of RLF related to MRO are defined

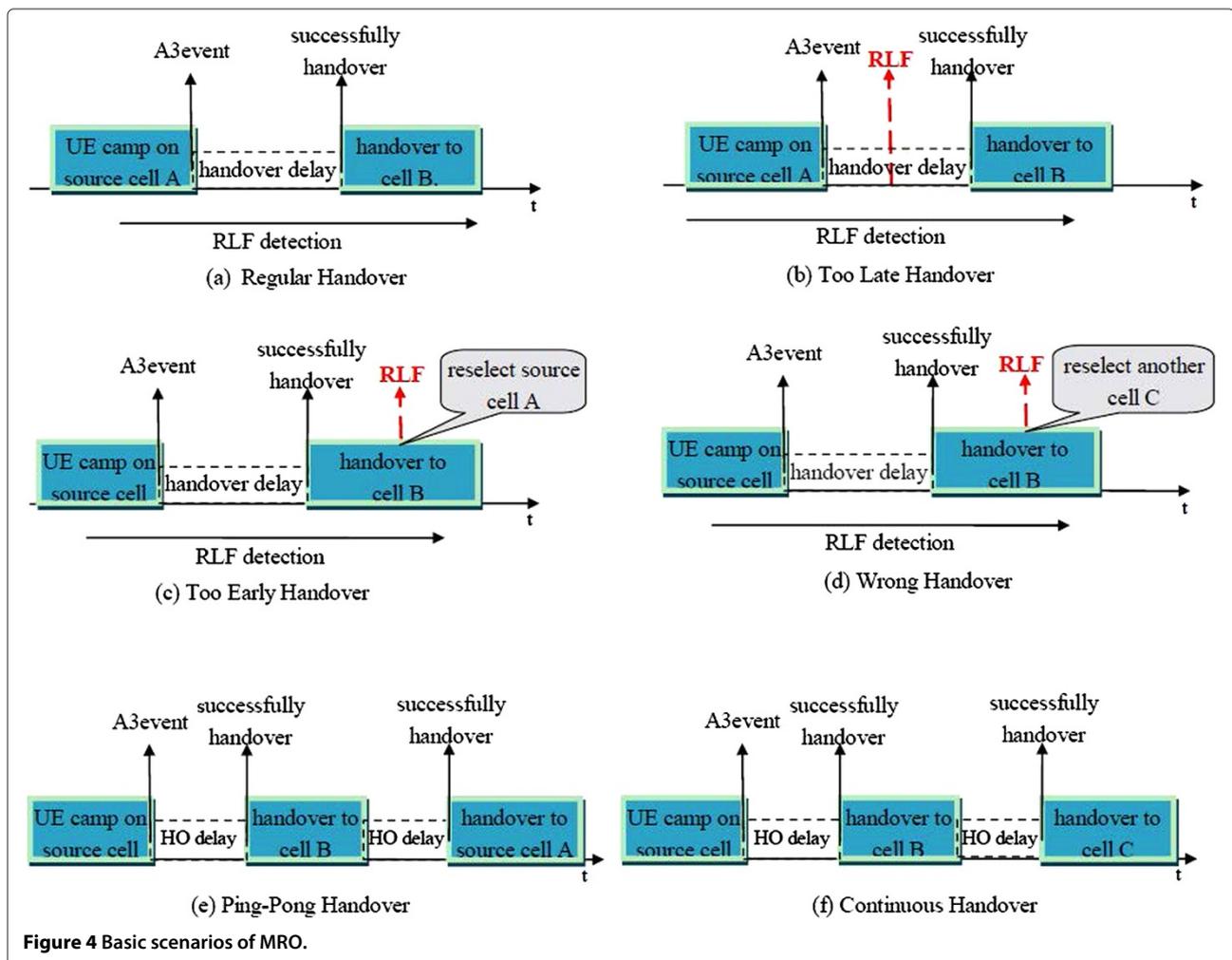


Figure 4 Basic scenarios of MRO.

in LTE SON: too late handover, too early handover, and wrong handover. Two types of unnecessary handover are defined in MRO: ping-pong handover and continuous handover. The characteristics of the mobility related unsuccessful handovers and the unnecessary handovers are described in Figure 4 [2]:

- (b) Too late handover: RLF occurs in serving cell before handover or during the handover procedure, then UE reconnect to the target cell (different from the serving cell).
- (c) Too early handover: RLF occurs shortly after a successful handover to the target cell, and then UE reconnect to the serving cell.
- (d) Wrong handover: RLF occurs shortly after a successful handover to the target cell, and then UE reconnect to another cell (neither the serving cell nor the target cell).
- (e) Ping-pong handover: handover to the serving cell from the target cell shortly after a successful handover to the target cell.
- (f) Continuous handover: handover to another cell (neither the serving cell nor the target cell) shortly after a successful handover to the target cell.

The proposed procedure of RLF detection related to handover and unnecessary handover is shown in Algorithm 1.

In Algorithm 1, Timer_UE_ID is the timer for UE, RLF_UE_ID is Cell ID of UE who undergoes RLF, Last_Visited_Cell_ID is the Cell ID the UE last visited, and Reconnected_Cell_ID is the Cell ID the UE reconnected after encountering RLF. The RLF event is detected and reported by UE once the RLF occurs, and then Timer_UE_ID, RLF_UE_ID, Last_Visited_Cell_ID, together with Reconnected_Cell_ID, are reported to the FBS/MBS, and are finally collected by the SON module where the MRO algorithm is implemented.

3 Gradient method based femtocell MRO

There are many handover parameters in femtocell MRO. Without loss of generality, we choose the CIO as the handover parameter to optimize in this article.

3.1 Cost function

In order to reduce RLFs and unnecessary handovers in femtocells, we propose a cost function based algorithm, which is defined as follows:

$$C = \sum_{i=1}^5 \alpha_i N_i \quad (3)$$

where $N_1, N_2, N_3, N_4,$ and N_5 denote the numbers of ping-pong handover, continuous handover, too late handover, too early handover and wrong handover, respectively, and

Algorithm 1 Detection of RLF and Unnecessary Handover.

```

1: Handover from cell A to cell B, start Timer_UE_ID;
2: if RLF Detected then
3:   Get the RLF_UE_ID and the Reconnected_Cell_ID;
4:   Stop Timer_UE_ID;
5:   if then
6:     if Reconnected_Cell_ID==Last_Visited_Cell_ID
       then
7:       Too_Early_Handover_Counter++;
8:     end if
9:     else if Timer_UE_ID<Time_Threshold then
10:      Wrong_Handover_Counter++;
11:    else
12:      Too_Late_Handover_Counter++;
13:    end if
14:  else if Handover from cell B to cell C then
15:    Stop Timer_UE_ID;
16:    if Timer_UE_ID<Timer_Threshold then
17:      if Target_Cell_ID==Last_Visited_Cell_ID then
18:        Ping-Pong_handover_Counter++;
19:      else
20:        Continuous_Handover_Counter++;
21:      end if
22:    end if
23:  end if
    
```

α_i denotes the weight factor of N_i . The value of N_i is determined by Algorithm 1.

3.2 Gradient method based MRO

In this section, we introduce the gradient algorithm [18,19,24] into femtocell MRO.

The optimization algorithm to be used in the minimization of the cost function is derived from first principles for one single parameter. It is easy to extend the derivation to the case of several parameters. Consider the general case of a cost function C to be minimized with respect to a parameter denoted by w . Let w^* be the value of w which minimizes C . Evaluating $C(w^*)$ using the Taylor series expansion about any value of w gives [19],

$$C(w^*) = C(w) + (w^* - w)C'(w) + \frac{(w^* - w)^2}{2}C''(w) \quad (4)$$

As $C(w^*)$ is a minimum point of C , then differentiating (4) with respect to w^* and letting the result equal 0 gives:

$$w^* = w - \frac{C'(w)}{C''(w)} \quad (5)$$

Consider a small change δw (> 0) in the value of the parameter w to give a new parameter value $w + \delta w$. The value of the cost function can then be approximated as,

$$C(w + \delta w) = C(w) + \delta w C'(w) + \frac{\delta w^2}{2} C''(w) \quad (6)$$

Consider a decrease in the value of w by $\delta w > 0$ to give a new parameter value $w - \delta w$. Once again the value of the cost function for this parameter value can be expressed as:

$$C(w - \delta w) = C(w) - \delta w C'(w) + \frac{\delta w^2}{2} C''(w) \quad (7)$$

Solving (6) and (7) for $C''(w)$ and $C'(w)$, we have,

$$C''(w) = \frac{C(w + \delta w) + C(w - \delta w) - 2C(w)}{\delta w^2} \quad (8)$$

$$C'(w) = \frac{C(w + \delta w) - C(w - \delta w)}{2\delta w} \quad (9)$$

Substituting (8) and (9) into (5), we get

$$w^* = w - \frac{\delta w(C(w + \delta w) - C(w - \delta w))}{2(C(w + \delta w) + C(w - \delta w) - 2C(w))} \quad (10)$$

Based on the above gradient algorithm, the pseudo code of the MRO algorithm is described in Algorithm 2.

Algorithm 2 Gradient Algorithm based MRO.

- 1: At time $t1$, the parameter value is w and the value of the cost function is $C(w, t1)$ based on Algorithm 1 network measurements of performance metrics according to Algorithm 1.
 - 2: At time $t1$, the value of w is changed to $w + \delta w$.
 - 3: At time $t2 = t1 + \delta t$, $\delta t > 0$, the value of the cost function $C(w + \delta w, t2)$ is evaluated based on Algorithm 1.
 - 4: At time $t2$, the parameter w is changed to $w - \delta w$.
 - 5: At time $t3 = t2 + \delta t$, the cost function is evaluated based on Algorithm 1.
 - 6: At time $t3$, a new value of w is calculated using Equation (10), where the values of $C(w)$, $C(w + \delta w)$ and $C(w - \delta w)$ are given respectively by $C(t1)$, $C(t2)$, $C(t3)$.
 - 7: These steps constitute one iteration of the algorithm and the algorithm can be iterated till convergence.
-

3.3 Signalling overhead analysis

For the gradient method used in Algorithm 2, the number of iterations required to minimize the cost function C is upper bounded by [25]

$$\frac{C(w_0) - C^*}{\gamma} + \log_2 \left(\log_2 \left(\frac{\varepsilon_0}{\varepsilon} \right) \right) \quad (11)$$

where w_0 is the initial value of w ; C^* is the minimum value of C ; γ is the smallest decrease of C in each iteration; ε_0

Table 1 Simulation parameters

Parameter	Value
MBS number	1
FBS number	20
Femtocell layout	A modified dual strip model (five apartments in a row)
Macrocell radius	577 m
Femtocell radius	6 m
Carrier frequency	2 GHz
System bandwidth	10 MHz
Antenna gain	UE: 0 dBi, MBS: 15 dBi FBS: omni-direction 5 dBi [27]
Log-normal shadowing	Standard deviation FBS-UE 10 dB Standard deviation MBS-UE 8 dB [27]
MBS transmitting power	46 dBm
FBS transmitting power	20 dBm
UE mobility model	Random way-point model [26]
MBS UE velocity	Uniform (0, 30) km/h
FBS UE velocity	Uniform (0, 5.4) km/h
UE distribution and mobility range	indoor UE: within FBS coverage area outdoor UE: within in the MBS coverage area
Traffic model	indoor UE: Full queue traffic outdoor UE: voice traffic
Outdoor UE to MBS	PL(dB) = 15.3 + 37.6lgR, R in m [27]
Indoor UE to MBS	PL(dB) = 15.3 + 37.6lgR + Low [27]
UE to FBS (UE is inside the same apt as FBS)	PL(dB) = 38.46 + 20lgR + q*Liw [27]
Indoor UE to FBS (UE inside a different apt as FBS)	PL(dB) = max(15.3 + 37.6lgR, 38.46 + 20lgR) + q*Liw + p*Low [27]
Sampling time	10 ms
Minimum coupling losses	70 dB
Out of Sync threshold	-4.5 dB
In Sync threshold	-2.1 dB

is a constant depending on w_0 ; and ε is the convergence tolerance.

For the detection of RLF and unnecessary handover in Algorithm 1, only three decision processes are needed to detect the handover event (e.g., lines 2, 5, 6 in Algorithm 1 to detect the too early handover). Moreover, the factors for detection in Algorithm 1, Timer_UE_ID, RLF_UE_ID, Last_Visited_Cell_ID, together with Reconnected_Cell_ID, are reported to the FBS/MBS,

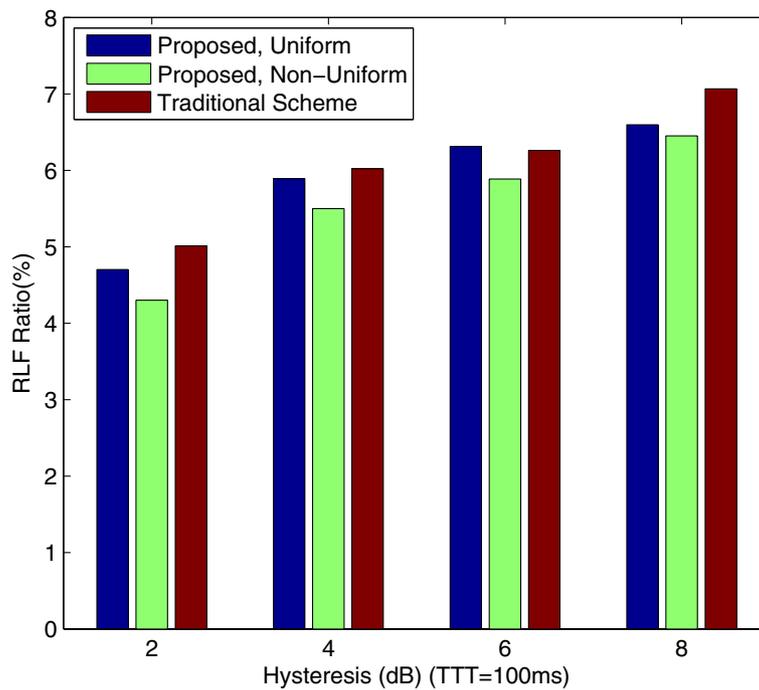


Figure 5 RLF ratio versus hysteresis (TTT = 100 ms).

when a RLF event is detected. A similar signalling overhead is needed for unnecessary handover detection. Therefore, both Algorithms 1 and 2 can be implemented with little modifications of existing LTE/LTE-A handover protocol.

4 Performance evaluation

The gradient based MRO algorithm proposed in this article is evaluated in a femtocell network by simulation. A macrocell and F femtocells coexisting scenario for the simulation is considered, and F is assumed to be 20. The

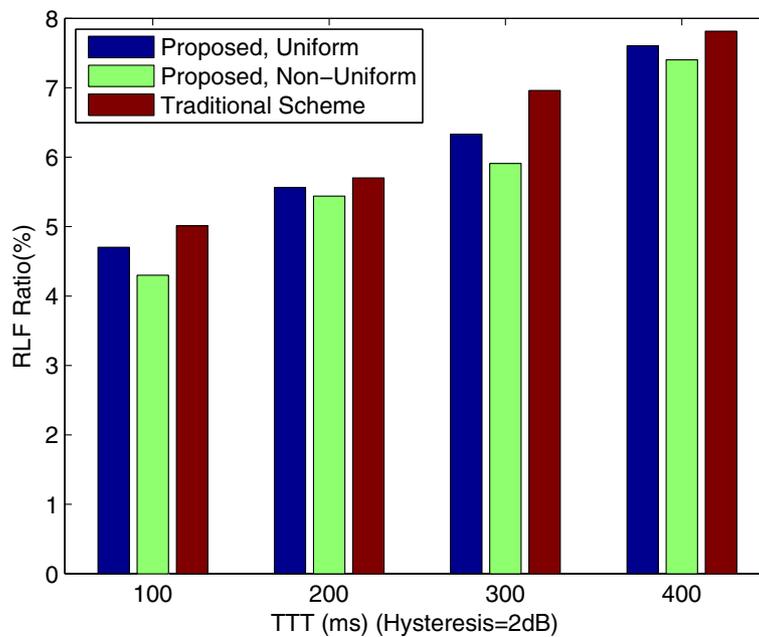


Figure 6 RLF ratio versus TTT (Hysteresis = 2 dB).

random way-point model [26] is used as UE mobility model. A UE's location is generated by a random destination and a random sample velocity with a sightline. Once reaching the destination, the UE stays for a random period of time. The macro UE (MUE) and FUE are assigned with different maximum velocity limits. The simulation parameters such as path loss, shadowing, and fading models, are given in Table 1.

For the purpose of examining the impact of weight factors, two settings of weighting factors α_i , which are named as uniform MRO and non-uniform MRO, respectively, are evaluated in the simulation. For uniform weighting, $\alpha_i = 1$, for $i = 1, \dots, 5$. For Non-Uniform weighting, $\alpha_i = 1$, for $i = 1, 2, 3$, and $\alpha_i = 3$, for $i = 4, 5$. To verify the effectiveness of the algorithms proposed, the Traditional Scheme, which uses fixed values of handover parameters, is compared with the proposed MRO algorithms. The RLF ratio is defined as the ratio of the number of RLFs caused by too late handover, too early handover, and wrong handover to the total handover attempts. Unnecessary handover ratio is defined as the proportion of ping-pong handover and continuous handover in the total handover attempts. Figures 5, 6, 7, and 8 show the comparison of the Traditional Scheme with the proposed gradient based MRO for different values of hysteresis and TTT.

As can be seen from Figure 5, the RLF rate increases as the hysteresis increases. This is because when the threshold of the handover parameter (hysteresis) increases, FUEs tend to stay in the camping femtocell, instead of handover to other femtocells, which may have better signals. Therefore, a bigger hysteresis results in less handover

attempts, but the too late handovers may still happens with Algorithm 1, resulting in increased RLF ratio. The number of too early handovers increases as the hysteresis/TTT increases, but the trend of RLF ratio does not change for the small number of too early handover in total RLFs. It can be seen from Figure 5 that the gradient based mobility robustness optimization has a better performance than traditional scheme in terms of RLF ratio. Since Non-Uniform MRO has a heavy weighting in Too Late Handovers, the proposed non-uniform scheme results in a lower RLF ratio.

In Figure 6, as the value of TTT increases, the RLF rate increases. This is because when the value of TTT increases, FUEs tend not to handover to other femtocells due to the extended observation window time. Therefore, more Too Late Handovers happen in Algorithm 1, which results in higher RLF ratio. The proposed schemes outperform the traditional scheme, and non-uniform MRO has a better performance than the Uniform one.

Figure 7 illustrates the unnecessary handover ratio of all schemes considered. As can be seen in Figure 7, unnecessary handover ratio decreases as hysteresis increases, because when the threshold of handover decision decreases, Too early handovers, ping-pong handovers, and continuous handovers are more likely to happen in Algorithm 1. Similar observations can be made in Figure 8, where unnecessary handovers occur because the observation window time is longer due to the shorter TTT. In both Figures 7 and 8, Uniform weighting factors has better performance than Non-Uniform ones, because unnecessary handover has a lower weighting factor in Non-Uniform

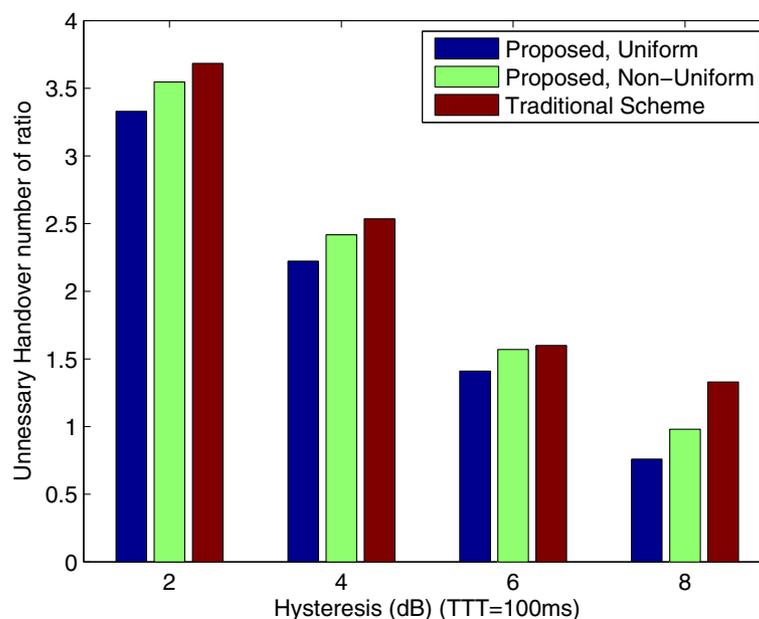


Figure 7 Unnecessary handover ratio versus hysteresis (TTT = 100 ms).

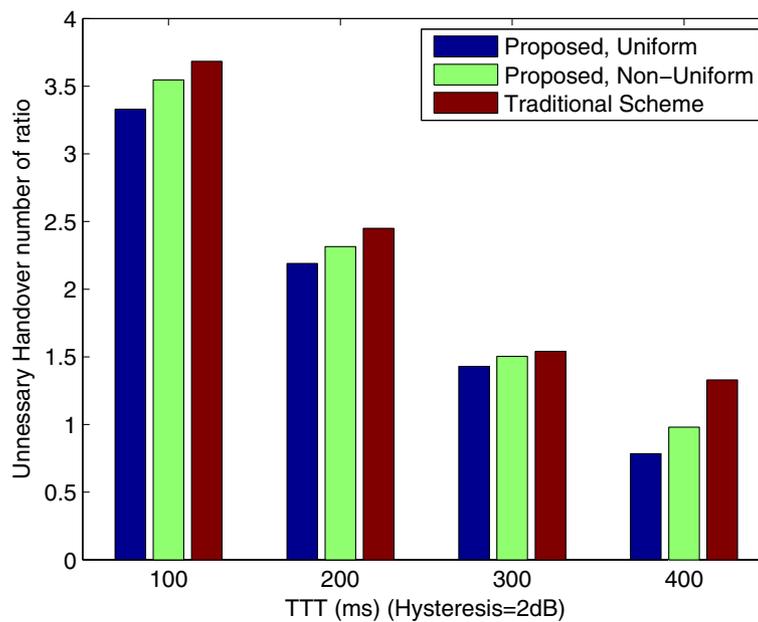


Figure 8 Unnecessary handover ratio versus TTT (Hysteresis = 2 dB).

schemes. The proposed algorithm obtains a good tradeoff between unnecessary handovers and unsuccessful handovers through adjusting weighting factors. Since RLF is more intolerant for users compared with unnecessary handover, the non-uniform weighting scheme more preferred.

5 Conclusion

In this article, a gradient based MRO scheme together with a detection algorithm of RLF and Unnecessary Handover is proposed in self-organizing LTE/LTE-A femtocell networks. After the successful detection using Algorithms 1 and 2 uses a gradient method to reduce the RLF and Unnecessary Handovers based on a cost function calculated by Algorithm 1. Simulation results have shown that the proposed scheme can achieve a reduction of RLF ratio and unnecessary handover Ratio with little handover signalling modifications.

Competing interests

The authors declare that they have no competing interests.

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Author details

¹Beijing Key Laboratory of Network System Architecture and Convergence, Beijing University of Posts and Telecommunications, Beijing 100876, China.

²Department of Electronic and Electrical Engineering, University of Sheffield, Sheffield, S1 3JD, UK.

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