

AN INTELLIGENT CALL ADMISSION CONTROL SYSTEM FOR WIRELESS CELLULAR NETWORKS BASED ON FUZZY LOGIC

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The mobile cellular systems are expected to support multiple services with guaranteed Quality of Service (QoS). However, the ability of wireless systems to accommodate expected growth of traffic load and broadband services is limited by available radio frequency spectrum. Call Admission Control (CAC) is one of the resource management functions, which regulates network access to ensure QoS provisioning. However, the decision for CAC is very challenging issue due to user mobility, limited radio spectrum, and multimedia traffic characteristics. To deal with these problems, in this paper, we propose a fuzzy CAC system. We compare the performance of the proposed system with Shadow Cluster Concept (SCC). We evaluate by simulation the performance of the proposed system. The proposed system has a good behavior on deciding the number of accepted connections while keeping the QoS for serving connections.

Keywords: Wireless networks, cellular networks, fuzzy theory, CAC.
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1 Introduction

The future telecommunications networks (such as the third-generation and forth-generation wireless networks) aim to provide integrated services such as voice, data, and multimedia via inexpensive low-powered mobile computing devices over wireless infrastructures. As the demand for multimedia services over the air has been steadily increasing over the last few years, wireless multimedia networks have been a very active research area [1]. To support various integrated services with a certain Quality of Service (QoS) requirement in these wireless networks, resource provisioning is a major issue.

The QoS support for future wireless networks is a very important problem. To guarantee the QoS, Call Admission Control (CAC) is a good strategy. CAC is a provisioning strategy that limits the number of call connections into the networks in order to reduce the network congestion and call dropping. A good CAC scheme has to balance the call blocking and call dropping in order to provide the required QoS [1].

CAC is not a problem that is unique to wireless networks. It is applicable to almost every type of networks, but in cellular wireless networks due to users' mobility the CAC becomes much more complicated. While in wired networks the resources are reserved for the call at set-up time and are not changed after that, in cellular wireless networks when the mobile

node moves from one cell to another one, the bandwidth must be requested in the new cell. During this process, the call may not be able to get a channel in the new cell to continue its service due to the limited resource in wireless networks, which will lead to the call dropping. Thus, the new and handoff calls have to be treated differently in terms of resource allocation. Since users are much more sensitive to call dropping than to call blocking, the handoff calls are assigned higher priority than new calls [2].

Future wireless networks are expected to support different broadband multimedia services with diverse QoS requirements. For more efficient utilization of the radio spectrum, the total area covered by a wireless network is divided into cells. Different traffic types can arrive at a cell. Traffic can also be handed over from other nearby cells when the mobile users cross cell boundaries. During the call setup phase, a cell must decide whether to admit a new (or a handed-over) call or not. This depends primarily on the availability of bandwidth to support the call. The amount of bandwidth allocated to a new call is based on various statistical properties of the call and on the number of calls already in session [3,4,5]. The call allowable delay and loss are specified by its QoS requirements. For example, voice traffic cannot be queued, because it cannot tolerate any delay, while data traffic is queue-able and a certain amount of delay can be acceptable.

CAC techniques are required to guarantee that all traffic types meet their QoS requirements. In order to improve the system performance at the call level (fairness in blocking), a CAC strategy may block additional calls even if there are enough resources for their service. CAC is based on the knowledge of the statistical characteristics of ongoing and arriving calls [2]. The decision to accept an additional call involves the calculation or estimation of the consequences of the call acceptance on blocking and delay of itself and other incoming calls [3].

Several schemes have recently been proposed for CAC in wireless cellular networks. One of the simplest CAC techniques is the Complete Sharing (CS) strategy. In CS, an arriving customer is served if there are enough free channels for its service. If the number of free channels is less than the channel requirements of the arriving customer, it is lost. This technique is easy to implement but it suffers from the fact that it is not fair to customers with large bandwidth requirements. The authors in [1] studied the performance of some widely known CAC protocols under more general (more accurate) assumptions and provided good approximations for the network performance. In [4], the authors proposed a model for heterogeneous multi-class environment that permits call transition between different classes. They also show that under some assumptions, the optimal policy has the shape of Multi-Priority Threshold Policy.

In [5], the authors propose a CAC algorithm that can maintain the desired level of QoS, while the successful call completion rate is very high. In the proposed algorithm, the new call arrival rate is estimated continuously, but when the estimated arrival rate is higher than a predetermined level, some new calls are blocked irrespective of the availability of channels. Several CAC schemes based on interference and power control are proposed, where the acceptance of a new request depends on Signal-to-Interference Ratio (SIR) value [2,6].

However, during the complexity of CAC in wireless environment, many simplified models and assumptions are made. Some schemes consider that each mobile node will make hand-over to neighboring cells with equal probability, which may be not accurate in general. For

this reason, in [7], the authors proposed a mobility prediction scheme that is motivated by computational learning theory. The authors derive the mobility prediction scheme from data compression techniques that are both theoretically optimal and good in practice.

Use of intelligent methods based on Fuzzy Logic (FL), Neural Networks (NN) and Genetic Algorithms (GA) can prove to be efficient for traffic control in telecommunication networks [8,9,10,11,12,13,14,15,16].

In this paper, in order to deal with CAC in wireless cellular networks, we propose a CAC system based on fuzzy logic [17]. Conventional CAC schemes for wireless networks must consider some measured parameters to make the decision. However, in wireless networks due to the user mobility and varying of channel condition the measurement obtained are not accurate. Also, it is very difficult to obtain the complete statistics of the input traffic. Therefore, the CAC decision must be made based on the uncertain or inaccurate information. The proposed fuzzy based system has 2 parts: fuzzy prediction scheme and fuzzy admission control scheme. The user movement is obtained by GPS and the fuzzy decision is based on the user speed, angle and distance from the Base Station (BS). We implement and evaluate the proposed system by simulation. We compare its performance with Shadow Cluster Concept (SCC) [18]. The proposed scheme can achieve a better prediction of the user behavior and a good admission decision compared with SCC.

The structure of this paper is as follows. In Section 2, we present the SCC. In Section 3, we introduce the proposed system. In Section 4, we discuss the simulation results. Finally, some conclusions are given in Section 5.

2 Shadow Cluster Concept

Consider a microcell wireless network system that can support mobile terminals running applications which demand a wide range of resources. Users can freely roam within the network's coverage area, and experience a large number of handoffs during a typical connection. The wireless network users expect good QoS from the system, e.g., low delays, small call dropping and packet loss probabilities.

The wireless network must provide the requested level of service even if an active mobile terminal moves to a congested cell. In this case, the corresponding BS must provide the expected service even if this implies denying network access to new connection requests. Ideally, BSs should deny network access to certain connection requests only when it is strictly necessary. This constitutes a problem that could only be optimized if knowledge were available regarding the future movement and call holding times of the active mobile terminals in the wireless network, as well as the future movements and call holding times of the mobile terminals with connection requests. As in related problems, solutions close to optimal can be obtained by using knowledge of past events to predict future behavior [18].

The fundamental idea of the SCC is that every mobile terminal with an active wireless connection exerts an influence upon the cells (and their BSs) in the vicinity of its current location and along its direction of travel. As an active mobile terminal travels to other cells, the region of influence also moves, following the active mobile terminal to its new location. The BSs (and their cells) currently being influenced are said to form a shadow cluster, because the region of influence follows the movements of the active mobile terminal like a shadow, as shown in Fig. 1. The shadow (and therefore the level of influence) is strongest near the active

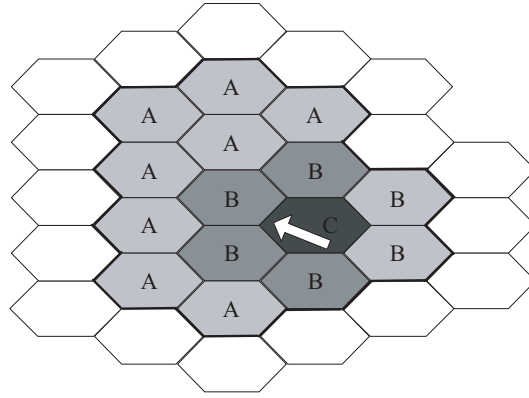


Fig. 1. SCC.

mobile terminal, and fades away depending on factors such as the distance to the mobile terminal, current call holding time and priority, bandwidth resources being used, and the mobile terminal's trajectory and velocity. Because of these factors, the shape of a shadow cluster is usually not circular and can change over time. The center of a shadow cluster is not the geometric center of the area described by the shadow, but the cell where the mobile terminal is currently located. This cell is considered as the mobile terminal's current home cell. A bordering neighbor is a cell that shares a common border with the shadow cluster's center cell. In contrast, a non-bordering neighbor cell, although being a part of the shadow cluster, does not share a border with the shadow cluster's center cell.

Conceptually, the number and "darkness" of the shadows covering a cell reflect the amount of resources that the cell's BS needs to reserve in order to support the active mobile terminals currently in its own and in neighboring cells. With the information provided by shadow clusters, BSs can determine, for each new call request, whether the request can be supported by the wireless network. In practice, a shadow cluster is a virtual message system where BSs share probabilistic information with their neighbors on the likelihood that their active mobile terminals will move to neighbor cells (while remaining active) in the near future. With the information provided by shadow clusters, BSs project future demands and reserve resources accordingly. BSs reserve resources by denying network access to new call requests, and by "waiting" for active users to end their calls.

The decision process for the acceptance of a new call request also involves a shadow cluster. Every new call request results in the implementation of a tentative shadow cluster. BSs exchange information on their new call requests, and decide, based on this and other information, which requests should be accepted and which requests should be denied.

After a handoff, BSs within the old shadow cluster are notified about this movement, and the mobile terminal's new current BS has to assume the responsibility of supplying the appropriate information to the BSs within the new shadow cluster. BSs which were in an old shadow cluster that has just moved away must delete any entries corresponding to the active mobile terminal that established the shadow cluster, and free reserved resources if appropriate. BSs which become part of the influence region of a shadow cluster must be

given appropriate information on the shadow cluster's active mobile terminal, such as the respective QoS requirements, e.g., bandwidth demands, call dropping probabilities, and any other useful information such as the wireless connection's elapsed time, for the establishment of the new shadow cluster.

3 Proposed Fuzzy Admission Control System

In order to make a more accurate decision for connection acceptance, we propose a fuzzy based CAC system.

The Fuzzy Logic Controller (FLC) is the main part of the proposed Fuzzy Admission Control System (FACS) and its basic elements are shown in Fig. 2. They are the fuzzifier, inference engine, Fuzzy Rule Base (FRB) and defuzzifier. As membership functions, we use triangular and trapezoidal membership functions because they are suitable for real-time operation [17]. They are shown in Fig. 3 and are given as:

$$f(x; x_0, a_0, a_1) = \begin{cases} \frac{x-x_0}{a_0} + 1 & \text{for } x_0 - a_0 < x \leq x_0 \\ \frac{x_0-x}{a_1} + 1 & \text{for } x_0 < x \leq x_0 + a_1 \\ 0 & \text{otherwise} \end{cases}$$

$$g(x; x_0, x_1, a_0, a_1) = \begin{cases} \frac{x-x_0}{a_0} + 1 & \text{for } x_0 - a_0 < x \leq x_0 \\ 1 & \text{for } x_0 < x \leq x_1 \\ \frac{x_1-x}{a_1} + 1 & \text{for } x_1 < x \leq x_1 + a_1 \\ 0 & \text{otherwise} \end{cases}$$

where x_0 in $f(\cdot)$ is the center of triangular function; $x_0(x_1)$ in $g(\cdot)$ is the left (right) edge of trapezoidal function; and $a_0(a_1)$ is the left (right) width of the triangular or trapezoidal function.

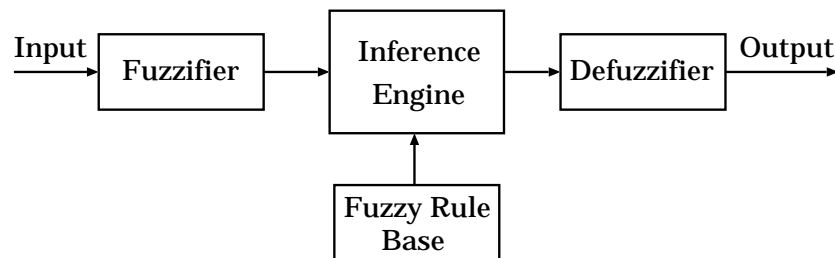


Fig. 2. FLC structure.

The proposed FACS considers the following parameters for acceptance decision: user Speed (S), user Angle (A), Distance between user and BS (D), Correction value (Cv), Required bandwidth (R), Counter state (Cs), Accept or Reject decision (A/R), Differentiated service (Ds), Real Time Counter (RTC), and Non Real Time Counter ($NRTC$). The structure of the proposed FACS is shown in Fig. 4.

3.1 FLC1 Design

The input parameters for FLC1 are: user Speed (S), user Angle (A), and the Distance between user and BS (D), while the output linguistic parameter is Correction value (Cv). The term

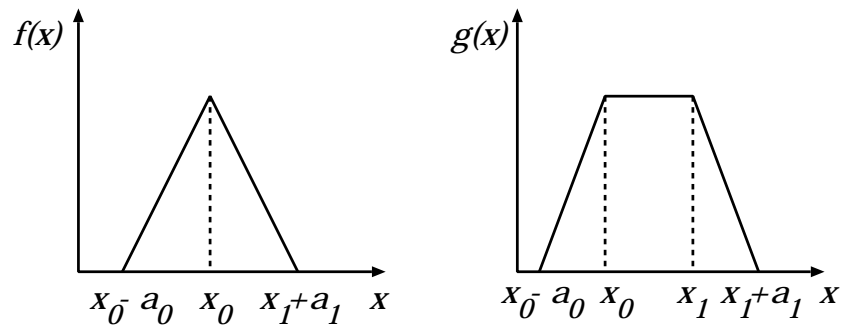


Fig. 3. Triangular and trapezoidal membership functions.

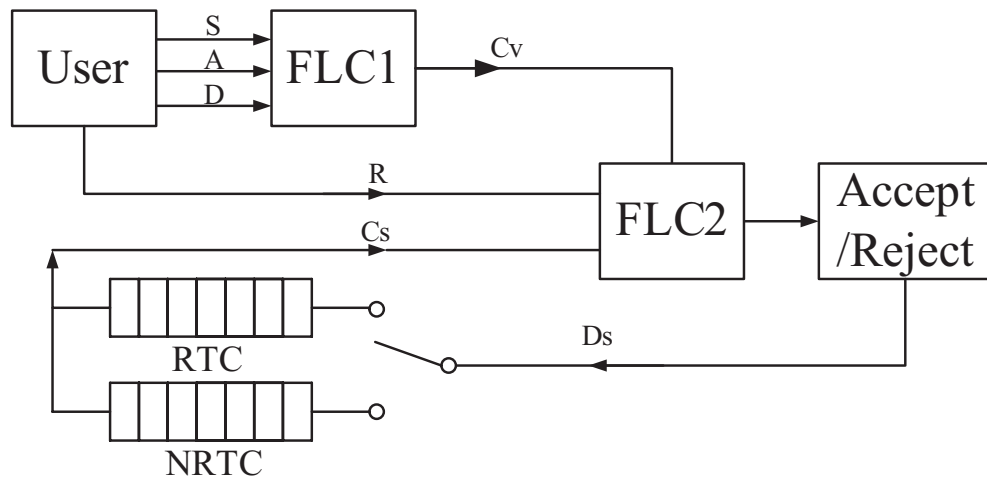


Fig. 4. Proposed system model.

sets of S , A , and D are defined respectively as:

$$\begin{aligned} T(S) &= \{Slow, Middle, Fast\} = \{Sl, M, Fa\}; \\ T(A) &= \{Back1, Left1, Left2, Straight, Right1, Right2, \\ &\quad Back2\} = \{B1, L1, L2, St, R1, R2, B2\}; \\ T(D) &= \{Near, Far\} = \{N, F\}. \end{aligned}$$

The membership functions for input parameters of FACS are defined as follows:

$$\begin{aligned} \mu_{Sl}(S) &= g(S; Sl_0, Sl_1, S_{w0}, S_{w1}); \\ \mu_M(S) &= f(S; M_0, M_{w0}, M_{w1}); \\ \mu_{Fa}(S) &= g(S; Fa_0, Fa_1, Fa_{w0}, Fa_{w1}); \\ \mu_{B1}(A) &= g(A; B1_0, B1_1, B1_{w0}, B1_{w1}); \\ \mu_{L1}(A) &= f(A; L1_0, L1_{w0}, L1_{w1}); \\ \mu_{L2}(A) &= f(A; L2_0, L2_{w0}, L2_{w1}); \\ \mu_{St}(A) &= f(A; St_0, St_{w0}, St_{w1}); \\ \mu_{R1}(A) &= f(A; R1_0, R1_{w0}, R1_{w1}); \\ \mu_{R2}(A) &= f(A; R2_0, R2_{w0}, R2_{w1}); \\ \mu_{B2}(A) &= g(A; B2_0, B2_1, B2_{w0}, B2_{w1}); \\ \mu_N(D) &= f(D; N_0, N_{w0}, N_{w1}); \\ \mu_F(D) &= f(D; Fa_0, Fa_{w0}, Fa_{w1}). \end{aligned}$$

The small letters $w0$ and $w1$ mean left width and right width, respectively.

The term set of the output linguistic parameter $T(Cv)$ is defined as $\{Correction\ value\ 1, Correction\ value\ 2, \dots, Correction\ value\ 9\} = \{Cv1, Cv2, \dots, Cv9\}$. The membership functions for the output parameter Cv are defined as follows:

$$\begin{aligned} \mu_{Cv1}(Cv) &= g(Cv; Cv1_0, Cv1_1, Cv1_{w0}, Cv1_{w1}); \\ \mu_{Cv2}(Cv) &= f(Cv; Cv2_0, Cv2_{w0}, Cv2_{w1}); \\ \mu_{Cv3}(Cv) &= f(Cv; Cv3_0, Cv3_{w0}, Cv3_{w1}); \\ \mu_{Cv4}(Cv) &= f(Cv; Cv4_0, Cv4_{w0}, Cv4_{w1}); \\ \mu_{Cv5}(Cv) &= f(Cv; Cv5_0, Cv5_{w0}, Cv5_{w1}); \\ \mu_{Cv6}(Cv) &= f(Cv; Cv6_0, Cv6_{w0}, Cv6_{w1}); \\ \mu_{Cv7}(Cv) &= f(Cv; Cv7_0, Cv7_{w0}, Cv7_{w1}); \\ \mu_{Cv8}(Cv) &= f(Cv; Cv8_0, Cv8_{w0}, Cv8_{w1}); \\ \mu_{Cv9}(Cv) &= g(Cv; Cv9_0, Cv9_1, Cv9_{w0}, Cv9_{w1}). \end{aligned}$$

The membership functions of FLC1 are shown in Fig. 5. The FRB forms a fuzzy set of dimensions $|T(S)| \times |T(A)| \times |T(D)|$, where $|T(x)|$ is the number of terms on $T(x)$. The FRB1

shown in Table 1 has 42 rules. The control rules have the following form: IF “conditions” THEN “control action”.

3.2 FLC2 Design

The input parameters for FLC2 are: the output parameter of the FLC1 (Cv), user Request (R), and the Counter state (Cs), which shows the capacity of the system. While, the output linguistic parameter is the Accept/Reject decision (A/R).

The term sets of Cv , R , and Cs are defined as:

$$\begin{aligned} T(Cv) &= \{Bad, Normal, Good\} = \{B, N, G\}; \\ T(R) &= \{Text, Voice, Video\} = \{T, Vo, Vi\}; \\ T(Cs) &= \{Small, Middle, Full\} = \{S, M, F\}. \end{aligned}$$

In order to have a soft admission decision, for the output linguistic parameter (A/R), we considered not only “accept” and “reject” but also “weak accept”, “weak reject”, and “not accept not reject” for the accept/reject decision. The membership functions for input and output linguistic parameters of FLC2 are shown in Fig. 6. The FRB2 shown in Table 2 has 27 rules.

The membership functions for input parameters of FLC2 are defined as follows:

$$\begin{aligned} \mu_B(Cv) &= f(Cv; B_0, B_{w0}, B_{w1}); \\ \mu_N(Cv) &= f(Cv; N_0, N_{w0}, N_{w1}); \\ \mu_G(Cv) &= f(Cv; G_0, G_{w0}, G_{w1}); \\ \mu_T(R) &= f(R; T_0, T_{w0}, T_{w1}); \\ \mu_{Vo}(R) &= f(R; Vo_0, Vo_{w0}, Vo_{w1}); \\ \mu_{Vi}(R) &= f(R; Vi_0, Vi_{w0}, Vi_{w1}); \\ \mu_S(Cs) &= f(Cs; S_0, S_{w0}, S_{w1}); \\ \mu_M(Cs) &= f(Cs; M_0, M_{w0}, M_{w1}); \\ \mu_F(Cs) &= f(Cs; F_0, F_{w0}, F_{w1}); \end{aligned}$$

The term set of the output linguistic parameter $T(A/R)$ is defined as {Reject, Weak Reject, Not Reject Not Accept, Weak Accept, Accept}. We write for short as {R, WR, NRNA, WA, A}. The membership functions for the output parameter A/R are defined as follows:

$$\begin{aligned} \mu_R(A/R) &= g(A/R; R_0, R_1, R_{w0}, R_{w1}); \\ \mu_{WR}(A/R) &= f(A/R; WR_0, WR_{w0}, WR_{w1}); \\ \mu_{NRNA}(A/R) &= f(A/R; NRNA_0, NRNA_{w0}, NRNA_{w1}); \\ \mu_{WA}(A/R) &= f(A/R; WA_0, WA_{w0}, WA_{w1}); \\ \mu_A(A/R) &= g(A/R; A_0, A_1, A_{w0}, A_{w1}). \end{aligned}$$

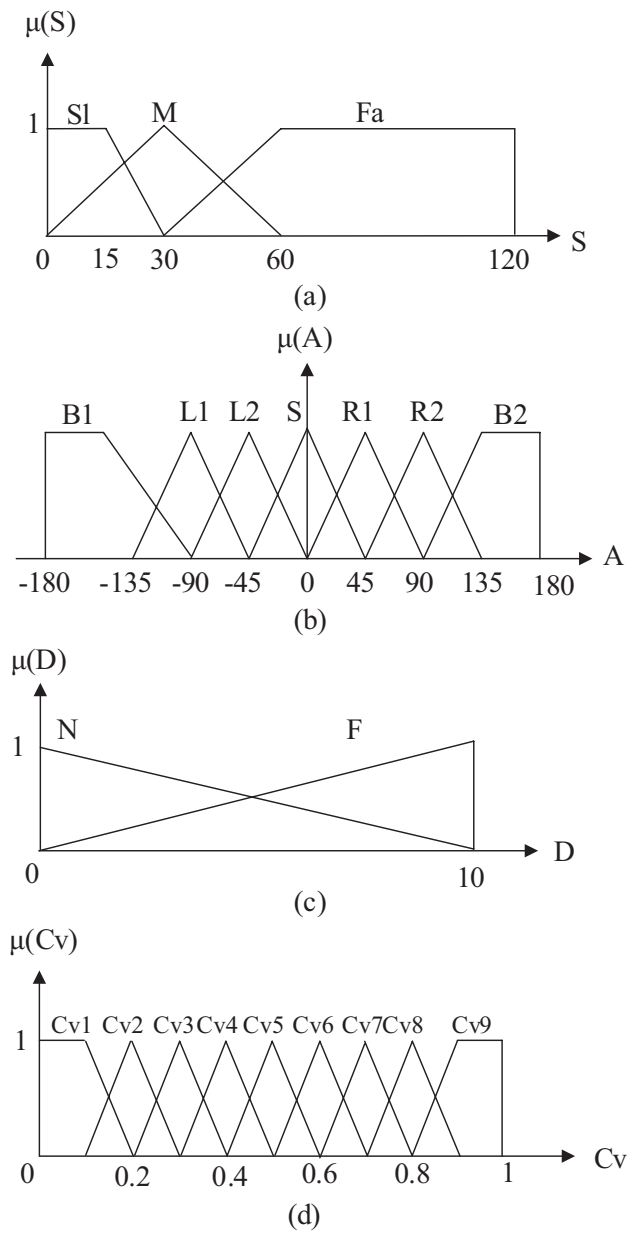


Fig. 5. FLC1 membership functions.

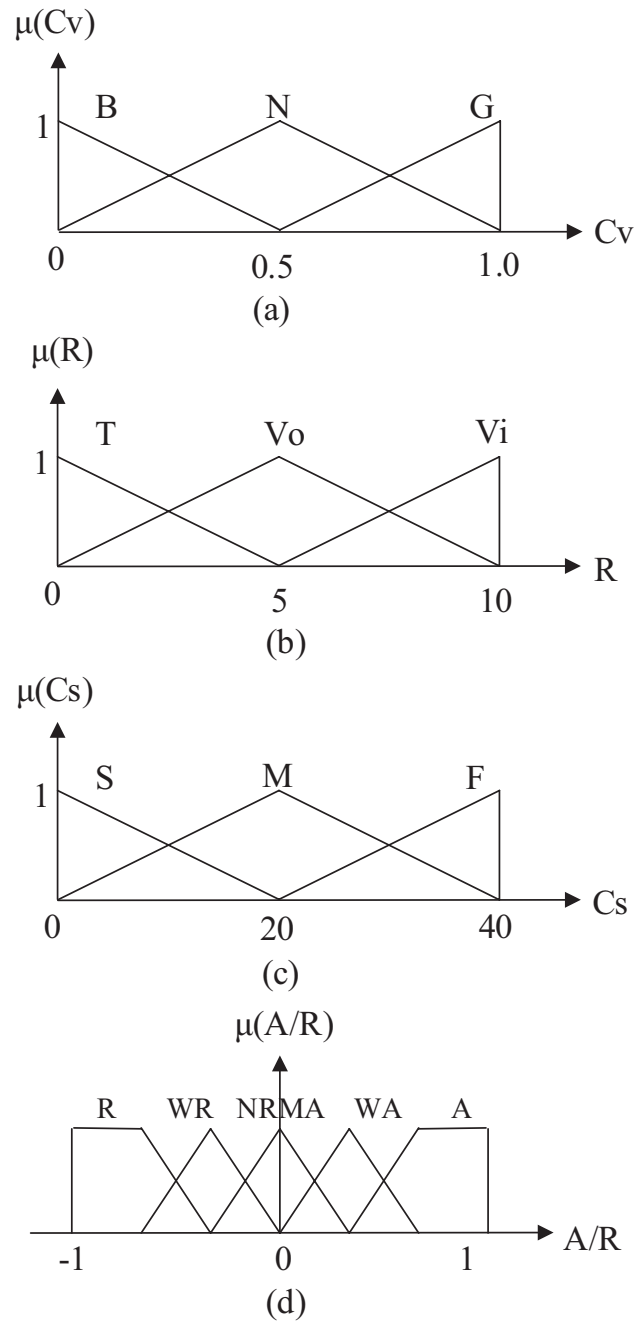


Fig. 6. FLC2 membership functions.

Table 1. FRB1.

Rule	S	A	D	Cv
0	Sl	B1	N	Cv3
1	Sl	B1	F	Cv1
2	Sl	L1	N	Cv4
3	Sl	L1	F	Cv2
4	Sl	L2	N	Cv5
5	Sl	L2	F	Cv3
6	Sl	St	N	Cv9
7	Sl	St	F	Cv3
8	Sl	R1	N	Cv5
9	Sl	R1	F	Cv2
10	Sl	R2	N	Cv4
11	Sl	R2	F	Cv2
12	Sl	B2	N	Cv3
13	Sl	B2	F	Cv1
14	M	B1	N	Cv2
15	M	B1	F	Cv1
16	M	L1	N	Cv4
17	M	L1	F	Cv1
18	M	L2	N	Cv8
19	M	L2	F	Cv5
20	M	St	N	Cv9
21	M	St	F	Cv7
22	M	R1	N	Cv8
23	M	R1	F	Cv5
24	M	R2	N	Cv4
25	M	R2	F	Cv1
26	M	B2	N	Cv2
27	M	B2	F	Cv1
28	Fa	B1	N	Cv1
29	Fa	B1	F	Cv1
30	Fa	L1	N	Cv1
31	Fa	L1	F	Cv2
32	Fa	L2	N	Cv6
33	Fa	L2	F	Cv8
34	Fa	St	N	Cv9
35	Fa	St	F	Cv9
36	Fa	R1	N	Cv6
37	Fa	R1	F	Cv8
38	Fa	R2	N	Cv1
39	Fa	R2	F	Cv2
40	Fa	B2	N	Cv1
41	Fa	B2	F	Cv1

4 Simulation Results

The simulation were carried out in Linux Fedora Core5 computer. We considered the following parameters for simulations: the user speed was from 0 to 120 km/h, the user direction was

Table 2. FRB2.

Rule	Cv	R	Cs	A/R
0	B	T	S	A
1	B	T	M	NRNA
2	B	T	F	NRNA
3	B	Vo	S	A
4	B	Vo	M	NRNA
5	B	Vo	F	WR
6	B	Vi	S	WA
7	B	Vi	M	NRNA
8	B	Vi	F	WR
9	N	T	S	A
10	N	T	M	NRNA
11	N	T	F	NRNA
12	N	Vo	S	A
13	N	Vo	M	NRNA
14	N	Vo	F	NRNA
15	N	Vi	S	WA
16	N	Vi	M	NRNA
17	N	Vi	F	NRNA
18	G	T	S	A
19	G	T	M	A
20	G	T	F	NRNA
21	G	Vo	S	A
22	G	Vo	M	A
23	G	Vo	F	WR
24	G	Vi	S	A
25	G	Vi	M	A
26	G	Vi	F	R

changed from -180 degree to +180 degree, the distance between users and BS was changed between 0 to 10 km. The required bandwidth for voice, video and text was 30%, 10%, and 60%, respectively. The requested size was 1, 5 and 10 Bandwidth Units (BU) for text, voice and video, respectively. The bandwidth of the BS was considered 40 BU.

In Fig. 7 is shown the relation between percentage of accepted calls versus number of requesting connections. In this simulation, we consider the user speed as a parameter. From the simulation results can be seen that with the increase of the user speed, the percentage of the number of the accepted calls is increased when the number of requesting connections increases. This happens because with the increase of the user speed, the user direction can not be changed easily, this results in a better prediction of the user direction and the network resources are used better. On the other hand, when the user speed is 4km/h or 10km/h, the user direction can be changed (this is the case for walking users). For this reason, the prediction of the user direction becomes difficult, which results in a small percentage of the accepted calls.

In Fig. 8, we consider the angle as a parameter. We show the simulation results for

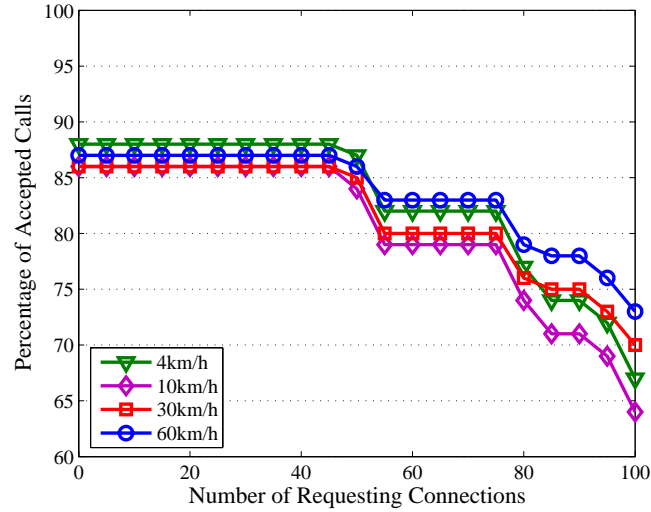


Fig. 7. Percentage of number of accepted calls vs. number of requesting connections for different speed values.

different angles from 0 to 90 degree. When the user angle is not changed, the percentage of accepted calls is close to 100% and is decreased with the increase of the number of requesting connections. With the increase of the angle, the user is going far from the BS, so there is not need to allocate the bandwidth for this user. This is why the percentage of the number of accepted calls is decreased with the increase of the angle value. When the angle is more than 90 degree, the percentage of accepted calls is almost zero. For this reason, we did not show in this figure.

In Fig. 9, the distance between the user and BS is considered as a parameter. With the increase of the number of requesting connections, the percentage of accepted calls is decreased. Also, when the distance between the user and BS is increased, the percentage of the accepted calls is decreased. However, the difference is not so big like in the case of the speed and angle. This shows that the speed and angle have strong effect compared with the distance.

In order to evaluate the performance of the proposed system, we compare its performance with SCC. The simulation results are shown in Fig. 10. When the number of requesting connections is less than 50, the percentage of accepted calls for proposed system is higher than SCC. However, when the number of requesting connections is larger than 50, the proposed system accepts less number of connections. This is because, the proposed system guarantees the QoS of ongoing calls and also considers the QoS for the requesting connections. When the number of requesting connections is less than 50 (the case when there is enough BU), the proposed system make a better allocation of the resources compared with SCC.

5 Conclusions

In this paper, we proposed a fuzzy based admission control system (called FACS) for wireless cellular networks. We evaluated the performance of the proposed system for different scenar-

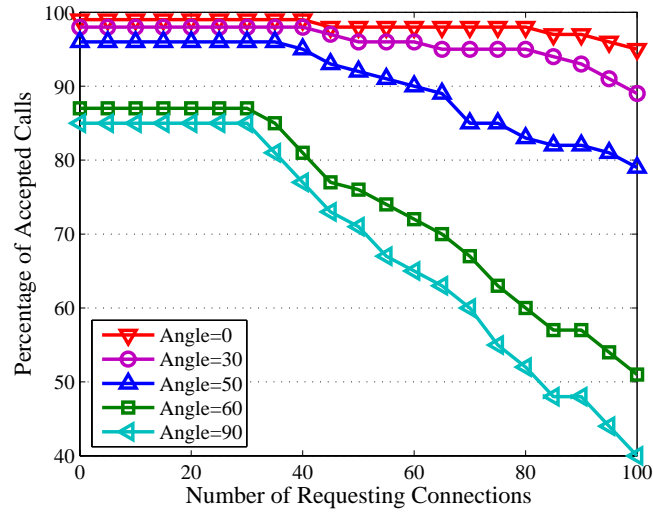


Fig. 8. Percentage of number of accepted calls vs. number of requesting connections for different angle values.

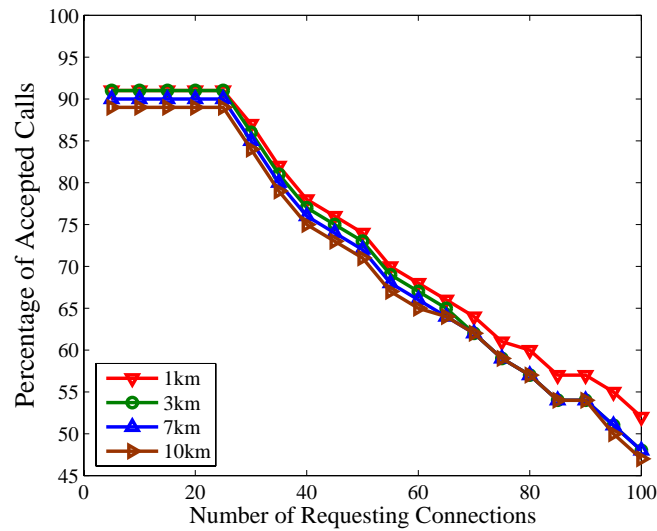


Fig. 9. Percentage of number of accepted calls vs. number of requesting connections for different distance values.

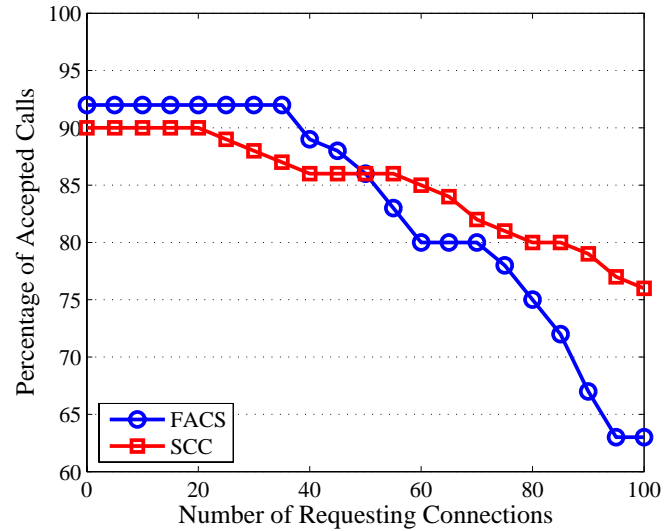


Fig. 10. Performance of proposed system and SCC.

ios. We also compared the performance of the proposed system with conventional SCC. From the simulations results, we conclude:

- with the increase of the user speed, the percentage of the number of the accepted calls is increased when the number of requesting connections increases;
- when the user speed is slow (the case for walking users), the prediction of the user direction becomes difficult, because the users can change their direction, which results in a small percentage of the accepted calls;
- when the user angle is not changed, the percentage of accepted calls is close to 100 % and is decreased with the increase of the number of requesting connections;
- when the distance between the user and BS is increased, the percentage of the accepted calls is decreased, but the difference is not so big like in the case of the speed and angle;
- the proposed system has better prediction and acceptance decision than SCC.

In this work, we did not consider the priority of the ongoing calls and requesting connections. This will be a problem to treated in the future work.

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