

Resource allocation based on fairness and QoS provisioning for OFDMA-WLAN system

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Abstract: To satisfy different service requirements of multiple users in the orthogonal frequency division multiple access wireless local area network (OFDMA-WLAN) system downlink transmission, a resource allocation algorithm based on fairness and quality of service (QoS) provisioning is proposed. Different QoS requirements are converted into different rate requirements to calculate the QoS satisfaction level. The optimization object is revised as a fairness-driven resource optimization function to provide fairness. The complex resource allocation problem is divided into channel allocation and power assignment sub-problems. The sub-problems are solved by the bipartite graph matching and water-filling based method. Compared with other algorithms, the proposed algorithm sacrifices less data rate for higher fairness and QoS satisfaction. The simulation results show that the proposed algorithm is capable of providing QoS and fairness, and performs better in a tradeoff among QoS, fairness and data rate.

Key words: QoS (quality of service) satisfaction level; fairness driven function; bipartite graph matching; water-filling; resource allocation

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The orthogonal frequency division multiple access (OFDMA) scheme has been intensively explored for offering greater flexibility in allocation of frequency resources^[1]. IEEE 802.16e and the femtocell system also use the OFDMA to exploit multi-user diversity for higher network capacity^[2]. It also has become the mainstream multiple access scheme for the downlink of the 3rd generation partner project long term evolution (3GPP LTE). Many researchers devote themselves to integrating the OFDMA technology into existing wireless communication networks by modifying the corresponding access algorithm^[3-4] to solve the problems of multipath fading or multiple access interference^[5]. A lot of documents focus

on resource allocation of the OFDMA system for the maximum network throughput^[6] and spectrum efficiency^[7] by using centralized or distributed algorithms. Some researchers combine the OFDMA technology with cognitive radio technology and study on the resource management for resource sharing between primary and secondary networks to ensure that the second user does not interfere with the primary user^[8-10]. The IEEE 802.11 work group has been trying to make standards for very high data rate wireless local area network (WLAN)^[12]. Some technical improvements have been studied to integrate multiuser dynamic OFDMA into the IEEE 802.11 WLAN^[11-12].

Since the OFDMA technology is important for multiuser system performance, many resource allocation algorithms have been studied for the OFDMA-based systems in the past few years. In the earliest studies, algorithms committed to find an efficient way to maximize the system sum rate with total power constraint^[13-15]. Then, the users' priority is considered during the resource allocation. Weighted sum-rate maximization and weighted sum-power minimization problems are proposed in Ref.[16] and solved by the Lagrange dual decomposition method. It is found that the complexity of the traditional optimization method is high; thus, the evolutionary algorithm is proposed to reduce the complexity^[17-18]. However, these references only consider the sum rate. The disadvantage is that maximizing the data rate may lead to unfair transmission and unsatisfied quality of service (QoS), although providing fairness and QoS guarantee will decrease the system data rate. In recent years, QoS and fairness provisioning have been the important research aspects for network resource optimization. Sacchi et al.^[19] proposed an OFDMA resource balance strategy based on the game theory, in which the object of optimization is the mean opinion score (MOS) but not data rate. By considering different QoS requirements of users, Ref.[20] integrated power control, relay selection and sub-carrier assign into resource allocation optimization to maximize system throughput, and supported QoS by QoS pricing. The optimal fair number of accessed real-time (RT) users and non-real-time (NRT) users is calculated in Ref.[21], but each RT user is only assigned one sub-channel and how to assign appropriate channels to users is not explained.

In the above references, different QoS requirements of different services are not considered in resource allocation.

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tion. If the resource is allocated only for bringing the highest data rate, the heterogeneous QoS requirements cannot be satisfied. When a user with a lower rate requirement is assigned a good channel with a higher data rate, the resource is wasted and it is unfair for other users with higher data requirements. In this paper, channels are assigned according to different users' QoS requests. Meanwhile, the fairness is considered in the allocation process. To prevent wasting of resources, channels should be properly allocated to users according to their demands. The optimization object is replaced by fairness-driven QoS satisfaction. The simulation results show that, the proposed resource allocation algorithm provides a better tradeoff among fairness, QoS guarantees of heterogeneous services and the system data rate.

1 System Model and Problem Formulation

As shown in Fig. 1, the OFDMA technique is integrated into the WLAN downlink transmission by the frame aggregation scheme and the link adaption scheme. The channel state information (CSI), available spectrum opportunities and the user's QoS request, which will be used in the allocation algorithm, are assumed to be available at the access point (AP) and remain unchanged during the allocation time period. AP assigns channels for downlink transmission and determines how much transmit power is allowed. The resource allocation results will be the input parameters of the frame aggregation and fragmentation module and the physical layer (PHY) processing module.

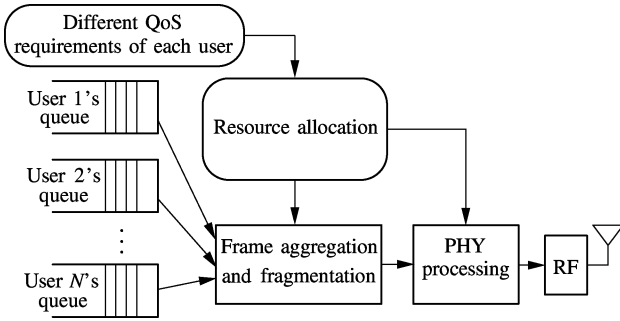


Fig. 1 Downlink transmission model at the AP

Assume that N stations (STAs) and one AP share K sub-channels. In this paper, the resource allocation optimization problem can be described as maximizing the sum satisfaction on the condition of system constraints and QoS constraints. The problem formulations are given as follows:

$$\max \sum_{i=1}^{i=N} \sum_{j=1}^{j=K} \frac{f(a_{ij})}{g(r_i^{\text{PER}}, \tau_i, r_i^{\text{data}})} \quad (1)$$

$$\text{s. t. } \sum_{i=1}^{i=N} a_{ij} = 1, a_{ij} \in \{0, 1\}, \forall j \in K \quad (2)$$

$$p_i^{\text{PER}} \leq \gamma_i^{\text{PER}} \quad (3)$$

$$t_i^D \leq \tau_i \quad (4)$$

$$\sum_{j=1}^{j=K} a_{ij} r_{ij} \geq r_i^{\text{data}} \quad (5)$$

$$\sum_{i=1}^{i=N} \sum_{j=1}^{j=K} p_{ij} \leq P \quad (6)$$

where $f(x)$ denotes the reward function with resource allocation strategy x ; $g(x)$ is the demand function with user's QoS requirement; $r_{ij} = B \log(1 + p_{ij} h_{ij} / (B n_0))$ represents user i 's data rate achieved on channel j . $a_{ij} = 1$ denotes channel j is assigned to user i ; otherwise $a_{ij} = 0$. One channel is only assigned to one user, and all channels are within available spectrum set K . The packet error rate (PER) and its threshold of user i are denoted as p_i^{PER} and γ_i^{PER} , respectively. t_i^D represents the packet delay time of user i containing waiting time and transmission time; τ_i is the time delay threshold. The data rate request of user i is represented as r_i^{data} . p_{ij} is the transmitting power assigned for user i on channel j , and the total power cannot exceed P .

2 Resource Allocation for Different QoS Requirements and Fairness

2.1 QoS-based resource optimization problem

With the constraints in section 1, the optimal result of problem (1) is difficult to be found. To decrease the complexity of optimization, QoS constraints should be handled first. Note that the data rate should at least reach a lower bound, so that the packet error rate (PER) will be below the threshold and the packet will be delivered in time.

The PER can be expressed as the increasing function of the average bit error rate (BER), and the data rate can be expressed as the decreasing function of the BER. So constraint (3) can be converted into the same form as constraint (5).

$$\sum_{j=1}^{j=K} a_{ij} r_{ij} \geq r_i^{\text{PER}} \quad (7)$$

where r_i^{PER} is the data rate due to the error bound, and the conversion is referred to Ref.[22].

For packets with the time delay threshold, the data rate at the current slot should be large enough to ensure that the most urgent packet can be delivered in time. Since the data rate can be expressed as the decreasing function of the past time after the urgent packet is created, constraint (4) can be converted into the same form as constraint (5).

$$\sum_{j=1}^{j=K} a_{ij} r_{ij} \geq r_i^{\text{delay}} \quad (8)$$

where r_i^{delay} is the lowest data rate satisfying the QoS requirement of time delay.

Now different QoS requirements can be converted into

different data rate requirements. According to problem formulations given in section 1, each user's QoS satisfaction is evaluated by the QoS satisfaction level (QSL), which is given as

$$s_i = \sum_{j=1}^{j=K} \frac{f(a_{ij})}{g(r_i^{\text{PER}}, \tau_i, r_i^{\text{data}})} = \frac{\sum_{j=1}^{j=K} a_{ij} r_{ij}}{r_i^{\text{req}}} \quad (9)$$

where $r_i^{\text{req}} \in \{r_i^{\text{PER}}, r_i^{\text{delay}}, r_i^{\text{data}}\}$.

So the QoS-based resource optimization problem can be expressed as

$$\max \sum_{i=1}^{i=N} s_i \quad (10)$$

s. t. constraint (2) and constraint (6)

2.2 Fairness-driven resource optimization problem

Problem (10) is a nonlinear programming problem. By relaxing integer constraint (2) to continuous values in range $[0, 1]$, problem (10) becomes convex and the optimal result is easy to be found by solving the Lagrangian function. However, problem (10) does not reflect the fairness allocation. Resource may be only assigned to the user with the highest data rate requirement. To achieve fairness, problem (10) can be revised by a fairness-driven utility function^[23] as

$$\max U_{\text{fair}} = \sum_{i=1}^{i=N} \log(s_i) \quad (11)$$

s. t. constraint (2) and constraint (6)

It is difficult to solve problem (11) because it involves two log functions. One of the log functions is the Shannon formula used for calculating data rate in Eq. (9). The optimal solution cannot be calculated by the Lagrangian function directly, so problem (11) is divided into several small problems to find a suboptimal solution. The channel allocation will be solved first, and then the power will be assigned based on the channel allocation result.

Given average power assignment, each user can calculate the fairness-driven utility $\log(s_i)$ on each channel. Assume that the number of users is equal to the number of channels. The channel allocation problem can be solved by bipartite graph matching. As shown in Fig. 2, let x set include all users, and y set include all the channels. Let the weight of edge is the fairness-driven utility. The object of optimal matching is the object of the maximization problem (11).

The channel allocation set can be calculated as

$$a = \operatorname{argmax} \sum_{i=1}^{i=N} \log(s_i) \quad (12)$$

s. t. constraint (2)

Given the channel allocation set, the power assignment

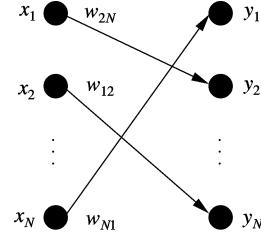


Fig. 2 Bipartite graph matching for channel allocation

can be solved by the water-filling-based method. The object of power assignment is the same as the object of problem (10). According to Ref. [22], the transmission power is assigned as

$$p_{ij} = \left(\frac{a_{ij}}{\mu r_i^{\text{req}}} - \frac{n_0}{h_{ij}} \right)^+ \quad (13)$$

where $(x)^+ = \max\{0, x\}$; μ is obtained by constraint (6). To provide fairness, μ will be revised during the iteration process by factor $1/(1 - s_i/\sum_{i=1}^{i=N} s_i)$.

2.3 Resource allocation based on fairness and QoS provisioning

According to the above analysis, the steps of the proposed algorithm for resource allocation based on fairness and QoS provisioning (RAFQ) can be given as follows:

- 1) Collect information: The AP collects CSI, available spectrum opportunities and QoS requirements of each user. Different QoS requirements will be converted into different data rate requirements. All the information should be collected at the beginning of every resource allocation circle.
- 2) Channel allocation: Given the average power assignment, AP calculates the channel allocation set by solving Eq. (12).
- 3) Power assignment: Given the channel allocation set, AP calculates the power assignment result by solving Eq. (13).
- 4) Repeat step 1) to 3) at every resource allocation circle.

3 Simulation Results

In this section, the performance of the proposed algorithm is evaluated and compared with three other algorithms which are discussed in Ref. [22]. The first algorithm is the maximum rate resource allocation algorithm (MRRA), in which the resource is always allocated to the user bringing the highest data rate. The second algorithm is the QoS provisioning channel allocation (QPCA) algorithm proposed in Ref. [21], in which the channel bringing the highest data rate is assigned to the user with the highest QSL. The third algorithm is the spectrum allocation based on the general genetic (SAGG) algorithm proposed in Ref. [17], but the fitness function is replaced

by the QSL. All the parameters used in the simulation are summarized in Tab. 1. Channels between the AP and wireless users are modeled as parallel AWGN channels with different channel gains. Each channel can only be allocated to one user.

Tab. 1 Parameters used for evaluation

Parameter	Value
Probability of crossover p_c	0.9
Probability of mutation p_m	0.02
Number of real-time users N	5
All STA users covered by one AP N_{AP}	1
Number of channels K	5
Maximum iteration number N_{iter}	100
Bandwidth of a channel B/MHz	1
$r_i^{req}/(\text{Mbit} \cdot \text{s}^{-1})$	1 to 3

A fairness index^[24] is used to evaluate the fairness performance of different algorithms. A higher value of $d(x)$ implies a higher degree of fairness.

$$d(s_i) = \frac{\left(\sum_{i=1}^N s_i\right)^2}{N \sum_{i=1}^N s_i^2} \quad (14)$$

In Fig. 3, the QSL of user 1 is much lower than that of user 3 when there is no fairness consideration. But with fairness consideration in the proposed algorithm, the QSL of user 1 is greater than that of user 3. This means that some resource of user 3 is re-allocated to user 1 to provide fair allocation. The fairness index with fairness consideration is 0.814 4, and the fairness index without fairness consideration is 0.698 4. The value increases by 16.61%, which means that the fairness of resource allocation is improved.

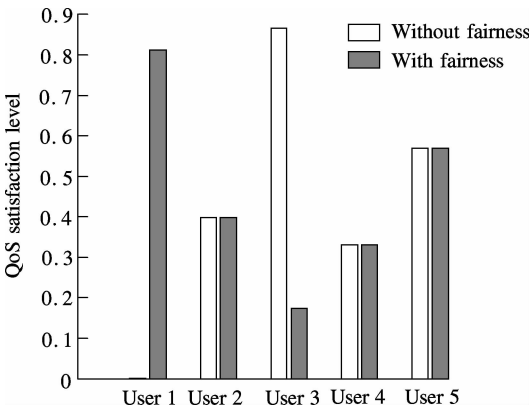


Fig. 3 The QSL of RAFQ with/without considering fairness

This result is also confirmed by Fig. 4, in which the fairness index of the proposed algorithm is higher than that of other algorithms. In three other algorithms, the object of power assignment is the sum rate; the power is assigned to improve some users' data rate while some others' requirements are ignored. Thus the fairness index of the QPCA algorithm decreases after the power allocation.

Fig. 4 shows that the fairness performance of the proposed algorithm is better than those of three other algorithms.

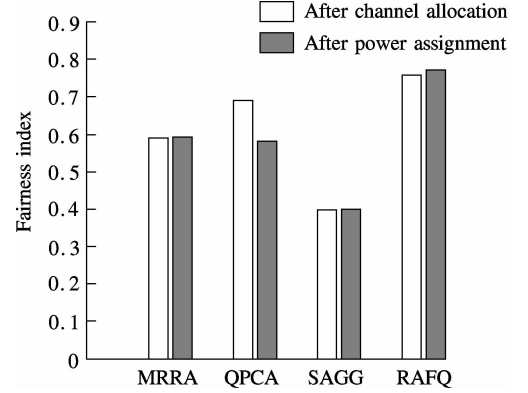


Fig. 4 Comparison of fairness index

The total QSL comparison is given in Fig. 5. The proposed algorithm has the highest total QSL value. It reveals that the RAFQ algorithm can provide different QoS guarantees and fairness. However, the RAFQ algorithm does not have the highest sum rate in Fig. 6. This is because the data rate is not the only target in the RAFQ algorithm; different QoS requirements are integrated into QoS satisfaction level; and the resource allocation process is driven by the fairness. The proposed algorithm sacrifices

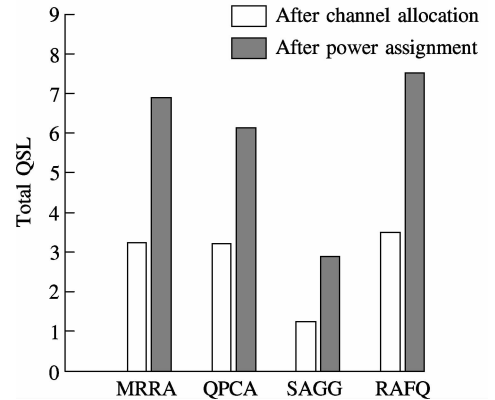


Fig. 5 Comparison of total QSL

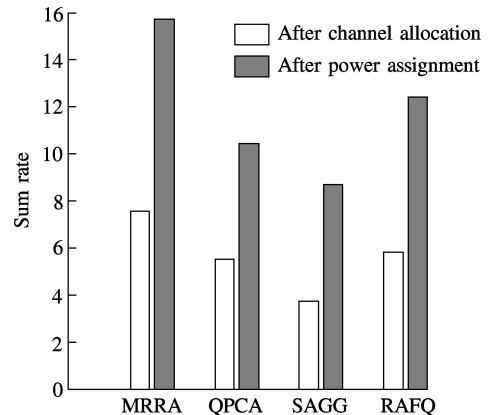


Fig. 6 Comparison of sum rate

some data rate to the QoS and fairness guarantee, but it can still obtain the second-highest data rate when it compares with other algorithms. So it is a good trade off among the data rate, QoS and fairness.

4 Conclusion

In this paper, a resource allocation algorithm is proposed for dynamic resource optimization with QoS and fairness guarantee. The system model is presumed as the OFDMA-WLAN downlink transmission system. Different QoS requirements of multiple users are converted into different data rate requirements, which are integrated into the QoS satisfaction level. The fairness-driven utility function is used to provide user fairness. The channels are allocated through bipartite graph matching. Power assignment is solved by the water-filling-based method, in which the correction factor is used to obtain fairness. The proposed RAFQ algorithm is compared with three other algorithms on total QSL, fairness index and sum rate. The simulation results show that the proposed algorithm improves fairness and QoS satisfaction with less data rate sacrifice, and performs a good tradeoff among QoS, fairness and data rate.

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基于公平性和 QoS 保障的 OFDMA-WLAN 系统资源分配

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摘要:为了满足 OFDMA-WLAN 系统下行通信中多用户的不同业务需求,提出一种基于公平性和 QoS 服务保障的资源分配算法.不同的 QoS 要求被转换成不同的速率要求来计算 QoS 满意等级;优化目标被修改为公平性驱动的优化函数以提供公平性保障;复杂的资源分配问题被划分为信道分配和功率分配问题,并通过二分图匹配和注水法得到分配结果.与其他算法相比,所提出的算法牺牲了较少的数据速率换取更高的公平性和 QoS 满意度.仿真结果表明所提算法具有保障 QoS 和公平性的能力,且在 QoS、公平性和速率之间权衡折中时表现更好.

关键词:QoS 满意等级;公平性驱动函数;二分图匹配;注水法;资源分配

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