Dynamic QoS-Aware Coalition Formation

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Abstract

Users of wireless devices increasingly demand access to multimedia content with specific quality of service requirements. Users might tolerate different levels of service, or could be satisfied with different quality combinations choices. However, multimedia processing introduces heavy resource requirements on the client side.

Our work tries to address the growing demand on resources and performance requirements, by allowing wireless nodes to cooperate with each other to meet resource allocation requests and handle stringent constraints, opportunistically taking advantage of the local ad-hoc network that is created spontaneously, as nodes move in range of each other, forming a temporary coalition for service execution. Coalition formation is necessary when a single node cannot execute a specific service, but it may also be beneficial when groups perform more efficiently when compared to a single node performance.

Keywords: Dynamic Real-time Systems, Quality of Service, Resource Management, Distributed Systems, Ad-hoc Networks

1 Introduction

Quality of Service (QoS) is considered an important user demand, receiving wide attention in real-time multimedia research [21]. However, in most systems, users do not have any real influence over the QoS they can obtain, since service characteristics are fixed when the systems are initiated. Furthermore, multimedia applications (and their users) can differ enormously in their service requirements as well as in the resources which need to be available to them [2].

New multimedia applications present increasingly complex demands on quality of service, reflected in multiple attributes over multiple quality dimensions. The use of laptop computers coupled with wireless network interfaces is growing rapidly. Recent technological development lead to the fusion of wireless ad-hoc networks, peer-to-peer computing and multimedia
content. As devices move within the range of each others a local ad-hoc network forms spontaneously, creating a new, highly dynamic and decentralized environment for multimedia applications.

Such an environment is expected to be heterogeneous, consisting of nodes with several resource capabilities. For some of those there may be a constraint on the type and size of applications they can execute with user's acceptable quality of service. For example, video conferencing systems often use compression schemes that are effective, but computationally intensive, trading CPU time for limited network bandwidth. A mobile client with limited CPU and memory capacity, but sufficient link speed, with nearby more powerful (or less congested) devices, can divide the computational intensive processing into tasks and spread it among different neighbors.

Therefore, in this paper we address distributed multimedia processing on wireless devices in scenarios where multimedia data is heavily processed when transferred between a server and the mobile device, maximizing user's quality requirements on several dimensions. We are interested in supporting spontaneous and opportunistic behavior in this new dynamic environment by enabling cooperation between available nodes. The purpose of service allocation to a group of nodes is to maximize the benefits associated to a cooperative execution of services, addressing the increasing demands on resources and performance. Nodes may cooperate either because they can not deal alone with resource allocation demands or because they can reduce the associated cost by working together.

We seek a generic model that enables a distributed service allocation, i.e., without a central authority distributing the services among nodes. Given a set of services, a distributed environment must seek the maximization of the associated QoS constraints. The nodes shall reach efficient service allocation by themselves, seeking a maximal outcome. This will be achieved via the formation of a temporary group of individual nodes, which, due to its higher flexibility and agility, is capable of effectively respond to new, challenging, requirements. We call these groups coalitions.

The basic components of such a smart environment will be nodes able to organize flexibly into a coalition for service processing. It is clear that such a group presents very significant challenges, especially at the architectural level. Major developments are required in the fields of communications protocols, data processing and application support. Our goal is to develop the architecture which enables the creation of a new generation of mobile nodes that can effectively network together, providing a flexible platform for the support of distinct network applications.

Note that this model does not preclude the existence of a fixed wired infrastructure collaborating with the wireless nodes. It is our goal to also encompass fixed set of nodes, even clusters of nodes (as in [6]), thus providing generic architecture support.

Nevertheless, currently we are focused in the ad-hoc infrastructure com-
ponents. The rest of this paper is structured as follows. The next section provides a brief description of the considered model for the system. Afterwards, section 3 presents our approach for QoS requirements representation and service requests. Section 4 presents the service allocation and coalition formation processes, while sections 5 and 6 present, respectively, the proposals formulation and evaluation. Finally, section 7 presents some conclusions.

2 Quality of Service in Ad-hoc Networks

Before building a model for this collaborative distributed processing, one must address the question if the integration is feasible and worth the effort.

Mobile devices are gaining popularity, and it is expected that the number of mobile devices will grow even more in the next few years. Very different types of mobile devices are currently available: telephones, PDAs, laptops, etc. Even the more simple ones are expected to have sooner or later sufficient processing resources to be used as multimedia clients.

Ad-hoc networks, i.e., networks without any fixed network infrastructure (such as base stations, etc.) are gaining much interest in research as well as in industry. With ad-hoc network mechanisms, clients that are in sufficiently close proximity are able to communicate directly without the need of further, externally provided, infrastructure.

At first glance, an individual mobile device may not have sufficient capacity and computation power for an effective integration in a distributed multimedia processing environment. However, if we exploit the aggregated mobile power instead of single, individual power and consider the exponential rise of mobile devices and the continuous developments in wireless technology, then one may conclude that this collaborative processing can be a valid solution.

The challenges introduced by these new requirements are already being addressed. Resource allocation in heterogeneous wireless networks has been previously addressed in [7]. The basic concept is that each service is delivered via the network that is most efficient to support the service. In [9] the authors introduced a fault tolerance architecture to provide continuous QoS support in case of network failures, by allowing users to access one of several wireless networks. Also, computation offloading on mobile computers has been previously explored [11][14][16][15]. In [14] the authors use offloading computation techniques for reducing execution time and energy consumption on handheld devices, assuming both input and output are local. A task partition/allocation scheme is proposed to allow the computation to be offloaded from the handheld to a server through a wireless LAN.
2.1 QoS Requirements Representation

Over the last years there has been a considerable amount of research within the field of QoS support for distributed multimedia systems [3].

In recent years, this research has gained significant importance. There is a lot of research on end-to-end architectures for QoS support [8][5][20][10] and on link, network and transport layers [22][19]. Some of this research has been focused in low-level systems’ mechanisms. While these issues are extremely important factors for QoS control, they are not sufficient for end users. Users may require different levels of service, or could be satisfied with different quality combination choices [12].

Research on adaptive QoS control takes user’s perspective into account by providing mechanisms for an application to adapt itself to changes in the environment, but the user has limited influence over the QoS delivered by the application. In [1] the authors propose a mechanism for QoS negotiation as a way to ensure graceful degradation. They suggest that a user should be able to express, in the service request, the spectrum of acceptable QoS levels, as well as the perceived utility of receiving service at each of those levels. However, these levels are statically mapped to certain quality choice combinations.

Research in QoS guarantees in wireless networks has been extensively studied in recent years. In [23] the authors classify the several proposed QoS schemes in three categories:

- Link adaptation in the physical layer
- Channel access coordination in the MAC layer
- Admission control strategies in MAC and higher layers

Essentially, the proposed approaches focus on different network layers and are tightly interrelated. Guaranteeing QoS in wireless networks is still a very challenging problem and needs further study [13].

3 Proposed Approach

In our approach, each application has its own QoS requirements. We propose to describe those requirements through a scheme that defines dimensions, attributes and values, as well as relations that maps dimensions to attributes and attributes to values.

\[
QoS = \{Dim, Attr, Val, DA_r, AV_r, Deps\}
\]

- \(Dim\) is the set of QoS dimensions identifiers.
- \(Attr\) is the set of attributes identifiers.
• Val is the set of attribute’s values identifiers. These are represented by the following structure:
  - Val = \{Type, Domain\}
  - Type = \{integer, float, string\}
  - Domain = \{continuous, discrete\}

• DA_r : Dim_i \rightarrow Atr, \forall Dim_i \in Dim is the relationship that assigns to each dimension in Dim a set of attributes in Attr.

• AV_r : Atr_i \rightarrow Val_k, \forall Atr_i \in Atr, \exists^1 Val_k \in Val is the relationship that assigns to each attribute in Atr a specific value in Val.

• Deps : \{Dep_{ij}\}, Dep_{ij} = f(Val_{ki}, Val_{kj}), \forall Atr_i, Atr_j \in Atr is a set of relationships defining the dependencies between attributes’ values.

As an example of this requirement description, within a video streaming application as an example, we may define a set of dimensions (and their attributes) that might be associated with any particular application (the following list is not intended to be exhaustive).

\begin{align*}
Dim &= \{Video \ Quality, \ Audio \ Quality\} \\
Attr &= \{color \ depth, \ frame \ rate, \ sampling \ rate, \ sample \ bits\} \\
Val &= \{\{1, integer, discrete\}, \{3, integer, discrete\}, \ldots, \\
&\quad\{[1, \ldots, 30], integer, continuous\}, \ldots\}
\end{align*}

\begin{align*}
DA_{Video \ Quality} &= \{color \ depth, \ frame \ rate\} \\
DA_{Audio \ Quality} &= \{sampling \ rate, \ sample \ bits\}
\end{align*}

\begin{align*}
AV_{color \ depth} &= \{1, 3, 8, 16, 24\} \\
AV_{frame \ rate} &= \{[1, \ldots, 30]\} \\
AV_{sampling \ rate} &= \{8, 16, 24, 44\} \\
AV_{sample \ bits} &= \{8, 16, 24\}
\end{align*}

3.1 Service Request

It is clearly infeasible to make the user specify the utility of every quality choice, for all the QoS dimensions of a particular application. There are simply too many choices. Instead, we impose a preference order over the dimensions, its attributes and their values on user’s service request. While
we want a user to provide a semantically rich request, so that the system
tries to achieve a service the more closely related to user’s preferences, we
also want to ensure that a user is actually able to express his preferences in
his request.

Suppose that, in a remote surveillance system, video is much more im-
portant to the user than audio. Assuming that for a particular user a gray
scale, low frame rate is fine for video, his request could be as follows:

1. Video Quality
   (a) frame rate: [10,...,5], [4,...1]
   (b) color depth: 3, 1

2. Audio Quality
   (a) sampling rate: 8
   (b) sample bits: 8

The relative decreasing order of importance imposed in dimensions, at-
tributes and values expresses user’s preferences, that is, elements identified
by lower indexes are more important than elements identified by higher in-
dexes.

In the example above, video is more important than audio, and frame
rate is more important than color depth in the Video Quality dimension. In
a similar way, the audio sampling rate is more important than the sampling
size. For each of these attributes, a preference order for the accepted values
is as well expressed.

4 Coalitions

Coalition formation is a very active research field in multi-agent systems
[18][17][4]. Most of this research assumes that agents are fully rational and
consider an optimal coalition. However, most systems in the real world are
real-time, dynamic and resource bounded. A single node has no complete
information about the environment and other nodes. As such it cannot make
fully rational reasoning when planning for a coalition. Due to the dynamic
nature and time constraints of our scenario a carefully rationalized coalition
planning may be useless or less useful by the time the coalition is formed.

A coalition’s life cycle can be decomposed in three phases:

Formation: Selection of individual partners, which will compose the coali-
tion, based in its specific skills, resources, costs and availability.

Operation: Control and monitoring of partners’ execution, resolution of
conflicts and, possibly, the coalition reconfiguration due to partial fail-
ures.
Dissolution: Termination of the coalition.

Our work is currently focused in the formation phase and our proposal concerns the development of a framework that supports the automatic coalition formation process.

QoS-aware applications are usually structured in such a way that they can provide different quality levels, which have associated estimations of the needed resources. As such, can dynamically change the executing quality level. A QoS manager can negotiate with applications the level they have to provide. This negotiation can be based on a contract model, trading of quality by resources. The final goal is to use resources in an efficient way and to maximize system quality, by cooperative execution of services.

In order to cope with limited resources, an effective resource allocation algorithm is required. The increasing demand on resources and performance of multimedia applications makes it appropriate for wireless nodes to cooperate with each other to meet resource allocation requests and handle stringent bandwidth constraints.

4.1 Cooperative Service Allocation

Consider a wireless ad-hoc network with several nodes, each one with its own particular resources $R_i$. There will be several services to be executed, each one with a set (for now) of independent tasks $T$. Each service has specific QoS constraints, defined by the user, and will compete for the finite set of resources $R$.

Let $Q_i$ be the set of QoS constraints associated with task $T_i$. Each $Q_kj$ is a finite set of quality choices for the $j^{th}$ attribute of dimension $k$. This can be either a discrete or continuous set.

In the next sections we use the following definitions:

Resource: A limited hardware or software quantity supplied by a specific node. These might be CPU time, memory, I/O bus bandwidth, network bandwidth.

Resource Manager: The object that manages a particular resource. This typically would be implemented by the device driver that manages the physical resource, by the scheduler that manages the CPU, or by software that manages other resources (such as memory).

QoS Provider: A server that negotiates access to node’s resources. Rather than reserving resources directly it will contact the Resource Managers to grant specific resource amounts to the requesting task.

Preferences: User’s preferences, for each of the QoS dimensions of a particular application.
Each node may supply a distinct set of resources, with different capacities. The nodes in this heterogeneous network will cooperate to achieve a common goal.

The objective of each coalition is to fulfill the resource allocation requests from users. Every member of the coalition provides some of the requested resources according to its own operation constraints. We view the distributed resource allocation problem as a cooperative process among nodes.

Let $N = 1, \ldots, n$ denote the set of possible nodes available for the application to request resources from. Let $P$ denote the set of user’s preferences on the QoS dimensions for an application, with its independent tasks. The basic coalition problem can be described as:

Given a set of nodes $N$ and a resource allocation demand enforced by $P$ they have to satisfy, if the resource demand cannot be satisfied by a single node or when a single node handles the request inefficiently, it is necessary for the nodes in the network to cooperate to fulfill the resource demand. With cooperation between nodes, by forming coalitions among themselves, resources can be allocated by splitting application’s tasks by a subset of $N$.

### 4.2 Coalition Formation

The coalition formation process should enable the selection of individual nodes that, based on their own resources and availability, will constitute the best group to satisfy user’s QoS requirements. In such a scenario, the adopted automatic negotiation mechanism has to be powerful enough to satisfy two important requirements:

- Ability to select the most promising nodes that should belong to the coalition. This means that nodes have to negotiate over requirements described through multiple attributes, which imply that the negotiation process must be enhanced with the capability to both evaluate and formulate multi-attribute proposals.

- In coalition formation process, each one of the individual nodes will contribute with at least one of its own resources. All these contributions may be, and they usually are, mutually dependent. The negotiation process has to be able to deal with those inter-dependencies, reaching a coherent solution.

When a user requests a service, with its specific QoS preferences, on a particular node the QoS Provider starts and guides all the negotiation process. It plays the role of Negotiation Organizer. Those nodes who are willing to belong to the future coalition (may include the node that starts the negotiation) have to submit their multi-attribute proposals, for each service’s task.

Our negotiation algorithm is described here:
1. The Negotiation Organizer broadcasts the description of each service, as well as user’s preferences on each QoS dimension.


3. The Negotiation Organizer, using a multi-attribute function, evaluates all received proposals and selects the one that offers the best utility.

4. Relevant data for task execution is sent to winning node.

The coalition is formed based on the set of proposals that presents:

- Lowest evaluation value, since it is the solution that includes values closer to the preferred ones. As our objective is to maximize user’s perceived utility, each task should be executed by the node that offers the QoS level closer to user’s preferences.

- Lowest communication cost.

- Lowest number of distinct nodes in coalition. Coalition operation’s complexity increases with the number of distinct members.

5 Proposals Formulation

Requests for task execution may arrive dynamically. To guarantee the request locally, the QoS Provider executes a local QoS optimization heuristic and formulates its proposal.

All entities that participate in the coalition formation process must provide sufficient resources to try to fulfill these QoS requirements. Therefore, each individual QoS Provider must map QoS constraints to resource requirements, and then reserve resources accordingly (resource reservations are made through the Resource Managers). This mapping is inherently difficult. To address this problem we (for now) assume that applications make a reasonable accurate analysis of their resource requirements, made a priori through resource monitoring tools, followed by run-time adaptation.

In order to make a proposal the QoS Provider contacts the required Resource Managers for resource availability, using the following algorithm, inspired in the local QoS optimization heuristic of [1]. Let each task $T_i$ have an associated set of preferences, specified by the user in relative decreasing order of preference. For each $k$ QoS dimensions there are $n$ possible attributes, $Q_{kn}$.

1. Start by selecting user’s preferred values for all QoS dimensions

2. While the set of tasks is not schedulable
(a) For each task $T_i$ receiving service at level $Q_{kj} < Q_{kn}$

(b) Determine the decrease in local reward resulting from degrading attribute $j$ to $j+1$

(c) Find task $T_m$ whose decrease is minimum and degrade it to the $j+1$'s level

The local reward is calculated by:

$$r = \begin{cases} 
  n & \text{if task is being served at } Q_{k1} \text{ for all dimensions} \\
  n - \sum_{j=1}^{n} \text{penalty}_j & \text{if } Q_{kj} > Q_{k1} 
\end{cases}$$

(1)

In equation 1 penalty is a parameter that decreases the reward value. This parameter can be defined according to user’s own criteria and its value increases with the distance for user’s preferred value.

6 Proposals Evaluation

The evaluation of received proposals implies taking into consideration multiple attributes, across several QoS dimensions. Attaching utility values to different attributes solves the problem of multi-attribute evaluation. Generally, an evaluation formula is a linear combination of attributes’ values, weighted by their corresponding utility values. As such, a multi-attribute evaluation is simply converted in a single attribute evaluation, where the result of the evaluation function can be seen as this single issue.

However, it can be very difficult to the user to specify absolute numeric values to quantify all the QoS dimension’s attributes. A more natural, and realistic way, is to simply impose a preference order over the dimensions, its attributes and their values. Therefore, in this work we consider a proposal to be formulated through the relative importance $(k = 1 \ldots n)$ of a set of QoS dimensions. Furthermore, for each dimension a relative importance order of attributes is also specified $(i = 1 \ldots attr_k)$, where $k$ is the number of attributes of dimension $k$. Note that $k, i$ are not the identifiers of dimensions and attributes, but their relative position in user’s service request. Identifiers are defined in the application’s QoS requirements representation (see section 3).

All admissible proposals are evaluated according to equation 2. A proposal is admissible if it can satisfy all the QoS dimensions requested by the user. Dimensions and their attributes are evaluated by decreasing order of importance to the user, that is, dimensions and attributes identified by lower indexes are more important than dimensions and attributes identified by higher indexes.
distance = \sum_{k=1}^{n} w_k \cdot dist(Q_k) \quad (2)

where \( n \) is the number of QoS dimensions and \( 0 \leq w_k \leq 1 \) is the relative importance of QoS dimension \( k, Q_k \), to the user, and can be defined as

\[ w_k = \frac{n - k + 1}{n} \quad (3) \]

In the equation above, QoS dimensions are presented in relative decreasing order of importance to the user. This order is specified in user’s service request, encoding user’s preferences in a qualitative way.

For each \( k \) dimension’s evaluation \( dist_k \), we propose an weighted sum of the differences between user’s preferred values and the values proposed by a specific node to that dimension’s attributes.

\[ dist(Q_k) = \sum_{i=1}^{attr_k} w_i \cdot dif(Prop_{ki}, Pref_{ki}) \quad (4) \]

where \( attr_k \) is the number of attributes in dimension \( k \). In equation 4, the function \( dist(Prop_{ki}, Pref_{ki}) \) quantifies, for an attribute \( i \), the degree of acceptability of the proposed value \( Prop_{ki} \), when compared to user’s preferred value \( Pref_{ki} \) and is defined as

\[
dif(Prop_{ki}, Pref_{ki}) = \begin{cases} 
\frac{Prop_{ki} - Pref_{ki}}{\max(Q_k) - \min(Q_k)} & \text{if continuous } Q_{ki} \\
\frac{pos(Prop_{ki}) - pos(Pref_{ki})}{\text{length}(Q_k) - 1} & \text{if discrete } Q_{ki}
\end{cases}
\]

If attribute \( i \) has a continuous domain, this quantification is a normalized difference between the proposed value and the preferred one.

For discrete domains equation 5 considers the preferences attached to \( Prop_{ki} \) and \( Pref_{ki} \) by using their relative position in the application QoS requirements specification. In [12] the authors use the notion of Quality Index, defining a bijective function that maps the elements of a discrete domain into integer values. We use a similar approach, by mapping the position (index) of that attribute in the domain specification into \( Prop_{ki} \)’s and \( Pref_{ki} \)’s scoring values.

The best proposal is the one that presents the lowest evaluation, since it is the one that contains the attributes’ values more closely related to user’s preferences, in all QoS dimensions.
7 Conclusions

Nodes in a wireless ad-hoc environment may need to cooperate in order to fulfill services. Given a set of services to be satisfied, we consider situations where a service is assigned to a group of nodes. Service allocation to several nodes is necessary when the processing cannot be performed by a single node or when a single node performs them inefficiently.

For example, playing downloaded movies may require decompression. On the other hand, transmitting data to the Internet from the mobile devices may require compression. Where to perform processing is a quite complex problem. By default, the responsibility associated with data processing is on the mobile device. However, such a default action may suffer time penalty and, possibly, battery energy loss. In the examples above, processing on the server may require additional data communication. Its possible to partition the entire process into tasks and divide them among different devices with spare resources.

Various groups of nodes may have different degrees of efficiency in service execution performance due to different capabilities of their members. As such, service allocation should be done with respect to those differences.

In this paper we addressed the specification of a generic model for enabling distributed service allocation, without a central authority distributing services among nodes, forming temporary coalitions, considering QoS-aware applications.

References


