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Sharing of Energy among Service Categories in Wireless Access Networks using Shapley Value

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Abstract—We investigate in this paper the sharing of energy consumption among service categories in the access of a wireless network. We focus on the fixed part of the energy consumption of the network, which is known to be significantly larger than the load-dependent variable part, and propose its sharing among the service categories based on coalition game concept, the Shapley value. We consider five service categories, two large players: streaming and web browsing, and three smaller ones: download, voice and other minor services, and compare our proposal with two other sharing methods: uniform and proportional which follows the same traffic proportions. Our results, applied on a real dataset extracted from an operational network in Europe, show that our proposal is more fair both towards small services in that it reduces their shares in comparison to the uniform approach, and towards larger services as it reduces their shares in comparison with the proportional one. Indeed, our Shapley-based model accommodates both short term network behavior, in which the fixed energy component is independent of the traffic load, and longer term behavior, in which it varies with the load and infrastructure. Uniform sharing accounts only for the short term, and the proportional one only for the longer term.

Index Terms—Service-oriented, Energy consumption, Wireless networks.

I. INTRODUCTION

Internet traffic is growing exponentially over the years, mainly due to the democratization of Smartphones and tablets and the increase of content. According to [1], overall IP traffic will grow at a compound annual growth rate (CAGR) of 23 percent from 2014 to 2019. In order to face this situation, Internet providers upgrade their networks so as to keep up or improve the users' perceived Quality of Experience (QoE). This leads to an increase in power consumption, resulting in turn in two main challenges: economical, as operators margin is decreasing and ecological, in a context aiming at reducing greenhouse gas emissions. That is why optimizing the power consumption of network equipments is such an active area of research.

If modeling the energy consumption of network elements is important for ecological and economical reasons, assessment of energy consumption of services is not less important. In fact, knowing the energy consumed by services should help in eco-design of applications and cost sharing model design.

We focus in this work on the fixed part of energy consumption in the access of mobile networks and its sharing among different service categories. We consider several service

categories representing players of different sizes, large and small, in terms of traffic loads. We decompose the energy model into variable versus fixed components and share the former in a manner that is proportional to the traffic load of each service category. As of the latter, we propose an approach based on coalition game concept, the Shapley value [2].

Our results show that the Shapley value allows to strike a good balance between the different service categories, in that it offers small players reduced cost, as compared to a uniform sharing, and encourages hence their transport as well as the introduction of novel, small ones. It also offers big players reduced cost as well, as compared to a sharing proportional to the traffic proportion of each service category. In doing so, it acknowledges their role as major drivers for network activity and increased deployment. This fairness comes from that, unlike the uniform sharing (which takes into account only the short term) and the proportional sharing (which takes into account only the long term), our Shapley-based model accommodates the double behavior of the fixed energy consumption which is independent of the network load over short periods of time, and varies with the load and infrastructure over longer periods of time, on the order of years.

The remainder of this paper is organized as follows. In section II, we review some literature related to the assessment of energy consumption per service category. In section III, we describe our Shapley-based models for sharing the fixed component of the energy consumption among service categories. We discuss some implementation issues of the Shapley-based model and how we tackle these issues in section IV. In section V, we run numerical applications, comparing our Shapley-based proposal to uniform as well as proportional sharing of energy, on a real dataset taken from an operational European network transporting three main service categories: streaming, browsing and download, in addition to voice and other minor services. Eventually, section VI concludes the paper.

II. RELATED WORK

Authors in [3] investigated the information and communication technology (ICT) services' energy consumption and CO2 emission at life-cycle of the equipment, including negative and positive impacts: positive impact refers to potential gains due to dematerialization, such as physical transport substitution. Negative impact refers to CO2 emissions notably. The model

for assessing the end-to-end energy consumption of a service is based on the concept of consumption rate, i.e., the energy consumed per unit of service, for example kWh/hour/user. However, no distinction is made between fixed and variable energy consumptions.

In [4], an energy model was developed to estimate the energy consumption of cloud applications; This model was applied to the case of sharing photos on Facebook. The approach consists in determining the energy consumed per bit on a device, then multiplied by the traffic volume of the service. Only the load-dependent power consumption was considered in the model, the fixed power component is ignored since it is independent of services.

Authors in [5] propose a statistical model to assess the overall energy output required for a digital service, from a Datacenter to the end user, using Monte Carlo analysis. No information was given about the nature of the considered energy consumption, fixed, variable or both.

To date, the investigations in the literature related to modeling the energy consumption of services are based on inputs that are very difficult or even impossible to measure. In [3] and [4] cited above, authors base their models on the energy consumed per transmitted bit of the service by the equipments implied on the path of the service flow. This approach has several limits. Firstly, it allows modeling much more the energy consumed by an application rather than the energy consumed by a service category. Secondly, it is quite impossible to measure the energy consumed per bit of service as most of the time, network equipments serve several services simultaneously. In order to overcome this complexity, the authors mostly refer to the power consumption models of the constructors which do not reflect the reality of the field. We propose to base our model on realistic inputs, i.e., the traffic volumes or proportions of the service categories.

Depending on which network segment is considered, one of these component is preponderant over the other. For example, the energy consumption in the core of the network is largely load-dependent because routers energy consumption varies significantly with utilization, while it is largely independent of the load in the RAN because the access is typically underloaded (typically $\rho < 50\%$ so as not to exceed some operating load threshold). Fig. 1 shows the power consumption of an operating 4G base station versus its traffic load. At 10% of load, the fixed power consumption represents 91% of the base station total power consumption.

III. MODELING OF THE ENERGY CONSUMPTION SHARING

A. Description of the system

A wireless network is composed mainly of three segments, the access, the transport and the core as depicted in [6], [7], running possibly several technologies, for instance 2G/GSM, 3G/UMTS and 4G/LTE, for the access, as shown in Fig. 2.

The RAN (Radio Access Network) is the segment of the mobile network interfacing the end-users and the mobile core network. The GSM EDGE Radio Access Network (GERAN)

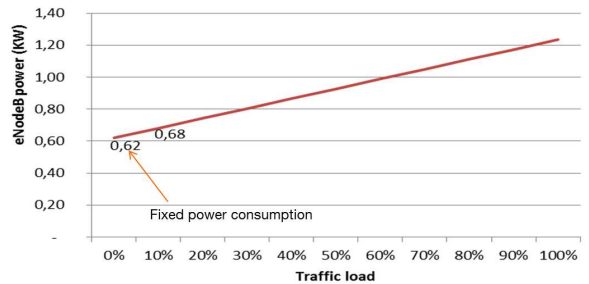


Fig. 1. Power consumption of a 4G base station.

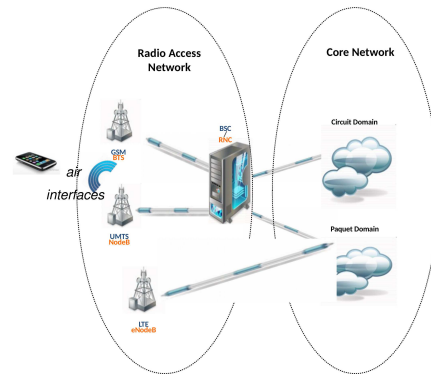


Fig. 2. The system.

is composed of the Base Transceiver Station (BTS) and the Base Station Controller (BSC).

The UMTS Terrestrial Radio Access Network (UTRAN) is composed of the NodeB and the Radio Network Controller (RNC). The RNC, as with GSM, is in charge of the management of the radio resource and implements Wideband Code Division Multiple Access (W-CDMA) as resource allocation algorithm.

The eNodeB hosts both the base station and the controller functions in a single equipment, for LTE networks. Orthogonal Frequency Division Multiplex (OFDM) is the modulation technology and OFDM Access (OFDMA) the resource allocation algorithm.

Alongside with the network elements on a site, there are the equipments of the Technical Environment (TE) composed of the cooling system, the rectifiers and the backup battery.

Traffic and energy measurements are regularly made on the network for management and investigation purposes. These measures will feed our models for the numerical applications.

B. Energy consumption model

Let us first consider a radio access network with only one radio technology (homogeneous network) transporting a set of \mathcal{N} service categories, consuming energy E to be shared among the service categories. As stated earlier, the energy consumed

by the access equipments is composed of a variable and a fixed components, denoted by E^v and E^f , respectively. Then we have

$$E = E^v + E^f \quad (1)$$

Denoting by E_i the energy consumption induced by service category i , with variable and fixed components E_i^v and E_i^f , respectively,

$$E_i = E_i^v + E_i^f \quad (2)$$

We first focus on the variable energy consumption due to service category i . Let us denote by v_i the traffic volume of service i .

$$E_i^v = \varphi_i \times E^v \quad (3)$$

where φ_i is the share of service category i in E^v , given by:

$$\varphi_i = \frac{v_i}{\sum_{k=1}^N v_k} \quad (4)$$

As of the fixed energy consumption component, it is given by:

$$E_i^f = \phi_i \times E^f \quad (5)$$

where ϕ_i is the share of service i in E^f .

Unlike the share of a service category in the variable energy consumption, that corresponds, as shown above, to its traffic proportion, we propose to determine ϕ_i using the Shapley value.

We begin by giving some introductory material on the Shapley value concept, based on [2], [8]. This mathematical tool has a number of applications in telecommunications, [9]–[11].

C. Shapley value: definition

The Shapley value is used in the theory of cooperative games to determine a solution for sharing the common revenue of a coalition. Let us consider a game $\xi(N, V)$ with N denoting the number of players and V the characteristic function, associating to each coalition of the game a value. A coalition is a set of players that cooperate so as to improve their revenue. The grand coalition is the coalition including all the players of the game. There are $N!$ possible scenarios of constructing the grand coalition.

Let S_i^σ denote the largest coalition not containing yet the player i in the construction of the grand coalition, with regard to scenario σ . We define the incremental cost vector associated to the scenario σ by:

$$c_{inc}^\sigma = (c_{inc}^\sigma(\{1\}), \dots, c_{inc}^\sigma(\{i\}), \dots, c_{inc}^\sigma(\{N\})) \quad (6)$$

where $c_{inc}^\sigma(\{i\}) = V(S_i^\sigma \cup \{i\}) - V(S_i^\sigma)$, i.e., the marginal contribution of the player i in $V(S_i^\sigma \cup \{i\})$.

The Shapley value $x_{Shapley}$ is the arithmetic mean of the incremental cost vectors associated to the scenarios of constructing the grand coalition, i.e.,

$$x_{Shapley} = \frac{1}{N!} \sum_{\sigma} c_{inc}^\sigma \quad (7)$$

D. Game without mandatory players

As stated earlier, the game is characterized by $\xi(N, V)$, with N the number of players and V the characteristic function. The characteristic function allocates to each coalition a cost, corresponding to a fraction of the fixed energy consumption of the network.

Let S denote a coalition of size s , with $s = |S|$, $|\cdot|$ the cardinal function. In the sequel, the payoffs of the players and values of the coalitions are normalized by the fixed energy consumption of the network, unless otherwise stated.

$$V(S) = \frac{\sum_{k_1=1}^s v_{k_1, S}}{\sum_{j_2=1}^{C_N^s} \sum_{k_1=1}^s v_{k_1, S_{j_2}}} \quad (8)$$

The value of the coalition S is the ratio of its traffic volume and the traffic volume of all the coalitions having the same size as S , whose number is C_N^s . $v_{k, S}$ is the traffic volume of the k^{th} element of the coalition S .

Now that the characteristic function of the game is defined, we use the Shapley value concept to compute the payoffs of the players. According to Shapley, the payoff ϕ_i of the player i is:

$$\phi_i(V, S) = \frac{1}{N!} \sum_{s=1}^N (N-s)!(s-1)! \sum_{j_1=1}^{C_{N-1}^{s-1}} \delta(\{i\}, S) \quad (9)$$

where $\delta(\{i\}, S) = V(S_{j_1, \{i\}}) - V(S_{j_1, \{i\}} \setminus \{i\})$ is the marginal contribution of the player i , in coalition S . It represents the cost gained or lost by the coalition S with the entry of the player i .

The computational complexity of (9) grows exponentially in the number of service categories, which may represent an obstacle for being implemented. We hence propose a closed-form expression for the Shapley value computation, derived from (9).

Let p_i denote the traffic proportion of the player i : $p_i = \frac{v_i}{v_t}$.

The closed-form expression of the Shapley value of the player i is :

$$\phi_i(N, p_i) = \left(\sum_{s=1}^N \frac{1}{s C_N^s} \right) p_i + \left(\sum_{s=2}^N \frac{(C_{N-1}^{s-2} - C_{N-1}^{s-1}) C_{N-2}^{s-2}}{C_{N-1}^{s-1} C_{N-1}^{s-2} s C_N^s} \right) (1 - p_i) \quad (10)$$

The derivation of this expression is found in Appendix A.

E. Game with a mandatory player

Let us now consider a game with a mandatory player. As stated above, this is the case for instance when an operator has the obligation, by the state, to offer a given service, notably voice, when deploying a network infrastructure. A mandatory player is such that there can not exist any coalition without him.

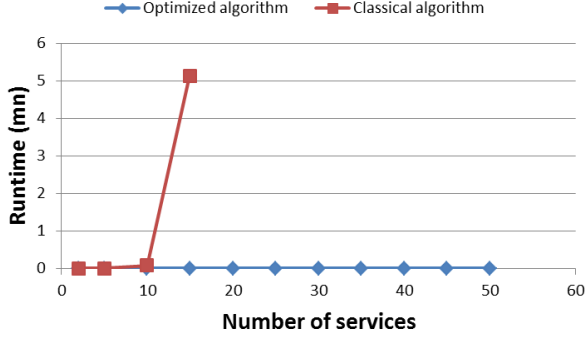


Fig. 3. Runtimes of the classical-based and closed-form-based Shapley value algorithms.

Let us denote by i^* the mandatory player and o a non mandatory player. The characteristic function of the game is defined as follows:

$$V(S) = \frac{\sum_{k_1=1}^s v_{k_1, S}}{\sum_{j_2=1}^{C_N^s} \sum_{k_1=1}^s v_{k_1, S_{j_2}}} \mathbf{1}_{i^* \in S} \quad (11)$$

ϕ_{i^*} and ϕ_o of the mandatory player i^* and a non mandatory player o , respectively, are obtained by (9).

The closed-form expression of the Shapley value of the mandatory player is :

$$\phi_{i^*}(N, p_{i^*}) = \left(\sum_{s=1}^N \frac{1}{s C_N^s} \right) p_{i^*} + \left(\sum_{s=2}^N \frac{C_{N-2}^{s-2}}{C_{N-1}^{s-1} s C_N^s} \right) (1 - p_{i^*}) \quad (12)$$

The derivation of this expression is found in Appendix B.

The closed-form expression of the Shapley value of a non mandatory player is :

$$\begin{aligned} \phi_o(N, p_{i^*}, p_o) = & \left(\sum_{s=2}^N \frac{(C_{N-1}^{s-2} - C_{N-1}^{s-1}) C_{N-2}^{s-2}}{C_{N-1}^{s-1} C_{N-1}^{s-2} s C_N^s} \right) p_{i^*} \\ & + \left(\sum_{s=2}^N \frac{C_{N-2}^{s-2}}{C_{N-1}^{s-1} s C_N^s} \right) p_o \\ & + \left(\sum_{s=3}^N \frac{(C_{N-1}^{s-2} - C_{N-1}^{s-1}) C_{N-3}^{s-3}}{C_{N-1}^{s-1} C_{N-1}^{s-2} s C_N^s} \right) (1 - p_{i^*} - p_o) \quad (13) \end{aligned}$$

The derivation of this expression is found in Appendix C.

IV. IMPLEMENTATION ISSUES

Fig. 3 shows the runtime (in minutes) of two algorithms for the computation of the Shapley values of service categories, one using (9) - denoted by Classical - and the other using the closed-form expression (10) - denoted by Optimized.

The algorithm using the closed-form expression (10) has a runtime independent of the number of service categories in the network (less than 1 second for up to 50 service categories, the maximum number of service categories one measures in the considered network), while the algorithm using (9) has a computational complexity growing exponentially in the

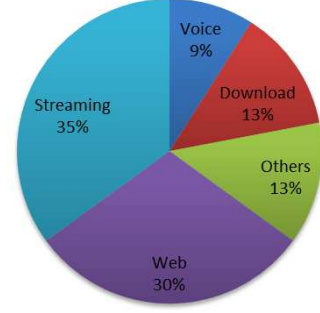


Fig. 4. Traffic proportions per service category.

number of service categories, it does not converge and has some resource limitation from a number of service categories (depending on the hardware and software environment). This comes from that (9) computes the marginal contribution of each service category in all $N!$ scenarios of constructing the grand coalition, whilst the closed-form expressions we derive from (9) are simple linear functions of the traffic proportions.

V. NUMERICAL APPLICATIONS

We now turn to the evaluation of our Shapley-based sharing model of energy between different service categories. We consider a realistic commercial network of a European country. The period of the study covers two years representing a mature 2G/3G network with early LTE deployments and associated traffic increase. We measure all voice and data services that are transmitted in the network with the following segmentation for the service categories: two large ones, namely streaming and web browsing, and three smaller ones: download, voice and other minor services. Fig. 4 shows their traffic proportions as taken from the real dataset. We consider just the traffic and energy consumption of the 3G sub-network (the network of NodeBs and RNCs).

The variable component of the energy consumption is shared proportionally to the traffic load, as this component is load-dependent. This implies that data services induced 90% of the UMTS Terrestrial Radio Access Network (UTRAN) variable energy consumption. These services are dominated by Over The Top (OTT) actors like Google.

A. Performance metric

In order to compare the three sharing strategies, i.e., Shapley-based, uniform and proportional, we introduce a satisfaction measure based on the concept of regret.

Let \mathcal{S} denote the set of strategies of the players. $\mathcal{S} = \{u, p, s\}$ with u denoting the uniform model, p the proportional model, and s the Shapley-based model. Each player has three strategies he can play. Let \mathcal{N} denote the set of players, x_i^k the fixed energy share of player i when playing strategy k . Let \tilde{x}_i denote the minimum share of player i , with regard to its strategies.

$$\tilde{x}_i = \min((x_i^k)_{k \in \mathcal{S}}) \quad (14)$$

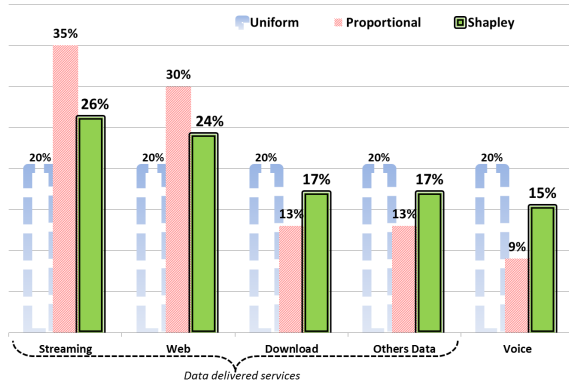


Fig. 5. Sharing of the fixed energy using three approaches: uniform, proportional and Shapley.

$\tilde{x} = (\tilde{x}_i)_{i \in \mathcal{N}}$ is the vector of minimum shares of the players. Let r_i^k denote the regret of player i when playing strategy k . It is the difference between its share when playing strategy k and its minimum share.

$$r_i^k = x_i^k - \tilde{x}_i \quad (15)$$

$r_k = (r_i^k)_{i \in \mathcal{N}}$ is the regret vector of players when strategy k is chosen. A strategy k maximizes the satisfaction of players if it minimizes the variance of the regret vector r_k with regard to other strategies. By minimizing the variance of regrets, the strategy k minimizes the difference between the regrets of satisfied and those of unsatisfied players. A player is satisfied when its regret is lower than the mean regret of the players, and is unsatisfied otherwise.

$$\text{var}(r_k) = \min((\text{var}(r_{k'}))_{k' \in \mathcal{S}}) \quad (16)$$

B. Energy sharing without a mandatory service category

We now turn to the fixed component of the energy consumption and show in Fig. 5 the sharing achieved by our Shapley-based proposal along with two other strategies: uniform sharing between the different service categories, independently of their traffic loads as on the short term the fixed energy consumption is independent of the network traffic load. And a proportional sharing which follows the traffic proportions of the service categories, given that traffic increase over a larger time scale causes network upgrades that in turn augment the fixed energy consumption.

It is worth to notice in the figure that the uniform approach favors "big services" (in terms of load) while "small ones" are favored by the proportional sharing. Our Shapley-based model achieves actually a trade-off among all the players, taking into consideration the double behavior of the fixed energy as it varies or not with the traffic load according to the time scale, unlike the uniform sharing that accounts only for the short term, and the proportional approach for the longer term. Indeed "big players", namely streaming and web services, have a lower impact in the network fixed energy consumption than they would have had with a proportional approach, as well

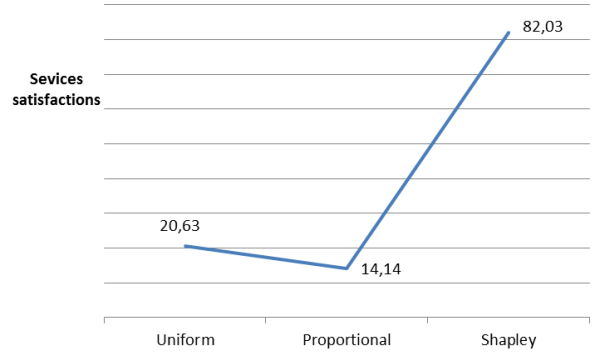


Fig. 6. Players' satisfactions per sharing approaches.

as "small players", namely voice, download and other minor services with regard to a uniform sharing.

This is a good trade-off for streaming and web services as it does not penalize them a lot and acknowledges the fact that they are major drivers for network activity, and is also a good trade-off for services with small loads as it does not make them too much responsible of the fixed energy consumption and encourages their transport as well as introduction of yet new, small ones.

The trade-off offers by our Shapley-based model to all the players results in the maximization of their satisfaction for this sharing strategy, as illustrated in Fig. 6. Maximizing the satisfaction is equivalent to reducing the difference between the highest and the lowest regrets of the players, since the players have the same mean regret whatever the strategy.

Based on our Shapley-based model, data services represent 85% of the UTRAN fixed energy consumption, versus 15% for voice service. We deduce that data services represent $0.91\theta_{3G} + 0.85(1 - \theta_{3G})$ of the total energy consumption (fixed and variable) of the 3G RAN. Typically $\theta_{3G} = 0.2$ because the access is under-loaded (say $\rho = 25\%$), finally data services represent 86% of the total RAN energy consumption. We consider the power consumption model of the base station in Fig. 1, i.e., $P(\rho) = 0.62(1 + \rho)$. For $\rho = 25\%$, $\theta_{3G} = \frac{0.775 - 0.62}{0.775} = 0.2$.

C. Energy sharing with a mandatory service category: voice

We now turn to the case where the voice service is mandatory due to legal constraints. In this scenario, voice is not considered in the selection of the best sharing strategy since it is a mandatory player, then must play whatever the sharing approach.

Based on Fig. 7, the best sharing model can not be the proportional approach as the regret of big services is very high. Uniform sharing is also eliminated because it induced higher regrets for all the players with regard to the Shapley-based model. The Shapley-based sharing appears as the strategy that minimizes the difference between the players regrets, and thus maximizing their satisfaction, as depicted in Fig. 8.

It is worth to notice that the Shapley-based model takes into account the mandatory nature of the voice service by aug-

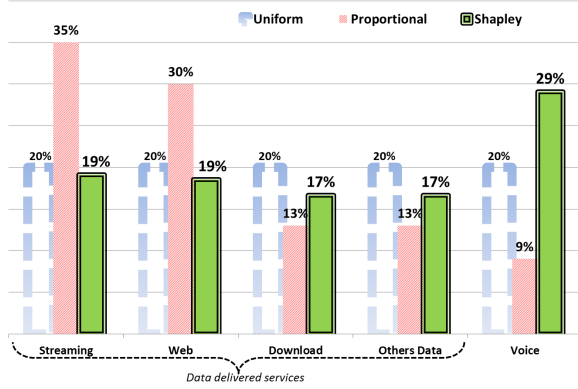


Fig. 7. Sharing of the fixed energy - voice a mandatory player.

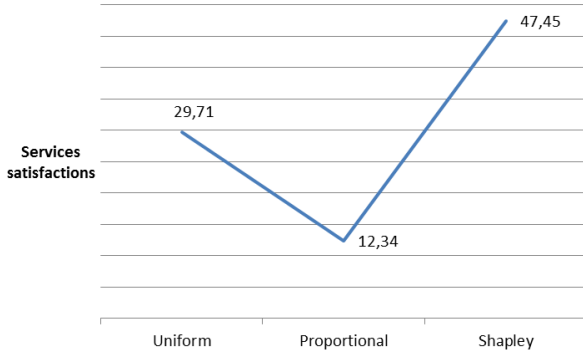


Fig. 8. Players' satisfactions per sharing approaches - voice a mandatory player.

menting significantly its share in the fixed energy consumption (from 15% to 29%). This results in a significant reduction of the impact of data services on the total energy consumption of the network. Data services represent now 57% of the total energy consumption, that corresponds to a decrease of 29% compared to the scenario where voice is not a mandatory player.

The mandatory nature of the voice service can be interpreted as if the network infrastructure was primarily deployed to deliver voice service in application of a legal obligation. That is reflected in the Shapley-based model, which is impossible with the uniform and proportional models.

VI. CONCLUSION

We investigated in this work the sharing of the energy consumed by a wireless access network among the provided service categories. We focused on the two energy components: small, load-dependent variable one, and significantly larger fixed one (since the RAN is most of the time under-loaded, with traffic load $\rho < 50\%$), load-independent over the short term but load-dependent over longer period of time (typically years). The former is to be shared among the service categories according to their traffic proportions. As of the latter, we

proposed a sharing model based on coalition game concept, the Shapley value, which allows dealing with the double behavior of that energy component as it varies or not with the network load according to the time scale. Our proposal favors small players, in terms of traffic load, when compared to uniform sharing (which takes into account only the short term) and favors larger players as compared to proportional sharing (which takes into account only the long term). This is appreciable as it gives incentives to the introduction and transport of small services, and acknowledges the role of larger service categories as major drivers for network activity.

Our next work will focus on the end-to-end path, from the content location in a datacenter for instance to the end user, and on the quantification as well as the sharing of the total energy consumption, again, among the different service categories in the overall network.

APPENDIX A. GAME WITHOUT MANDATORY PLAYERS

The characteristic function of the game without a mandatory player is as follows:

$$V(S) = \frac{\sum_{k_1=1}^s v_{k_1,S}}{\sum_{j_2=1}^{C_N^s} \sum_{k_1=1}^s v_{k_1,S_{j_2}}}$$

The value of a coalition S is the ratio of the traffic volume of that coalition and the traffic volume of all the coalitions having the same size as S , whose number is C_N^s , and size is s . $v_{k,S}$ is the traffic volume of the k^{th} element of the coalition S .

The marginal contribution of the player i is the gain or loss of the coalition S due to the entry of the player i in the coalition. It is determined as follows:

$$V(S) - V(S \setminus \{i\}) = \frac{\sum_{k_1=1}^s v_{k_1,S}}{\sum_{j_2=1}^{C_N^s} \sum_{k_1=1}^s v_{k_1,S_{j_2}}} - \frac{\sum_{k_2=1}^{s-1} v_{k_2,S \setminus \{i\}}}{\sum_{j_4=1}^{C_N^{s-1}} \sum_{k_2=1}^{s-1} v_{k_2,S_{j_4}}}$$

$$\sum_{j_1=1}^{C_{N-1}^{s-1}} V(S_{j_1}) - V(S_{j_1} \setminus \{i\}) = \frac{\sum_{j_1=1}^{C_{N-1}^{s-1}} \sum_{k_1=1}^s v_{k_1,S_{j_1},\{i\}}}{\sum_{j_2=1}^{C_N^s} \sum_{k_1=1}^s v_{k_1,S_{j_2}}} - \frac{\sum_{j_1=1}^{C_{N-1}^{s-1}} \sum_{k_2=1}^{s-1} v_{k_2,S_{j_1} \setminus \{i\}}}{\sum_{j_4=1}^{C_N^{s-1}} \sum_{k_2=1}^{s-1} v_{k_2,S_{j_4}}}$$

C_{N-1}^{s-1} is the number of coalitions of size s containing the player i , $\sum_{j_2=1}^{C_N^s} \sum_{k_1=1}^s v_{k_1,S_{j_2}}$ and $\sum_{j_4=1}^{C_N^{s-1}} \sum_{k_2=1}^{s-1} v_{k_2,S_{j_4}}$ are respectively the traffic volumes of the coalitions of size s and $s-1$. They are constant for a given coalition size s , hence we can get them out of the sum over coalitions of same size.

$\sum_{j_1=1}^{C_{N-1}^{s-1}} \sum_{k_1=1}^s v_{k_1,S_{j_1},\{i\}}$ is the sum of the traffic volumes of the coalitions of size s containing the player i . $v_{k,S_{j_1},\{i\}}$ is the traffic volume of the k^{th} element of the j^{th} coalition of size s containing the player i . The player i of course is present in all the C_{N-1}^{s-1} coalitions, while each other player is present

in C_{N-2}^{s-2} coalitions. In fact there are C_{N-2}^{s-2} coalitions of size s with both player i and a given player k .

$\sum_{j_1=1}^{C_{N-1}^{s-1}} \sum_{k_2=1}^{s-1} v_{k_2, S_{j_1} \setminus \{i\}}$ is the sum of the traffic volumes of the coalitions of size $s-1$ not containing the player i . $v_{k, S_j \setminus \{i\}}$ is the traffic volume of the k^{th} element of the j^{th} coalition of size $|S_j|$ not containing the player i . $|\cdot|$ is the cardinal function. Similarly, a given player k appears in C_{N-2}^{s-2} coalitions among the C_{N-1}^{s-1} coalitions of size $s-1$, derived from the C_{N-1}^{s-1} coalitions of size s containing i .

Then we have:

$$\sum_{j_1=1}^{C_{N-1}^{s-1}} V(S_{j_1}) - V(S_{j_1} \setminus \{i\}) = \frac{C_{N-1}^{s-1} v_i + C_{N-2}^{s-2} \sum_{k_3=1, k_3 \neq i}^N v_{k_3}}{C_{N-1}^{s-1} \sum_{k_4=1}^N v_{k_4} - \frac{C_{N-2}^{s-2} \sum_{k_3=1, k_3 \neq i}^N v_{k_3}}{C_{N-1}^{s-2} \sum_{k_4=1}^N v_{k_4}}}$$

$$\phi_i(v) = \frac{1}{N!} \sum_{s=1}^N (N-s)!(s-1)! \frac{C_{N-1}^{s-1} v_i + C_{N-2}^{s-2} \sum_{k_3=1, k_3 \neq i}^N v_{k_3}}{C_{N-1}^{s-1} \sum_{k_4=1}^N v_{k_4} - \frac{C_{N-2}^{s-2} \sum_{k_3=1, k_3 \neq i}^N v_{k_3}}{C_{N-1}^{s-2} \sum_{k_4=1}^N v_{k_4}}} \phi_{i^*}(v) = \frac{1}{N!} \sum_{s=1}^N (N-s)!(s-1)! \frac{C_{N-1}^{s-1} v_{i^*} + C_{N-2}^{s-2} \sum_{k_3=1, k_3 \neq i^*}^N v_{k_3}}{C_{N-1}^{s-1} \sum_{k_4=1}^N v_{k_4}}$$

$$C_N^s = \frac{N!}{s!(N-s)!} \implies (N-s)!(s-1)! = \frac{N!}{s C_N^s}$$

Hence,

$$\phi_i(v) = \frac{1}{v_T} \left(\sum_{s=1}^N \frac{1}{s C_N^s} \right) v_i + \frac{1}{v_T} \left(\sum_{s=2}^N \frac{(C_{N-1}^{s-2} - C_{N-1}^{s-1}) C_{N-2}^{s-2}}{C_{N-1}^{s-1} C_{N-1}^{s-2} s C_N^s} \right) (v_t - v_i)$$

Let p_i denote the traffic proportion of the player i .

$$p_i = \frac{v_i}{v_T}$$

$$\phi_i(N, p_i) = \left(\sum_{s=1}^N \frac{1}{s C_N^s} \right) p_i + \left(\sum_{s=2}^N \frac{(C_{N-1}^{s-2} - C_{N-1}^{s-1}) C_{N-2}^{s-2}}{C_{N-1}^{s-1} C_{N-1}^{s-2} s C_N^s} \right) (1 - p_i)$$

APPENDIX B. PAYOFF OF THE MANDATORY PLAYER

Let i^* denote the mandatory player. The value of a coalition with the mandatory player is :

$$V(S) = \frac{\sum_{k_1=1}^s v_{k_1, S}}{\sum_{j_2=1}^{C_N^s} \sum_{k_1=1}^s v_{k_1, S_{j_2}}} \mathbf{1}_{i^* \in S}$$

The payoff of the mandatory player is :

$$\phi_{i^*}(v) = \frac{1}{N!} \sum_{s=1}^N (N-s)!(s-1)! \sum_{j_1=1}^{C_{N-1}^{s-1}} V(S_{j_1, \{i^*\}})$$

In fact $V(S \setminus \{i^*\}) = 0 \forall S$

$$\sum_{j_1=1}^{C_{N-1}^{s-1}} V(S_{j_1, \{i\}}) = \frac{\sum_{j_1=1}^{C_{N-1}^{s-1}} \sum_{k_1=1}^s v_{k_1, S_{j_1, \{i\}}}}{\sum_{j_2=1}^{C_N^s} \sum_{k_1=1}^s v_{k_1, S_{j_2}}}$$

$$\sum_{j_1=1}^{C_{N-1}^{s-1}} V(S_{j_1, \{i\}}) = \frac{C_{N-1}^{s-1} v_i + C_{N-2}^{s-2} \sum_{k_3=1, k_3 \neq i}^N v_{k_3}}{C_{N-1}^{s-1} \sum_{k_4=1}^N v_{k_4}}$$

Hence,

$$\phi_{i^*}(v) = \frac{1}{v_T} \left(\sum_{s=1}^N \frac{1}{s C_N^s} \right) v_{i^*} + \frac{1}{v_T} \left(\sum_{s=2}^N \frac{C_{N-2}^{s-2}}{C_{N-1}^{s-1} s C_N^s} \right) (v_T - v_{i^*})$$

$$\phi_{i^*}(N, p_{i^*}) = \left(\sum_{s=1}^N \frac{1}{s C_N^s} \right) p_{i^*} + \left(\sum_{s=2}^N \frac{C_{N-2}^{s-2}}{C_{N-1}^{s-1} s C_N^s} \right) (1 - p_{i^*})$$

APPENDIX C. PAYOFF OF A NON MANDATORY PLAYER

Let o denote a non-mandatory player. The value of a coalition with a non mandatory player and the mandatory player is:

$$\phi_o(v) = \frac{1}{N!} \sum_{s=1}^N (N-s)!(s-1)! \sum_{j_3=1}^{C_{N-2}^{s-2}} V(S_{j_3, \{i^*, o\}}) - \frac{1}{N!} \sum_{s=1}^N (N-s)!(s-1)! \sum_{j_3=1}^{C_{N-2}^{s-2}} V(S_{j_3, \{i^*, o\}} \setminus \{o\})$$

C_{N-2}^{s-2} is the number of coalitions of size s containing both players i^* and o .

$$V(S) - V(S \setminus \{o\}) = \frac{\sum_{k_1=1}^s v_{k_1, S}}{\sum_{j_2=1}^{C_N^s} \sum_{k_1=1}^s v_{k_1, S_{j_2}}} - \frac{\sum_{k_2=1}^{s-1} v_{k_2, S \setminus \{o\}}}{\sum_{j_4=1}^{C_N^{s-1}} \sum_{k_2=1}^{s-1} v_{k_2, S_{j_4}}}$$

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$$\sum_{j_3=1}^{C_{N-2}^{s-2}} V(S_{j_3, \{i^*, o\}}) - V(S_{j_3, \{i^*, o\}} \setminus \{o\}) = \frac{\sum_{j_3=1}^{C_{N-2}^{s-2}} \sum_{k_1=1}^s v_{k_1, S_{j_3, \{i^*, o\}}} - \sum_{j_2=1}^{C_N^s} \sum_{k_1=1}^s v_{k_1, S_{j_2}}}{\sum_{j_3=1}^{C_{N-2}^{s-2}} \sum_{k_2=1}^{s-1} v_{k_2, S_{j_3, \{i^*, o\}} \setminus \{o\}} - \sum_{j_4=1}^{C_N^{s-1}} \sum_{k_2=1}^{s-1} v_{k_2, S_{j_4}}}$$

$$\sum_{j_3=1}^{C_{N-2}^{s-2}} V(S_{j_3, \{i^*, o\}}) - V(S_{j_3, \{i^*, o\}} \setminus \{o\}) = \frac{C_{N-2}^{s-2} v_i * + C_{N-2}^{s-2} v_o + C_{N-3}^{s-3} \sum_{\substack{k_5=1 \\ k_5 \neq i^*, o}}^N v_{k_5}}{C_{N-1}^{s-1} \sum_{k_4=1}^N v_{k_4} - \frac{C_{N-2}^{s-2} v_i * + C_{N-3}^{s-3} \sum_{\substack{k_5=1 \\ k_5 \neq i^*, j}}^N v_{k_5}}{C_{N-1}^{s-2} \sum_{k_4=1}^N v_{k_4}}}$$

In the case of a mandatory player, any non mandatory player can not form a coalition of less than 2 members.

$$\begin{aligned} \phi_o(v) &= \frac{1}{N!} \sum_{s=2}^N (N-s)!(s-1)! \frac{C_{N-2}^{s-2} v_i * + C_{N-2}^{s-2} v_o + C_{N-3}^{s-3} \sum_{\substack{k_5=1 \\ k_5 \neq i^*, o}}^N v_{k_5}}{C_{N-1}^{s-1} \sum_{k_4=1}^N v_{k_4}} \\ &- \frac{1}{N!} \sum_{s=2}^N (N-s)!(s-1)! \frac{C_{N-2}^{s-2} v_i * + C_{N-3}^{s-3} \sum_{\substack{k_5=1 \\ k_5 \neq i^*, o}}^N v_{k_5}}{C_{N-1}^{s-2} \sum_{k_4=1}^N v_{k_4}} \end{aligned} \quad [10]$$

Hence,

$$\begin{aligned} \phi_o(v) &= \frac{1}{v_T} \left(\sum_{s=2}^N \frac{(C_{N-1}^{s-2} - C_{N-1}^{s-1}) C_{N-2}^{s-2}}{C_{N-1}^{s-1} C_{N-1}^{s-2} s C_N^s} \right) v_{i^*} \\ &+ \frac{1}{v_T} \left(\sum_{s=2}^N \frac{C_{N-2}^{s-2}}{C_{N-1}^{s-1} s C_N^s} \right) v_o \\ &+ \frac{1}{v_T} \left(\sum_{s=3}^N \frac{(C_{N-1}^{s-2} - C_{N-1}^{s-1}) C_{N-3}^{s-3}}{C_{N-1}^{s-1} C_{N-1}^{s-2} s C_N^s} \right) (v_T - v_{i^*} - v_o) \end{aligned}$$

$$\begin{aligned} \phi_o(N, p_{i^*}, p_o) &= \left(\sum_{s=2}^N \frac{(C_{N-1}^{s-2} - C_{N-1}^{s-1}) C_{N-2}^{s-2}}{C_{N-1}^{s-1} C_{N-1}^{s-2} s C_N^s} \right) p_{i^*} \\ &+ \left(\sum_{s=2}^N \frac{C_{N-2}^{s-2}}{C_{N-1}^{s-1} s C_N^s} \right) p_o \\ &+ \left(\sum_{s=3}^N \frac{(C_{N-1}^{s-2} - C_{N-1}^{s-1}) C_{N-3}^{s-3}}{C_{N-1}^{s-1} C_{N-1}^{s-2} s C_N^s} \right) (1 - p_{i^*} - p_o) \end{aligned}$$