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ProceedingsPaper Energy-Saving Techniques in Next Generation of Mobile Communication Networks ⁺

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Abstract: Generalized integration of new communication technologies are expected to improve hu-8 man's life and bring the Sustainable Development Goals achievement closer in the next ten years. 9 People will benefit from rapid exchange of information, high speed of data transfer, high-quality 10 protection of uploaded and transmitted information, greater accessibility to medicine, the develop-11 ment of smart cities and homes, etc. But there are factors that significantly hinder the spread of fifth 12 generation mobile networks. And the most critical of them is a high energy consumption that comes 13 with rapid growing number of devices, machines and mobile subscribers. Most often, base stations 14always operate at full capacity, but there are statistics that show that the flow of base station clients 15 is not always the same, so the energy consumption may be managed. It was found that cell zooming 16 algorithms that mostly consider clients flow can save 14-80 % of the base station power consump-17 tion, however, not all studies pay attention to the quality of service. By balancing energy saving and 18 quality of service, base station energy savings of about 15 % can be achieved. In modern research, 19 dynamic network scaling algorithms are more often presented and the location management inves-20 tigation is identified as a solution that can help improve energy saving techniques and keep the 21 quality of service on the optimal level. 22

Keywords: Energy efficiency; Energy saving; Cell zooming; Traffic load; Green wireless communi-23cation; 5 G24

1. Introduction

Research conducted by mobile communication organizations such as Ericsson and 27 The Next Generation Mobile Networks Alliance (NGMN) demonstrate a growing trend 28 in the number of mobile application users and subscribers. Consequently, the number of 29 base stations required to cater to the growing user base is increasing, leading to the neces-30 sity of power conservation and meeting key performance indicators to ensure the quality 31 of service and the quality of the user experience. Multiple scientific investigations have 32 validated the feasibility of managing power consumption in a base station, and several 33 effective techniques have been proposed to achieve this aim. 34

This literature review aims to present an analysis of the existing research related to as energy-saving techniques in NGMN, with a focus on Traffic Driven Cell Zooming and Advanced Sleep Modes that consider the impact of traffic load. 37

The present review is structured into the following sections. At the outset, a detailed 38 summary of the prevailing trends in NGMN and their impact on the energy efficiency of 39 mobile networks. Subsequently, two key energy-saving techniques that consider the traffic load described and derived conclusions on their effectiveness based on an analysis of 41 the pertinent scientific literature. Finally, the paper concludes with a succinct summary of 42 the key findings. 43

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Copyright: © 2023 by the authors. Submitted for possible open access publication under the terms and conditions of the Creative Commons Attribution (CC BY) license (https://creativecommons.org/licenses/by/4.0/). Overall, this literature review aims to provide an understanding of the latest research in energy-saving techniques in NGMN. The identification of the principal trends and gaps in the existing literature serves as a potential roadmap for future research in this field.

2. The Present State and Future Prospects of Next Generation Mobile Networks

According to Ericsson's Mobility Report [1], Long-Term Evolution (LTE) in telecom-5 munications has become the predominant technology in Central and Eastern Europe, 6 comprising 75 % of all mobile subscriptions as of 2022. Projections indicate that LTE will 7 remain the dominant technology in that region, accounting for 56 % of mobile subscrip-8 tions by 2028, while 5G subscriptions are estimated to constitute 43 %. In contrast, the 9 implementation of 5G technology in Western Europe is occurring at a faster rate, with an 10 estimated 88 % of all mobile subscribers expected to utilize 5G, while 12 % are expected 11 to use LTE. 12

Huawei's annual report [2] predicts that by 2030, there will be 200 billion global con-13 nections, of which around 100 billion will be wireless connections, including passive cel-14 lular connections, while the other 100 billion connections will be wired, Wi-Fi, and short-15 range connections. The industrial sector will account for a large portion of these connec-16 tions, consisting of various sensors such as pressure, photoelectric, temperature, and hu-17 midity sensors, as well as intelligent cameras, autonomous vehicles, drones, and robots. 18 To support the increasing number of connected devices, industrial networks will need to 19 adopt universal broadband technologies that can replace different narrowband technolo-20 gies, providing high-speed, reliable, and low-latency connectivity to various devices and 21 applications. 22

The impact of mobile networks on global electricity use and carbon emissions is relatively small, 0.6 % and 0.2 % respectively, but the industry faces challenges in 24 managing energy use, costs, and carbon emissions due to the need for constant network 25 expansion. However, Ericsson believes that 5G can be scaled up while reducing overall 26 energy consumption. 27

According to [3], about 60 % to 80 % of the energy consumed by wireless mobile 28 networks comes from Base Stations (BSs). The NGMN Alliance reported [4] (p.15) that the 29 primary equipment consisting of the radio unit, baseband, and main control is the biggest 30 power consumer at a site, accounting for roughly 50 % of total power consumption. Air 31 conditioning is the second biggest power consumer, representing around 40 % of the total. 32 When analyzing a BSs, the radio processing component is responsible for 40 % of the en-33 ergy consumption. This component converts the digital signal from the baseband into am-34 plified radio waves. As the power amplifier in the radio unit consumes the most power, 35 it is the primary determinant of the unit's efficiency. 36

An analysis of emerging trends in NGMN suggests that improving the energy efficiency of base stations is critical due to their high energy consumption in mobile networks. 38 As base stations operate at maximum power continuously, it is essential to investigate 39 energy-saving technologies that optimize energy usage based on traffic load analysis. 40

3. Sustainable NGMN: A Survey of Energy-saving Techniques

According to the findings of NGMN Alliance [5] (p.6), there are numerous organizations that assess either the performance and user experience delivered by a particular network operator (referred to as "performance benchmarks") or the wider Environmental, Social and Governance (ESG) and sustainability standards without considering the particular service they offer. 46

Disregarding quality of service considerations in favor of enhancing the energy efficiency of a network is not an optimal strategy. The importance of Quality of Service (QoS) 48 in real-time applications cannot be overstated, as even slight deviations from the expected 49 level of performance can result in significant degradation in Quality of user's Experience 50 (QoE). 51

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Several energy-saving techniques use traffic load analysis to achieve optimal QoS 1 and improve energy efficiency in BSs. These techniques include User Association, Traffic 2 Variation, Cell Load, Cell Zooming, and On-OFF Switching Control, as shown in Figure 1. 3 Additionally, other techniques from Figure 1 can also impact the energy efficiency of the 4 base station. 5



Figure 1. Energy Efficiency techniques illustration. Note: adapted from [3].

The effectiveness of energy-saving techniques may depend on network deployment characteristics, traffic patterns, and optimization algorithms. Therefore, comprehensive evaluations are necessary before deploying them in production environments.

3.1. Traffic Driven Cell Zooming

Cell zooming is a system-level technique that can be implemented within a geographical area served by multiple base stations without requiring any hardware changes. 13 It involves adjusting the size of cells based on the network's traffic load and user needs. 14 This energy-saving technique monitors the network's traffic load and enables cell zooming when the load falls below the set traffic threshold to reduce the power consumption 16 of the base station. However, if the load exceeds the threshold, cell zooming is not enabled 17 to avoid compromising the QoS. 18

Cell zooming algorithms can be classified into two principal categories based on their 19 fundamental similarity: regular or static switch-off algorithms and dynamic switch-off al-20 gorithms. In regular algorithms, cells are switched off in a predefined pattern by the net-21 work control server during a predefined time period when low traffic arrival is expected 22 [6]. However, the limitation of this algorithm is that it cannot be used for full-day opera-23 tion, where time-dependent traffic fluctuation and traffic spatial distribution are high or 24 when traffic arrival patterns are inconsistent day by day. Despite these limitations, the 25 regular switch-off algorithm is advantageous due to its simplicity and power-saving ben-26 efits during low traffic hours. 27

Dynamic algorithms involve real-time monitoring of traffic conditions and sharing 28 that information among collaborating base stations to make decisions about cell zooming. 29 This can save power and solve traffic congestion problems if they occur. Dynamic 30

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In the work [7], a traffic driven cell zooming strategy was implemented in homoge-4 neous network, which regulated the radiation radius of individual base station to reduce 5 the energy consumption. The proposed algorithm adjusts the number of base stations 6 (BSs) in a cluster based on real-time traffic load, switching off BSs with no traffic to achieve 7 zero dynamic power consumption while ensuring service quality. When traffic increases, 8 neighboring cells handle it until a control signal triggers a switched-on BS. Transmitted 9 power is increased, extending cell radius. Compared to the Business-as-Usual scenario, 10 the algorithm is more energy-efficient, balances load, and maintains coverage and service. 11 This traffic-driven cell zooming saves energy and resolves load imbalances [7]. The power 12 consumption saving was measured in real time environment and found to be 20 % con-13 firming the previous simulated results, which showed that 14.4 % of energy saving could be obtained. Author mentions that load balancing is the key task in cell zooming.

3.2. Advanced Sleep Modes

The implementation of sleep modes represents a means of reducing power consumption, as it does not necessitate any alterations to the underlying hardware. Consequently, this technique can be readily deployed on existing architectures, facilitating its widespread adoption.

As indicated in [8], the sleep mode approaches aim to conserve energy by observing 22 the network's traffic load and determining which network components to power off or on. 23 By implementing such sleep mode techniques, wasteful energy consumption such as run-24 ning air conditioning in under-utilized BSs can be prevented. These approaches typically 25 entail toggling various components, including but not limited to power amplifiers, signal 26 processing units, cooling equipment, entire BSs, or the entire network, between sleep 27 mode and active mode. 28

On a research level, four different levels of sleep modes called Advanced Sleep 29 Modes (ASMs) can be extracted. The ASMs are exclusively determined by the time it takes 30 for the BS's components to transition, but there is no clarity on the duration of sleep for 31 individual components. In [9], it is assumed that the minimal sleep duration in each sleep 32 level is in the same time scale of the transition time corresponding to that level, and that 33 both the activation and deactivation delays are half of this transition time. These ASM 34 levels are shown on Table 1. 35

Level of sleep	Description	Sleep time (min. duration)
SM_1	Primarily deactivates the radio unit power amplifiers.	71 µs
	Deactivation of more BS components that may be retrieved from the	
	radio frequency (RF) integrated circuit or from the digital processing	
	unit.	
	Foresees the deactivation of more BS components than the both pre-	⁵ 10 ms
SM ₃	vious levels for longer duration. Similar to the 2nd level, components	
	to switch off at this level may be retrieved from the RF integrated cir-	
	cuit or from the digital processing unit.	
SM4	This level of sleep mode foresees the deactivation of more BS compo-	
	nents than all the previous levels. This level of sleep typically cannot	
	be used without loss of connectivity.	

Table 1. Sleep mode levels with minimum sleep time and depth required for their activation [9]. 36

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As was shown in work [10], the energy consumption can be reduced up to 90 % while 1 very low load. However, the delay increase may be up to 5 ms which is critical for some 2 5G use cases. For this reason, a delay-sensitive orchestration of the different ASM levels 3 is needed in order to know how much each of them is allowed to be used depending on 4 the length of the idle periods of a BS and according to the network operator's needs with 5 respect to latency [9]. 6

In work [11], a self-adaptive sleep and wake-up scheduling approach for small BSs 7 is proposed. The author suggests that the current approach to network power saving is 8 not significant due to active small cells in distant areas from the macro-cell and propose a 9 new approach that combines adaptive small cell expansion and small cell sleeping mech-10 anism to use range expanded small cells to cover the traffic system from nearby sleeping 11 small cells in the edge area of the macro-cell to save more power. This differs from existing 12 small cell sleeping mechanisms where all the traffic from a sleeping small cell is handed 13 over to the macro-cell to decrease power consumption. 14

3.3. Hybrid Energy-saving Technique

While both Cell Zooming and Sleep Modes techniques are effective in reducing energy consumption, they can also have an impact on the QoS provided by the network. For 17 example, cell zooming can lead to increased interference in neighboring cells, while sleep 18 modes can result in delays when devices need to be reactivated. To overcome these chal-19 lenges, some researchers have proposed combining cell zooming and sleep modes to 20 achieve network energy efficiency while maintaining QoS. This can be done by dynami-21 cally adjusting the size of cells based on user demand and activity levels, and putting idle devices into sleep mode when they are not needed.

In a more recent study [12], a power consumption model as a function of the traffic 24 is developed for macrocell base stations based on measurements on an actual base station. 25 This model allows to develop energy-efficient wireless access networks by author's Green 26 Radio Access Network Design (GRAND) tool which develops an always-on network with 27 a minimal power consumption for a predefined area, and an algorithm that introduces 28 sleep modes and cell zooming to reduce a power consumption. Green-field deployments 29 and optimization of existing networks are investigated. It was found that introducing 30 sleep modes and cell zooming in the network can reduce the power consumption by up 31 to 14.4 % compared to the network without sleep modes and cell zooming. Optimizing 32 existing networks by applying GRAND tool (without sleep modes and cell zooming) 33 results in a power consumption reduction of 34.5 % compared to the original network. 34 The authors conclude that for forthcoming network implementations, supporting sleep 35 modes and cell zooming techniques is advisable. In addition, future research should 36 consider incorporating adaptive capacity demands to further optimize network 37 performance. 38

The authors of [13] propose a joint cell zooming and sleeping strategy for a downlink 39 two-tier heterogeneous network modeled under the Poisson Point Process. The cell 40zooming scheme uses game theory to adjust the cell zooming factor, improving the 41 association probability of each BS and allowing more users to connect to micro BS instead 42 of macro BS, resulting in better network experiences. The two-step sleeping strategy 43 includes analyzing the energy consumption of candidate base stations to ascertain which 44 ones are kept, and measuring the overall overlapping degree of each base station to make different sleep decisions based on the radius of the area covered by MBS, resulting in more 46 energy savings. The proposed strategy balances user benefits and communication system 47 operating cost. The simulation results demonstrate an enhancement in both QoE and 48energy savings. 49

Overall, the combination of Cell Zooming and Sleep Modes techniques has the po-50 tential to significantly reduce energy consumption in wireless networks while still provid-51 ing high-quality service to users. 52

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	4. Summary	1
	This literature review discusses the current state and future prospects of Next Gen- eration Mobile Networks (NGMN) and describes energy-saving approaches for base sta- tions (BSs). The review highlights Cell Zooming and Sleep Modes as effective algorithms for reducing energy consumption by up to 20 % on real equipment, but also mentions their limitations. Based on the literature analysis, combining these techniques can help address these limitations, with load balancing being a key task in Cell Zooming. The ap- proach that combines adaptive small cell expansion and small cell sleeping mechanism is	2 3 4 5 6 7 8
	a possible method for further research.	9
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