

Chapter 14

Deep Reinforcement Learning Methods for Energy-Efficient Underwater Wireless Networking

Ahmed Ali Saihood

 <https://orcid.org/0000-0002-0147-7121>

University of Thi-Qar, Iraq

Laith Alzubaidi

University of Information Technology and Communications, Iraq

ABSTRACT

The wireless sensor networks have been developed and extended to more expanded environments, and the underwater environment needs to develop more applications in different fields, such as sea animals monitoring, predict the natural disasters, and data exchanging between underwater and ground environments. The underwater environment has almost the same infrastructure and functions with ground environment with some limitations, such as processing, communications, and battery limits. In terms of battery limits, many techniques have been proposed; in this chapter, the authors will focus in deep reinforcement learning techniques.

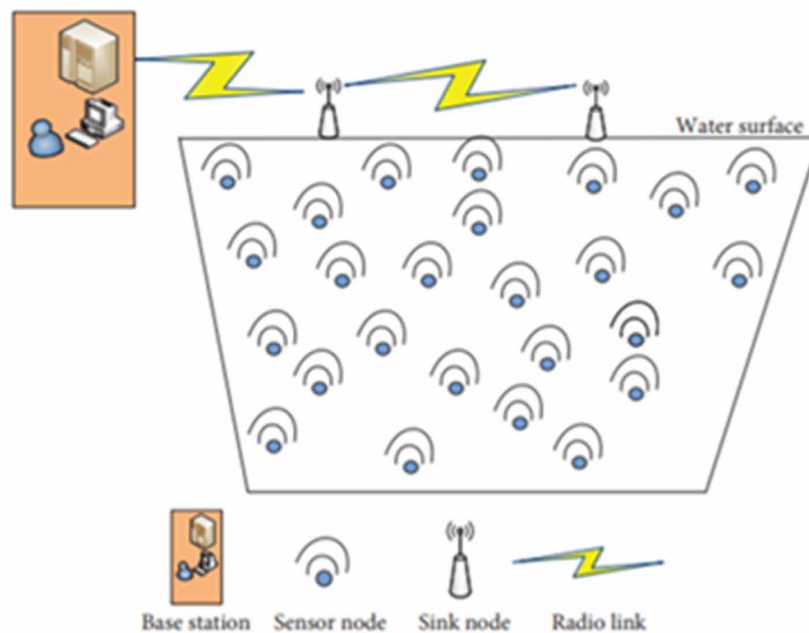
INTRODUCTION

During the recent years, Underwater Wireless Sensor Networks (UWSNs) have gained a lot of attention. These networks provide a large number of applications including, for example in studies on marine mammals as well as investigating commercial exploitations strategies. The main driving force for investigating the design process for underwater wireless sensor networks involves the need for collecting oceanography data in order to provide the opportunity for human beings to study large areas underwater in various ways (Jin, Ma, Su, Li, & Fu, 2017). The UWSN constitutes a number of sensors and vehicles

DOI: 10.4018/978-1-7998-3640-7.ch014

specifically designed for the target underwater domain in order to collect the required data and provide the opportunity for collaborative monitoring. In a general sense, in order to proceed with monitoring the ocean bed, there have been a number of oceanographic sensors developed and proposed that are capable of recording data in a stable manner, while also providing the opportunity to recover the utilized instruments once the task of collecting the required data is completed. One of the problems with using conventional methods involves the fact that there is no interactive communication between the two ends, increasing the likelihood of losing the recorded data during the transmission stage. Moreover, in such methods, the recorded data will be damaged in case of failure (Awan, et al., 2019). On the other hand, some of the major concerns and problems influencing underwater routing involve the issue of the mobility of the nodes, the limited available bandwidth, the low speed of sound, major possibilities of failure, and the limitations in energy consumption. In general, when developing routing algorithms, keeping a complete picture of the problem in mind is a highly difficult task. Therefore, there is a need for achieving a tradeoff among various limitations. Similar to any other network, in UWSNs, it is highly important and integral to set up appropriate paths and provide a secure and efficient way for transmitting the collected data (Dosaranian-Moghadam & Amo-Rahimi, 2018). The schematic architectural configuration of UWSNs is presented in Figure 1 below. There is a strong similarity between Underwater Wireless Sensor Networks (UWSNs) and Terrestrial Wireless Sensor Networks (terrestrial WSNs). However, the challenges facing the UWSNs differ from those facing the terrestrial WSNs due to difference in the technology utilized in the physical layers along with the difference in the propagation medium. In clearer terms, radio signals are used in terrestrial WSNs to transmit the data; however, such a technology is not appropriate for underwater settings since it requires a lot of energy and radio signals dissipate rapidly in

Figure 1. The architecture for the underwater sensor network (Zhu & Wei, 2018)



10 more pages are available in the full version of this document, which may be purchased using the "Add to Cart" button on the publisher's webpage:

www.igi-global.com/chapter/deep-reinforcement-learning-methods-for-energy-efficient-underwater-wireless-networking/262246

Related Content

Emerging Technologies in Transportation Systems: Challenges and Opportunities

Antonio Guerrero-Ibáñez, Carlos Flores-Cortés, Pedro Damián-Reyes and JRG Pulido (2012). *International Journal of Wireless Networks and Broadband Technologies* (pp. 12-40).

www.irma-international.org/article/emerging-technologies-in-transportation-systems/94552

WiFiMon: A Tool for Wi-Fi Performance Monitoring and Verification

Christos Bouras, Kurt Baumann, Vasileios Kokkinos, Nikolaos Papachristos and Kostas Stamos (2019). *International Journal of Wireless Networks and Broadband Technologies* (pp. 1-18).

www.irma-international.org/article/wifimon/237188

Deep Reinforcement Learning Methods for Energy-Efficient Underwater Wireless Networking

Ahmed Ali Saihood and Laith Alzubaidi (2021). *Energy-Efficient Underwater Wireless Communications and Networking* (pp. 212-223).

www.irma-international.org/chapter/deep-reinforcement-learning-methods-for-energy-efficient-underwater-wireless-networking/262246

Dynamic Cache Management of Cloud RAN and Multi-Access Edge Computing for 5G Networks

Deepika Pathinga Rajendiran, Yihang Tang and Melody Moh (2021). *Research Anthology on Developing and Optimizing 5G Networks and the Impact on Society* (pp. 276-308).

www.irma-international.org/chapter/dynamic-cache-management-of-cloud-ran-and-multi-access-edge-computing-for-5g-networks/270196

Next Generation Cellular Network Planning: Transmission Issues and Proposals

Spiros Louvros and Athanassios C. Iossifides (2005). *Mobile and Wireless Systems Beyond 3G: Managing New Business Opportunities* (pp. 156-182).

www.irma-international.org/chapter/next-generation-cellular-network-planning/26435