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# Machine Learning Modeling for IoUT Networks

Internet of Underwater Things



Springer

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ISSN 2191-5768                    ISSN 2191-5776 (electronic)  
SpringerBriefs in Computer Science  
ISBN 978-3-030-68566-9        ISBN 978-3-030-68567-6 (eBook)  
<https://doi.org/10.1007/978-3-030-68567-6>

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# Preface

Our objective of writing this book is to draw some outlines to enable the reader to find a good source of knowledge that could help them to understand applied technologies in underwater communication as well as the existing challenges and some suggested solutions.

The first aim of this book is to shed light on the leading variable physical properties of water that affect the transmission of the core carrier for underwater communications, i.e., acoustic waves, and what leads to changes in its characteristics, e.g., intensity, resulting in inaccurate production of the deployed technologies that build on the assumption of fixed speed transmission of sound in underwater environments.

The second major aim is to investigate the application of machine learning (ML) techniques in settling diverse challenges that are encountered during deployment of underwater technologies, such as the Internet of Underwater Things (IoUT) and multi-modal underwater networks, and to capitalize on their merits. In addition, ML has the capabilities to treat the traditional underwater model-driven problems by considering the enormous measured data into appropriate data-driven problems and handle them to design a proper and adaptive behavioral modeling of these problems. This overcomes the main underwater problem where there is still no generic model that exists for the underwater environments because of the extremely harsh and fluctuating nature of such ambiance over the spatio-temporal domains since the ML techniques, unlike the theoretical systems, do not rely on explicit or certain propagation models or assumptions.

The book provides IoUT network and node structure and the ML modeling for underwater communication in Chap. 1, the key physical variables of water and their interrelationships in Chap. 2, an example of channel modeling and an adaptive transmission framework for underwater networks in Chap. 3, two examples of the positioning systems in Chap. 4, and application of the decision tree as a classifier and dynamic modeling using neural networks for underwater techniques. The book also

proposes in Chap. 6 some challenges faced by underwater communication and some glimpses of the solutions from both communication and data science perspectives.

Finally, we hope this book satisfies the need of the community members and they find it interesting and fruitful.

Nanshan, China  
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Ahmad A. Aziz El-Banna  
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# Contents

|          |   |           |
|----------|---|-----------|
| <b>1</b> | <b>Introduction to Underwater Communication and IoUT Networks .....</b> | <b>1</b>  |
| 1.1      | Underwater Communication.....   | 1         |
| 1.2      | IoUT Network and Node Structure.....                                    | 2         |
| 1.2.1    | Network Architecture.....   | 3         |
| 1.2.2    | Sensor Node Architecture .....  | 4         |
| 1.3      | Machine Learning Modeling for Underwater<br>Communication .....         | 5         |
|          | References .....  | 7         |
| <b>2</b> | <b>Seawater's Key Physical Variables .....</b>                          | <b>9</b>  |
| 2.1      | Key Physical Variables (KPVs) .....                                     | 9         |
| 2.1.1    | Temperature .....   | 9         |
| 2.1.2    | Salinity .....  | 10        |
| 2.1.3    | Pressure .....  | 11        |
| 2.1.4    | Density .....   | 12        |
| 2.1.5    | pH .....  | 12        |
| 2.1.6    | Internal Waves .....  | 12        |
| 2.1.7    | Conductivity .....  | 14        |
| 2.1.8    | KPV Interrelationships .....  | 14        |
|          | References .....  | 15        |
| <b>3</b> | <b>Opportunistic Transmission in IoUT Networks .....</b>                | <b>17</b> |
| 3.1      | Underwater Communication.....   | 17        |
| 3.1.1    | Channel Model.....  | 17        |
| 3.1.2    | Transmission System Analysis .....                                      | 19        |
| 3.2      | Confidence Metric .....   | 20        |
| 3.2.1    | Derivations of the Confidence Metric .....                              | 20        |
| 3.2.2    | TPL Control and Adaptive Modulation .....                               | 23        |
| 3.2.3    | Employing the Confidence Metric in the<br>Transmission Framework.....   | 24        |
| 3.3      | Performance Analysis .....  | 26        |
|          | References .....  | 33        |

|  |    |
|--|----|
| <b>4 Localization and Positioning for Underwater Networks .....</b>      | 37 |
| 4.1 Introduction .....   | 37 |
| 4.2 System Modeling Using TDoA .....                                     | 37 |
| 4.3 TDoA-Based Positioning Approach .....                                | 39 |
| 4.4 System Modeling Using RSS Method .....                               | 40 |
| 4.5 RSS Positioning Approach .....                                       | 42 |
| References .....   | 43 |
| <b>5 ML: Modeling for Underwater Communication in IoUT Systems .....</b> | 45 |
| 5.1 Classification for Transmission Methods .....                        | 45 |
| 5.2 Dynamic Modeling Using Neural Networks for Position Prediction ..    | 47 |
| 5.3 Performance Evaluation .....   | 50 |
| 5.3.1 Dataset and Preprocessing .....                                    | 50 |
| 5.3.2 Training and Evaluating the DyNets .....                           | 51 |
| References .....   | 53 |
| <b>6 Open Challenges for IoUT Networks .....</b>                         | 55 |
| 6.1 Communication Challenges .....                                       | 55 |
| 6.1.1 Media Characteristics .....  | 55 |
| 6.1.2 Underwater Channel Modeling .....                                  | 56 |
| 6.1.3 Complexity of the Hardware and Other Combined Technologies .....   | 56 |
| 6.2 ML Challenges .....  | 57 |
| 6.2.1 Datasets Availability .....  | 57 |
| 6.2.2 Problem Formulation and Model Selection .....                      | 58 |
| 6.2.3 Feasibility of Online Learning .....                               | 58 |
| 6.2.4 Reduced Computational Complexity Models .....                      | 58 |
| References .....   | 58 |
| <b>Index .....</b>   | 61 |

# Acronyms

|        |                                      |
|--------|--------------------------------------|
| ADCs   | Analog-to-Digital Converters         |
| AF     | Amplify-and-Forward                  |
| ANNs   | Artificial Neural Networks           |
| AoA    | Angle of Arrival                     |
| AWGN   | Additive White Gaussian Noise        |
| BER    | Bit-Error-Rate                       |
| Bval   | Best Validation                      |
| CFNN   | Cascade-Forward Neural Network       |
| CM     | Confidence Metric                    |
| DDNN   | Distributed Delay Neural Network     |
| DF     | Decode-and-Forward                   |
| DNN    | Deep Neural Network                  |
| DT     | Decision Tree                        |
| DyNets | Dynamic Neural Networks              |
| FFNN   | Feed-Forward Neural Network          |
| GPS    | Global System Positioning            |
| IoT    | Internet of Things                   |
| IoUT   | Internet of Underwater Things        |
| KPV    | Key Physical Variables               |
| ML     | Machine Learning                     |
| MSE    | Mean Squared Normalized Error        |
| NAR    | Nonlinear Autoregressive             |
| NARX   | Nonlinear Autoregressive Exogenous   |
| NNs    | Neural Networks                      |
| pH     | Power of Hydrogen                    |
| PSK    | Phase Shift Keying                   |
| QAM    | Quadrature Amplitude Modulation      |
| RNN    | Recurrent Neural Network             |
| ROV    | Remotely Operated Underwater Vehicle |
| RSS    | Received Signal Strength             |
| RSSI   | Received Signal Strength Indicator   |

|       |                                     |
|-------|-------------------------------------|
| SER   | Symbol Error Rate                   |
| TDLs  | Tapped Delay Lines                  |
| TDNN  | Time Delay Neural Network           |
| TDoA  | Time Difference of Arrival          |
| ToA   | Time of Arrival                     |
| TPLs  | Transmission Power Levels           |
| UWSNs | Underwater Wireless Sensor Networks |