

## Abstract Marine Oil Spill Detection with Deep Neural Networks <sup>+</sup>

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- <sup>†</sup> Presented at the 3rd International Electronic Conference on Processes—Green and Sustainable Process Engineering and Process Systems Engineering (ECP 2024), 29–31 May 2024; Available online: https://sciforum.net/event/ECP2024.

Keywords: artificial intelligence; deep neural networks; image classification; oil spill detection

Oil spills, primarily due to accidents involving pipelines, tankers, and storage facilities, significantly impact marine life, particularly fish and shellfish. These leaks harm the fish and shellfish living in the seas. Within the scope of this study, open-source ocean images were used, with or without oil spills. In the dataset, there are 150 images in each class, and a total of 300 images. Before the training in deep neural networks for oil spill detection, data augmentation by rotating the dataset at certain angles to the right and left and various data preprocessing processes were applied. After the data augmentation process, the total amount of the dataset was increased from 300 to 400. In addition, as a dataset distribution, there are 320 images in the training data and 80 images in the validation data. Validation data were also used as test data. Deep learning-based deep neural networks MobileNetV2, Convolutional Vision Transformer (CvT), and ConvNeXT were used for classification processes in which oil spill detection with artificial intelligence is performed. When the application results were examined, it was observed that the classification results in deep neural networks had an accuracy rate of over 90%. Future work will aim to develop a real-time object detection interface that classifies and accurately locates oil spills.

**Author Contributions:** Conceptualization, F.U., M.G. and F.H.; methodology, F.U., M.G. and F.H.; software, F.U.; validation, F.U., M.G. and F.H.; writing—review and editing, F.U., M.G. and F.H. All authors have read and agreed to the published version of the manuscript.

Funding: This research received no external funding.

Institutional Review Board Statement: Not applicable.

Informed Consent Statement: Not applicable.

**Data Availability Statement:** The open source dataset used in the study was collected from Google Images in January 2023 under a Creative Commons license. This dataset are available from the corresponding author upon reasonable request.

Conflicts of Interest: The authors declare no conflicts of interest.

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Citation: Uysal, F.; Güven, M.; Hardalaç, F. Marine Oil Spill Detection with Deep Neural Networks. *Proceedings* 2024, 105, 4. https://doi.org/10.3390/ proceedings2024105004

Academic Editor: Michael C. Georgiadis

Published: 28 May 2024



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