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Deep Learning for Detecting Amphoras in Ancient Shipwrecks

by

Tianyao Chen

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Supervisor: Prof. Dr. Andreas Birk

Jacobs University Bremen | Department of Computer Science and Electrical Engineering

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Abstract

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1 Introduction

1.1 Motivation

1.1.1 Relevance of Amphoras

The name *amphora* is derived from the Greek word *amphoreus*, which literally means "two-handled" [1, 2]. It is the combination of two linguistic roots: *amphi* (on both sides) and *phoreus* (bearer) [1, 2]. Amphoras (or amphorae) were commercially used from 1500 B.C.E. to 500 C.E. to ship products throughout the Mediterranean, supplying the ancient Greek and Roman empires [2]. Amphoras were designed to ship large quantities of liquid (wine, olives, and oils) and dry products (grain, nuts, and salted fish) [2].

Like many measures that are named after the packages, amphoras were also a semi-standard unit of liquid measure [2]. A cargo ship's capacity was measured by the number of amphoras it could carry instead of by weight [2, 3].

The structurally strong egg-like shape and the high volume-to-weight ratio made amphoras very efficient packages [2]. Amphoras were by far the most common cargo type in Mediterranean shipwreck analysis; more than half of the ships only carried amphoras [2, 4].

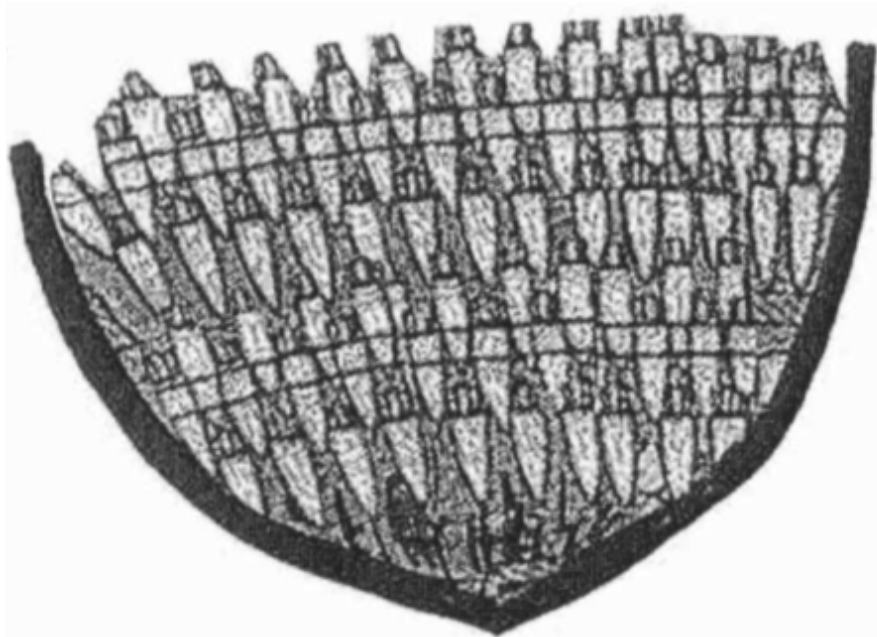


Figure 1: The egg-like shape enabled amphoras to interlock and minimize the waste of space on a ship. Source: [2].

Amphoras' various shapes and markings - which changed by time, region, producer, contents, and brand identity - were used to identify the package status and the different products inside [2].

Amphoras have great significance in archaeology. They can be used as evidence for the trade patterns throughout the Mediterranean [2]. As they were usually discarded at the destination of a trade and have been found in shipwrecks, archaeologists have been

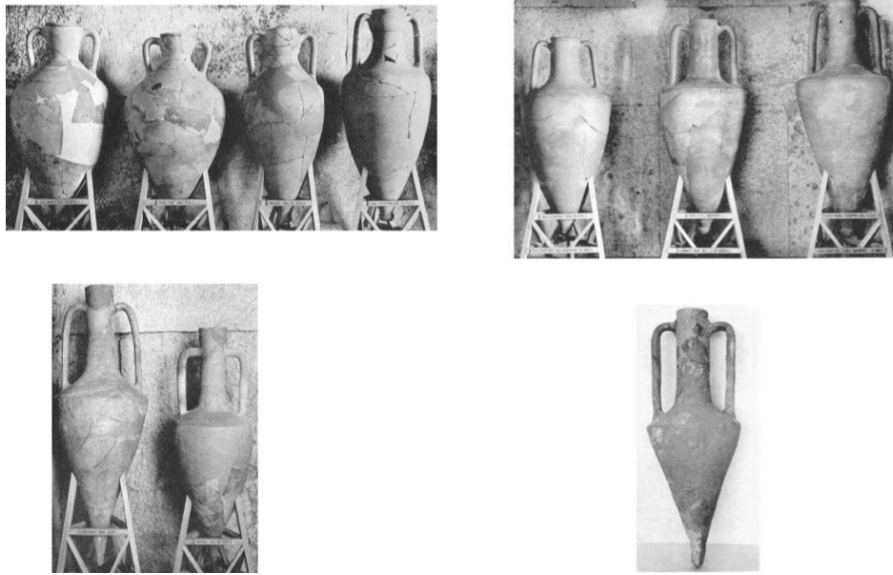


Figure 2: Amphoras have various shapes. Source: [2].

using them to recreate the transit routes [2]. Furthermore, researchers have been able to classify different amphoras, which also helps to date ruins and shipwrecks [2].

1.1.2 Computer Vision for Underwater Archaeology

Computer vision is the science of perceiving and understanding the world through images and videos [5]. There have been multiple exciting applications of computer vision, including image classification [6], object detection and localization [7, 8], art generation (neural style transfer) [9], image creation with Generative Artificial Networks (GAN) [10], face recognition [11], action and activity recognition [12], human pose estimation [13], and image recommendation system [14].

However, there is still limited research for the application of computer vision and machine learning in archaeology, especially underwater archaeology, compared to other domains [15, 16]. Computer vision, instead of visual inspection, could be used to automate the detection, assessment, and classification of artifacts [15].

Underwater computer vision has proven to be challenging, largely due to: 1) the distortion and attenuation caused by light propagation in water, and 2) the unrestricted natural environment with the abundance of marine life and suspended particles [16, 17, 18].

Despite the challenges, computer vision has lower cost [17] compared to sonar imagery [19] and laser scanning [20]. Plus, the increasingly abundant visual data obtained through autonomous underwater vehicles (AUVs), unmanned underwater vehicles (UUVs) [18, 21], and seafloor cabled observatories [16] enables us to utilize deep learning.

Furthermore, the research for deep-water shipwrecks is even more limited, mostly due to the lack of information and accessibility [22]. However, the need to study deep-water sites are in high demand, as the threats to these sites are increasing [22]. One major threat is the new forms of trawling that destroy the surface of these sites and interfere with the readability [22]. This means that many shipwrecks are likely to be damaged before they

can be studied [22]. It is thus crucial to implement efficient, accessible, and accurate techniques like deep learning based computer vision to study deep-water shipwrecks.

1.2 Deep Learning

Deep learning is a class of machine learning algorithms that build neural networks with more than one layer between input and output [23, 24]. Deep learning builds complex representations by combining simpler ones from the previous layers [25].

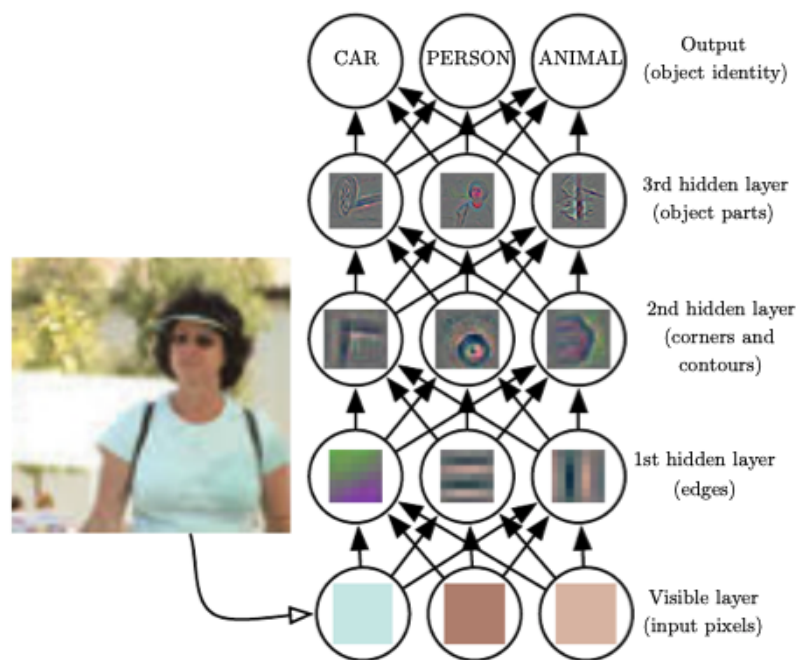


Figure 3: A deep learning system that learns the representations of a person. This is achieved by combining simpler features like corners and contours, which are further expressed by combining simpler features like edges. Source: [25]

1.2.1 Artificial Neural Networks (ANN)

1.2.2 Convolutional Neural Networks (CNN)

1.2.3 Deep Learning vs. Traditional Computer Vision

1.3 Object Detection

Define object detection and introduce the sliding CNN approach.

1.3.1 Fully Convolutional Networks (FCN)

1.3.2 General Object Detection Framework Components

Region Proposals

Network Predictions

Non-Maximum Suppression (NMS)

Metrics

1.3.3 Region-Based Convolutional Neural Networks (R-CNN)

R-CNN

Faster R-CNN

Faster R-CNN

1.3.4 Single Shot Detector (SSD)

1.3.5 You Only Look Once (YOLO)

YOLO

YOLOv2

YOLOv3

YOLOv4

YOLOv5

2 Related Work

3 Data and Methods

This is the technical core of the thesis. Here you lay out your how you answered your research question, you specify your design of experiments or simulations, point out difficulties that you encountered, etc.

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3.1 Data

3.2 Model

3.3 Model Training

4 Evaluation

This section discusses criteria that are used to evaluate the research results. Make sure your results can be used to published research results, i.e., to the already known state-of-the-art.

(target size: 5-10 pages)

Number	Description
7	A lucky number in Western culture
8	A lucky number in Chinese and other Asian cultures
42	Answer to the ultimate question of life, the universe, and everything
404	Not found

Table 1: Useless insights I gained with no further meaning

4.1 Visual Evaluation

4.2 Metric Evaluation

5 Conclusions

Summarize the main aspects and results of the research project. Provide an answer to the research questions stated earlier.

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6 Future Work

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