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Learning-Based Hyperspectral Image Compression Using A Spatio-Spectral Approach

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Eidesstattliche Erklärung

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Acknowledgements

Abstract

This template is intended to give an introduction of how to write diploma and master thesis.

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List of Acronyms

3GPP 3rd Generation Partnership Project AJAX Asynchronous JavaScript and XML API Application Programming Interface

AS Application Server

CSCF Call Session Control Function

CSS Cascading Stylesheets
DHTML Dynamic HTML

DOM Document Object Model

FOKUS Fraunhofer Institut fuer offene Kommunikationssysteme

GUI Graphical User Interface
GPS Global Positioning System

GSM Global System for Mobile Communication

HTML Hypertext Markup Language HSS Home Subscriber Server HTTP Hypertext Transfer Protocol

I-CSCF Interrogating-Call Session Control Function

IETF Internet Engineering Task Force

IM Instant Messaging

IMS IP Multimedia Subsystem

IP Internet Protocol
J2ME Java Micro Edition
JDK Java Developer Kit

JRE Java Runtime Environment
JSON JavaScript Object Notation
JSR Java Specification Request
JVM Java Virtual Machine
NGN Next Generation Network

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1 Introduction

1.1 Motivation

Hyperspectral imaging, the technique of capturing images with a camera that has a spectrometer for every pixel of the image, is a quickly growing field. There are many applications for the use of these images as much information can be extracted from the combination of spatial and spectral data in them.

One such application is in agriculture where it is used to detect the presence of the toxic metal cadmium in soil, lessening the need for chemical methods that require the plant to be harvested to test for cadmium stress. Using hyperspectral imaging the detection of cadmium can be performed on living plants [1].

Hyperspectral imaging is also being used to separate plastic waste where a machine learning model based on hyperspectral images can separate twelve kinds of plastics better than previously used methods such as near-infrared technology [2].

Another field where the use of hyperspectral imaging is rising in importance is in geology and environmental sciences. An example of this is the EnMAP mission which launched a satellite into earths orbit in 2022 that takes images of the surface of the earth with a comparatively high spatial resolution, allowing for regional geographic analysis [3].

This has already lead to many interesting studies including of glacier ice surface properties of the Ice Sheet in South-West Greenland, moisture content of soil below grassland and identification of specific crop traits such as the chlorophyll content or the leaf water content [4][5][6].

A dataset created from the images produced by the EnMAP mission is also the main dataset studied in this thesis.

The increasing usage of hyperspectral imaging makes it essential to address the disadvantes of the technology. One major disadvantage of this type of imaging is the size of the resulting images. Because there are much more spectral bands compared to the three bands of red, green and blue in traditional photography the resulting file size rises

accordingly. A hyperspectral image might for example have 300 bands resulting in an image that is a hundred times larger than an RGB image with the same pixel density. In addition to this it is important to notice that even RGB images are too large for many use cases which is why there are many compression standards for RGB images in wide usage.

Furthermore, compression of hyperspectral images is a complex problem. This arises from the combination of normal image compression complexities and the added challenge of encoding information in the spectral dimension which RGB compression algorithms rarely address. The latter is one of the reasons why many RGB compression algorithms do not perform well on hyperspectral data. In addition to this there are more technical reasons for the difficulty of adapting algorithms for RGB compression that will be explored in this thesis.

For these reasons it is clear that it is important to find good compression algorithms for hyperspectral images.

Possible algorithms for image compression can be categorized into multiple broad groups. They can firstly be grouped by whether they compress the image losslessly. Lossless algorithms restore the compressed image exactly whereas lossy algorithms introduce some distortion during the compression. The disadvantage of lossless compression is however that there are mathematical limits to the compression rate given by the entropy in the data that is to be compressed. Lossy compression methods can instead adapt their rate of compression based on either the desired amount of distortion or the target compression rate. They can also be combined with a lossless compression method to further optimise their performance. For these reasons this thesis focuses on lossy compression methods.

The category of lossy compression algorithms can be further divided into traditional compression methods and the learning-based compression methods. Traditional compression methods employ algorithms such as the discrete cosine transform which is used in the JPEG image compression standard or the wavelet transform, often in combination with a lossless entropy coding method.

Learning-based approaches use the powerful capability of artificial neural networks to universally approximate arbitrary functions using gradient descentCITATION. These networks are then used to build models that learn to compress and decompress images. Lately these networks have been shown to outperform traditional compression methods for many applications, especially for RGB images.

Compression for hyperspectral images is in the early stages of research, however even in this domain there are promising results showing the capabilities of learned autoencoders for this task.

1.2 Objective

This thesis addresses the problem of hyperspectral image compression using machine learning models consisting of two parts, an encoder and a decoder.

The encoder learns to map the original image to a smaller latent dimension. Analogously, the decoder is trained to map elements from this latent dimension to full-size images that are as close as possible to the original image.

Contributions to the research in on learned hyperspectral image compression are made in this thesis by introducing a new architecture that enables the use of spatial compression algorithms such as models that might be used for RGB image compression by combining them with a model that performs compression in the spectral domain while keeping spatial relationships intact. In this way, it is possible to compress the spatial information in hyperspectral images using models that are not ordinarily applicable to these images.

Using this architecture, multiple model combinations are designed and compared with each other as well as with the base versions of these models.

These models include models based on convolutional neural networks (CNNs) as well as transformer-based architectures.

A model using a hyperprior architecture is also used for the spatial compression. This architecture yields much higher compression ratios than other models by using an arithmetic encoder in the bottleneck. With this model compression ratios much higher than the current state of the art for hyperspectral image compression are achieved while distortion is only reduced by a comparatively small amount. TODO besser formulieren, bzw. überlegen wie viel hier erklärt werden sollte zu hyperprior etc.

Additionally, the latent spaces of both the encoder in the spectral domain as well as the latent space of the spatial encoder will be analysed.

1.3 Outline

This section gives a brief introduction into the chapters in the thesis, which is split into 7 chapters.

Chapter 2 is called 'Related Work'. In this chapter the current state of research is described. This relates to the research of hyperspectral image compression by itself,

however, as it is a recent and relatively unexplored field of study, the adjacent research topic of RGB image compression is also explored. There is also some exploration of the general studies done on convolutional neural networks as well as transformers.

Chapter 3, 'Theoretical Foundations' gives an overview of the theoretical ideas used in the proposed methods, such as CNNs, transformers, arithmetic coding as well as the hyperprior architecture for compression.

Chapter 4, 'Methodology', details the individual submodels making up the models that are proposed to address the hyperspectral image compression problem as well as the models in their totality. It also gives an overview over the loss functions used for training the models, one of which is a loss function specifically designed for the models proposed in this thesis.

Chapter 5, 'Implementation', gives a brief overview of the frameworks and libraries used in the implementation of the methods used in the thesis. It also describes the overall code architecture and structure. Furthermore the tools used for structuring and viewing the experiments are explained.

Chapter 6, 'Experiments', explains the design of the experiments done, the dataset that is used for these experiments as well as the results of these experiments. It also details some of the challenges encountered during the experimentation phase.

Chapter 7, 'Conclusion and discussion', gives a summary of the results from the thesis as well as possible improvements that could be made as well as ideas that could be explored in future research.

2 Related Work

- 2.1 Hyperspectral image compression
- 2.1.1 CNN-based architectures
- 2.1.2 Transformer-based architectures
- 2.2 RGB image compression
- 2.2.1 CNN-based architectures
- 2.2.2 Transformer-based architectures
- 2.3 ZZZ

3 Theoretical foundations

This chapter describes the implementation of component X.

- 3.1 Convolutional Neural Networks
- 3.2 Transformer Models
- 3.3 Arithmetic Coding
- 3.4 Hyperprior Architecture

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4 Methodology

- 4.1 Spectral Autoencoder Methods
- 4.1.1 Pixel-Wise Convolutional Neural Network
- 4.1.2 Two-dimensional CNN with Low Kernel Size
- 4.1.3 Spectral Transformer Model
- 4.2 Spatial Autoencoder Methods
- 4.2.1 CNN-based Spatial Autoencoder
- 4.2.2 Hyperprior-based Spatial Autoencoder
- 4.2.3 Attention-based Model Using Hyperprior Architecture
- 4.3 The Combined Model
- 4.4 Loss functions
- 4.4.1 MSE Loss
- 4.4.2 Rate Distortion Loss
- 4.4.3 Dual MSE Loss

5 Implementation

This chapter describes the implementation of component X.

- 5.0.1 Frameworks and libraries used
- 5.0.2 Model implementation structure
- 5.0.3 Metrics and data set implementation
- 5.0.4 Weights and Biases for experiments

6 Experiments

This chapter describes the implementation of component X.

- 6.1 Description of the Data Set
- 6.2 Design of Experiments
- 6.2.1 Metrics
- 6.3 Results of the XX
- 6.4 Results of the YY
- 6.5 Results of the ZZ

7 Conclusion and Discussion

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Appendix