Imperial College London

MENG INDIVIDUAL PROJECT

DEPARTMENT OF COMPUTING

IMPERIAL COLLEGE OF SCIENCE, TECHNOLOGY AND MEDICINE

Transfer Learning for bespoke automatic contouring of cervical cancer radiotherapy planning

Author: Anton Zhitomirsky Supervisor: Ben Glocker

Second Marker: TODO

January 9, 2024

Abstract

Clinicians target cancerous tumours by studying 3D contrasting images of cancerous tumours and surrounding soft tissues to plan targets for radiation therapy. The Royal Marsden Hospital is a key contributor of data for this project, which uses this approach to delineate tumours for cervical cancers. Typically after a gross tumour volume (GTV) is extrapolated from the relevant imaging modality, clinicians append tailored safety margins to also account for the microscopic cancerous spreads not visible in the scan to generate the planned target volume (PTV).

The PTV area has to be generous enough to attempt to treat the problem in one-shot, yet conservative enough to not harm surrounding healthy tissue with radiation over the course of the treatment. Compounded with small sample size of labelled data this proposes a significant challenge for developing deep-learning segmentation models to identify an optimal PTV.

Thus we propose a transfer learning strategy to utilize imaging models in similar domains to attempt to learn from the limited input size to provide clinicians with a faster and more accurate segmentation method.

Contents

1	Intro	oduction	1							
	1.1	Clinical Context	1							
	1.2	Motivation	1							
	1.3	Current Solutions	1							
	1.4	Outline of Report	1							
2	Bacl	nckground								
	2.1	Clinical Context	2							
	2.2	Vanilla Image Segmentation Models	2							
		2.2.1 Convolutional Neural Networks (CNN)	2							
		2.2.2 U-Net	2							
		2.2.3 nnU-Net	2							
		2.2.4 Traditional Limitations	2							
	2.3	Transfer Learning	2							
		2.3.1 Intuition	2							
		2.3.2 Total Segmentator	2							
		2.3.3 UniverSeg	2							
		2.3.4 SAM	2							
	2.4	Summary	2							
3	Base Line and Data									
	3.1	Data	3							
		3.1.1 Available Data	3							
		3.1.2 Data Cleaning	3							
	3.2	Evaluation Metrics	3							
		3.2.1 Geometric	3							
		3.2.2 Honorable mentions	3							
	3.3	Baseline Results	3							
		3.3.1 nnU-Net	3							
		3.3.2 Total Segmentator	3							
		3.3.3 UinverSeg	3							
	3.4	Summary	3							
4	Prop	posal	4							

CONTENTS

5	Ethics							
	5.1	Patient disclosures	5					
	5.2	Use in practice	5					
	5.3	Anonymisation of Medical Data	5					
	5.4	Using the tool	6					
Bibliography								

Introduction

- 1.1 Clinical Context
- 1.2 Motivation
- 1.3 Current Solutions
- 1.4 Outline of Report

Background

- 2.1 Clinical Context
- 2.2 Vanilla Image Segmentation Models
- 2.2.1 Convolutional Neural Networks (CNN)
- 2.2.2 U-Net
- 2.2.3 nnU-Net
- 2.2.4 Traditional Limitations
- 2.3 Transfer Learning
- 2.3.1 Intuition
- 2.3.2 Total Segmentator
- 2.3.3 UniverSeg
- 2.3.4 SAM
- 2.4 Summary

Base Line and Data

- 3.1 Data
- 3.1.1 Available Data
- 3.1.2 Data Cleaning
- 3.2 Evaluation Metrics
- 3.2.1 Geometric
- 3.2.2 Honorable mentions
- 3.3 Baseline Results
- 3.3.1 nnU-Net
- 3.3.2 Total Segmentator
- 3.3.3 UinverSeg
- 3.4 Summary

Proposal

Ethics

This project involves very intimate and personal information of many female patients with the end goal of learning to plot accurate contours for radiotherapy planning. Researchers should dedicate a large effort to anonymize data because the lack there-of may result in "stigma, embarrassment, and discrimination" [2]. It is important in any setting that data used (for instance in a learning environment) cannot be reverse engineered.

5.1 Patient disclosures

Royal Marsden Hospital has the right to share their clinical data with outside entities if this is disclosed to the patient, and if formalities are arranged with external parties. With no specific customer consent, Royal Marsden is entitled to share data with outside entities as long as they are made aware of the ways their data may be used and as long as the external entities act as "ethical data stewards" [1]. As such Royal Marsden Hospital mentions in their privacy policy that "explicit consent may not be required if the information being used has been de-identified/anonymized" [3]. Without the anonymisation of data patients may be reluctant to provide candid and complete disclosures of their sensitive information, even to physicians which may prevent a full diagnosis if their data isn't maintained in an anonymous fashion.

5.2 Use in practice

The MIRA team acts as responsible data stewards by storing anonymized data within a folder on the college network. This folder contains security measures which limit the availability of data only to those with specific access rights. Furthermore, operating on the pre-amble of de-identified data further reduces individual patient risk in the event that data is ever brought outside the confines of this folder.

How was the data transferred securely?

Data protection and deleting the data after a number of years? (we're not subject to data protection laws because it is anonymous)

5.3 Anonymisation of Medical Data

5.4. USING THE TOOL Chapter 5. Ethics

Maybe refer back to section about describing the data [3.1.1] In progress to try and see the metadata.

5.4 Using the tool

Bibliography

- [1] David B. Larson et al. "Ethics of Using and Sharing Clinical Imaging Data for Artificial Intelligence: A Proposed Framework". In: (2020). URL: https://pubs.rsna.org/doi/full/10.1148/radiol.2020192536.
- [2] Nass SJ, Levit LA, and Gostin LO. "The Value and Importance of Health Information Privacy". In: (2009). URL: https://www.ncbi.nlm.nih.gov/books/NBK9579/.
- [3] The Royal Marsden NHS Foundation Trust. "Privacy Note". In: (2023). URL: https://rm-d8-live.s3.eu-west-1.amazonaws.com/d8live.royalmarsden.nhs.uk/s3fs-public/2023-10/T22020ac_Revised%20privacy%20policy_V1_AW_WEB.pdf.