

# Méthodes pour l'automatisation de la dosimetrie pour les traitements radiothérapiques.

Methods for automatization of the dosimetry for radiotherapy treatments.

#### Thèse de doctorat de l'université Paris-Saclay

Spécialité de doctorat: ... École doctorale n° 573 Interfaces : matériaux, systèmes, usages, ED INTERFACE Graduate School: Sciences de l'Ingénierie et des Systèmes, SIS

Thèse préparée dans les unités de recherche Radiothérapie (Institut Régionale du Cancer de Montpellier), Advanced Research (TheraPanacea), et MICS, Mathématiques et Informatique pour la Complexité et les Systèmes (Université Paris-Saclay, CentraleSupélec), sous la direction de Nikos Paragios, Professeur, et la co-direction de Paul-Henry Cournède, Professeur

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Rapporteur & Examinateur / trice

Examinateur ou Examinatrice

Examinateur ou Examinatrice



Titre: Méthodes pour l'automatisation de la dosimetrie pour les traitements radiothérapiques.

Mots clés: Mathématiques, Intelligence Artificielle, Radiothérapie

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Title: Methods for automatization of the dosimetry for radiotherapy treatments.

Keywords: Mathematics, Artificial Intelligence, Radiotherapy

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# Acknowledgments

A PhD is more than just hard work; it thrives on mentorship, collaboration, and unwavering support. [...]

# List of Contributions

- Teaching: Consistency and Reproducibility of Grades in Higher Education: A Case Study in Deep Learning replace icon
- ArXiV: Radiotherapy Dosimetry: A Review on Open-Source Optimizer
- ESTRO: A Novel Framework for Multi-Objective Optimization and Robust Plan Selection Using Graph Theory
- SFPM: Dose Volume Histograms Guided Deep Dose Predictions
- AIME: Radiotherapy Dose Optimization via Clinical Knowledge Based Reinforcement Learning (full paper coming soon)
- ASTRO: Clinically Dependent Fully Automatic Treatment Planning System
- SFRO: Attention Mechanism on Dose-Volume Histograms for Deep Dose Predictions

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# Background

#### Abstract

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### 1.1 Medical context

#### 1.1.1 10 cancer markers

- cell proliferation
- reprogram cellular metabolism
- stop cell growth arrest
- evade apoptosis
- escape immune system
- ability to undergo a sufficient number of successive cell cycles of growth and division to generate macroscopic tumors
- create new blog vessels to get nutriments
- allow cell escape and metastasis formation
- change cellular response phenotypic via plasticity
- senescence

#### 1.1.2 4 cancer conditions

- mutation
- epigenetic reprogramming
- inflammatory context
- disruption of microbiota

#### 1.1.3 phases of cancer

initiation

promotion

tumorigenesis + neo angiogenesis

evolution (local, regional, metastasis)

### 1.1.4 cancer classification:

tumor, node, mestastasis

#### stages classification:

- 1. stage 0 which corresponds to a so-called in situ tumor
- 2. stage 1 which corresponds to a single, small tumor
- $3. \,$  stage 2 which corresponds to a larger local volume
- 4. stage 3 which corresponds to invasion of the lymph nodes or surrounding tissues
- 5. stage 4 which corresponds to a wider extension in the body in the form of metastases

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## 1.1.5 treatment types

surgery

RT

chemotherapy

combination

## 1.2 Patient Path

- 1.2.1 Detection / diagnostic
- 1.2.2 RT Prescription
- 1.2.3 CT scan
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## 1.4 Irradiations techniques

#### 1.4.1 IMRT

Step and Shoot

Sliding Window

#### 1.4.2 VMAT

## 1.5 Treatment Planning Systems

#### 1.5.1 Manufacturer

Eclipse (Varian)

ONE | Planning (Elekta)

Precision (Accuray)

## 1.5.2 Non-manufacturer

RayStation (RaySearch)

matRad (German Cancer Research Center - DKFZ)

AutoPlan (TheraPanacea - coming soon)

# 1.6 Dosimetry steps

Challenges

- 1.6.1 BOO
- 1.6.2 FMO
- 1.6.3 LF
- 1.7 Simulation

# Introduction

#### Abstract

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2.2       Problematic         2.3       State of the Art								·	·										·																																								٠	٠					•	٠				٠	٠	٠	٠	•	٠		
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4 Unsolved problems																																																																													

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## 2.1 Context

Cancer; RT; optim to be done

## 2.2 Problematic

Manual optim is time consuming; need to automate

## 2.3 State of the Art

# 2.4 Unsolved problems

## 2.5 Contribution

# **Dosimetry Optimization**

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3.1	Optim engine: classic and dose mimicking	
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# Automation: Classical Approach

Abstract

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# $4.1 \quad RL \, + \, classic \, \, optim \, \, algo \, \, (AIME \, / \, \, ASTRO)$

# Automation: Deep Dose

Abstract

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# 5.1 DVH guided deep dose + dose mimicking algo (SFPM / SFRO)

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# Conclusion

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