Automation on RaySearch

ICTP School of Medical Physics for Radiation Therapy:

Dosimetry and Treatment

Planning for Basic and Advanced Applications

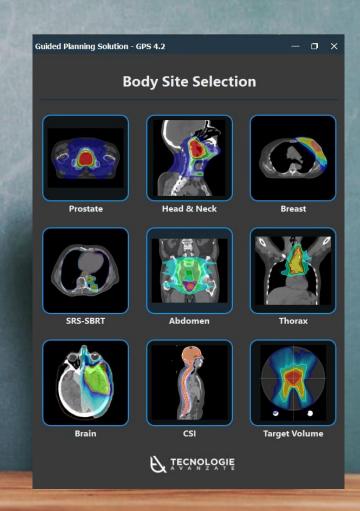
12 September 2025

Trieste



Christian Fiandra, PhD







RayStation infrastructure Citta della Salute e della Scienza, Turin, Italy Three hospitals

S. Giovanni







S. Anna

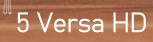








Radixact



RayStation server

First automation step: Images loading e contouring

Create patient in the database with images and structures

Philips CT Big Bore





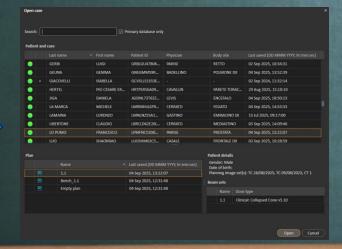
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Tag Dicom (0018,1030)	Structure template RayStation		
STANDARD/TECSPEC			
STEREO ENC 1,5 mm/TECSPEC	Brain Template and Couch SRS Template		
ENC 3 mm Vol/TECSPEC	Brain_Template		
TESTACOLL 2 mm Vol./TECSPEC	HN_Template		
Pulm Gating 4D PRO/TECSPEC	Lung_Template		
LIVER Gating 4D PRO/TECSPEC	Liver_Template		
FEGATO 4D SEMI 2 mm/TECSPEC	Liver_Template		
FEGATO / SURRENE NO 4D/TECSPEC	Liver_Template		
TORACE 3 mm/TECSPEC	Lung_Template		
SBRT VERTEBRATORACICHE/TECSPEC	Lung_Template		
TORACE 3 mm ESOFAGO/TECSPEC	Esoph_Template		
TORACE 3 mm LINFOMA/TECSPEC	LIMP_Template		
ABLAZIONE CARDIACA/TECSPEC	Lung_Template		
MAMMELLA 3 mm/TECSPEC	BREAST_Template		
MAMMELLA 3 mm C-PAP/TECSPEC	BREAST_Template		
PELVI 3 mm PROSTATA/TECSPEC	PROSTATE_Template		
PROSTATA IPO TOMO 2mm/TECSPEC	PROSTATE_Template		
PELVI 3 mm/TECSPEC	Rectum_Template		
SBRT VERTEBRALOMBARE/TECSPEC	Rectum_Template		
T.M.I. 5mm/TECSPEC	TMI/TMLI_Template		
T.M.L.I. 5mm/TECSPEC	TMI/TMLI_Template		
MEDULLO 3 mm/TECSPEC	TMI/TMLI_Template		
T.B.I. 5 mm/TECSPEC	TBI_Template		

First automation step Images loading e contouring

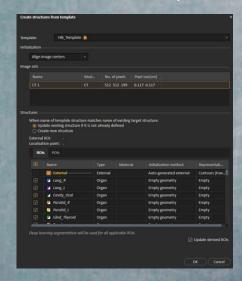
Database creation

(0018,0090) Data Collection Diameter | Keep | 600 | (0018,1020) | Software version(s) | Keep | 2.3.0 | (0018,1030) | Protocol Name | Keep | STEREO enc 1,5 mm/TECSPEC | (0018,1030) | Protocol Name | Keep | 600 | (0018,1030) | Protocol Name | Keep | 600 | (0018,1030) | Protocol Name | Keep | 600 | (0018,1030) | Protocol Name | Keep | 600 | (0018,1030) | Protocol Name | Keep | 600 | (0018,1030) | Protocol Name | Keep | 600 | (0018,1030) | (0018,1030) | Protocol Name | Keep | 600 | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,1030) | (0018,103

Dicom



Structures template





Deep learning segmentation	×
Primary image set	
CT: TC 26/08/2025 [26 Aug 2025, 12:43:43 (hr:min:sec)]
Select model ROI(s) (0 of 123 selected) ▼ □ ≫ RSL DLS CT	
➤ ☐ Abdomen ➤ ☐ Breast	
▶ □ Vessels	

Second automation step Planning



MCO

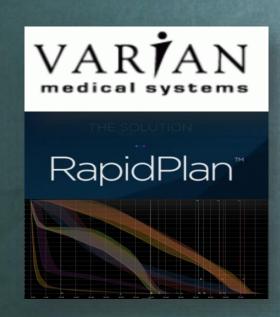
Targets:				
PTVAndEsoph+6, Max Dose	-6-			9
PTV, Min Dose	■ Ø•		-	
F PTV, Max Dose	-6·			
Organs at risk:				_
EsophMinFTV+5, Max Dose	-6-			
LungsMinPTV, Max EUD	-6-			
SpinalCord+4, Max Dose	-6-			.,4
External, Dose Fall-Off	86·			







Guided Planning Solution



Elekta ONE® Planning



M-cycle



AAPM REPORT NO. 263



Standardizing Nomenclatures in Radiation Oncology

The Report of AAPM Task Group 263

January 2018

Standardization is key to improve scriptbased plan evaluation Reduce variability and inconsistencies

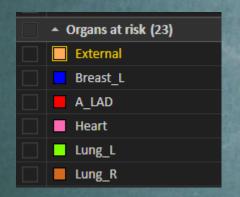


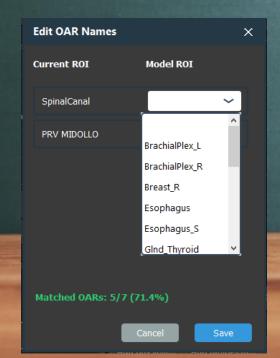
External Brainstem Brainstem_PRV SpinalCord SpinalCord_PRV Parotid_R Parotid_L Cavity_Oral Mandible Cochlea_R Cochlea_L Glnd_Thyroid

Prostate



Breast

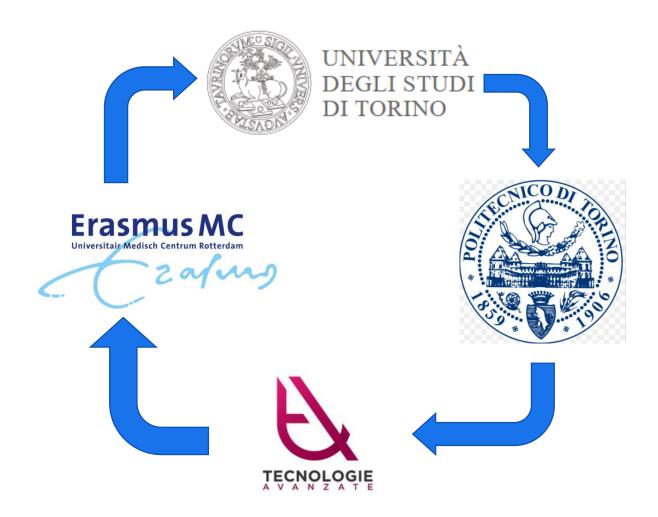




SBRT Lung



Genetic Planning Solution - GPS



GPS

OARs

$$EUD = \left(rac{1}{N}\sum_{i=1}^{N}d_i^a
ight)^{rac{1}{a}}$$

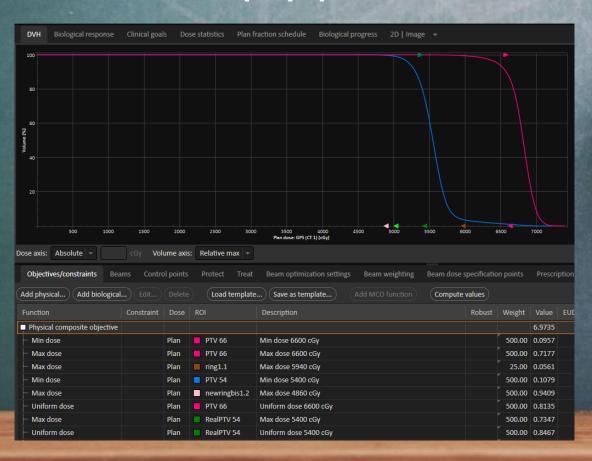
where:

- N = number of voxels (or dose bins) in the structure
- d_i = dose in voxel i
- a = tissue-specific parameter
 - Negative $a \rightarrow$ makes the calculation tumor-control-oriented (focuses on underdosed regions)
 - Positive a → makes the calculation normal-tissue-oriented (focuses on hotspots)
 - $a o \infty$ o EUD approaches the maximum dose
 - a=1 \rightarrow EUD becomes the mean dose



··· Max EUD	Plan	Femur_Head_R	Max EUD 685 cGy, Parameter A 1
··· Max EUD	Plan	Femur_Head_L	Max EUD 765 cGy, Parameter A 1
··· Max EUD	Plan	Bladder1	Max EUD 1611 cGy, Parameter A 1
··· Max EUD	Plan	Rectum1	Max EUD 1464 cGy, Parameter A 1
···· Max dose	Plan	Bowel_Small	Max dose 5000 cGy
··· Max EUD	Plan	Bowel_Small	Max EUD 4000 cGy, Parameter A 10
··· Max EUD	Plan	Bowel_Small1	Max EUD 477 cGy, Parameter A 1
··· Max EUD	Plan	PenileBulb1	Max EUD 1946 cGy, Parameter A 20

PTV



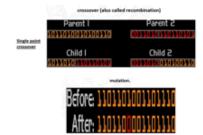
Genetic Planning Solution 2.2

Genetic algorithm









Same fitness function for all centers no center-specific tuning is applied!!







A genetic algorithm was implemented in the RayStation treatment planning system (Version 8A) using Python code inside a platform called Genetic Planning Solution

Heuristic Optimization - Genetic Algorithm



Automated Heuristic Optimization of Prostate VMAT Treatment Planning

Christian Fiandra', Alessandro Alparone', Elena Gallio', Claudio Vecchi', Gabriella Balestra', Sara Bartoncini', Samanta Rosati', Riccardo Ragona', Umberto Ricardi'

"Department of Choolings, Radiation Choolings, Colorently of Turins, Turin, Italy "Princolings of Amuses T.A. Sell, Turin, Italy "Medical Physics Units, A.O.U. Onto della Subner della Science, Turin, Italy "Department of Herchrostics and Tolerommunication, Politomico di Torrion, Turin, Italy "Radiation Oscolings Units, A.O.U. Onto della Subner e della Science, Turin, Italy "Radiation Oscolings Units, A.O.U. Onto della Subner e della Science, Turin, Italy

GAs are a particular class of evolutionary algorithms that use techniques inspired by evolutionary biology such as inheritance, mutation, selection, and crossover (also called recombination)

Abstra

Purpose: To investigate a genetic algorithm approach to automatic treatment planning. Methods: A Python script based on genetic algorithm (GA) was implemented for VMAT treatment planning of prostate tumor. The script was implemented in RayStation treatment planning system using Python code. Two different clinical prescriptions were considered: 78 Gy prescribed to planning target volume in 39 fractions (GROUP 1) and simultaneous integrated boost (70.2 Gy to prostate bed and 61.1 Gy to seminal vesides) in 26 fractions (GROUP 2). The script automatically optimizes doses to PTV and OARs according to GA. A comparison with corresponding plans created with Monaco TPS (M) and Auto-Planning module of Pinnacle' (AP) was carried out. The plans were evaluated with a total score (TS) of PlanIQ software in terms of target coverage and spuring of OARs as well as clinical score (CS) performed by a Radiation Oncologist. Results: In GROUP 1, mean value of TS were 150.6 ± 30.7, 146.3 ± 36.1 and 137.4 ± 35.7 for AP, GA and M respectively. For GROUP 2, mean value for TS were 163.5 \pm 16.8, 163.4 \pm 24.7 and 162.9 ± 16.6 for AP, GA and M respectively with no significance differences. In terms of CS, the highest value has been attributed to GA in four patients out of five for both GROUP 1 and 2. Conclusions: Genetic approach is practicable for prostate VMAT plan generation and studies are underway in other anatomical sites such as Head and Neck and Rectum.

AAPM REPORT NO. 263



Standardizing Nomenclatures in Radiation Oncology

The Report of AAPM Task Group 263

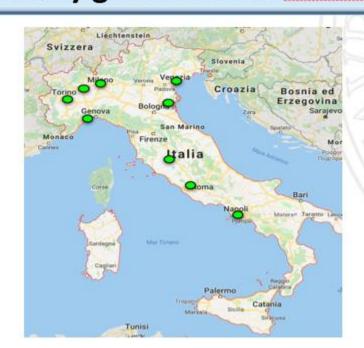
January 2018

VMAT Prostate radiotherapy treatment planning

10 participating Italian centers

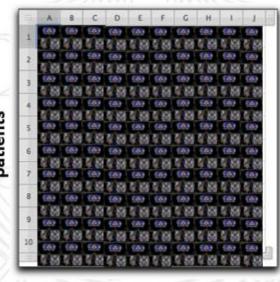
10 prostate patients

manually generated Vs AutoGPS



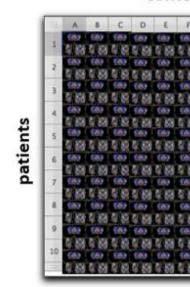
100 Manual plans

centers



100 AutoGPS plans

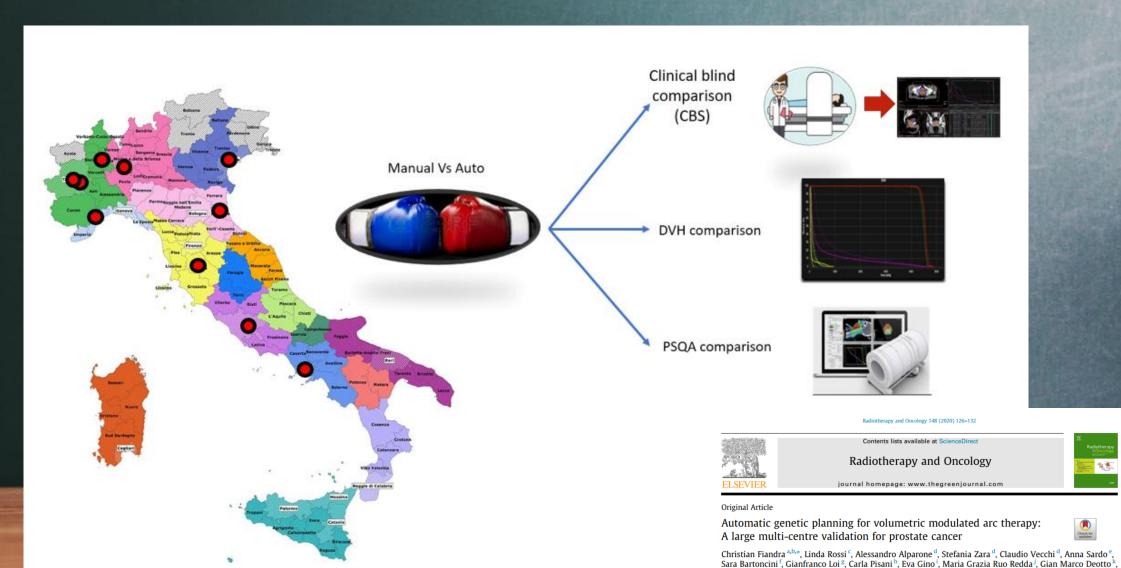
centers





Vs

GPS in clinical practice - prostate

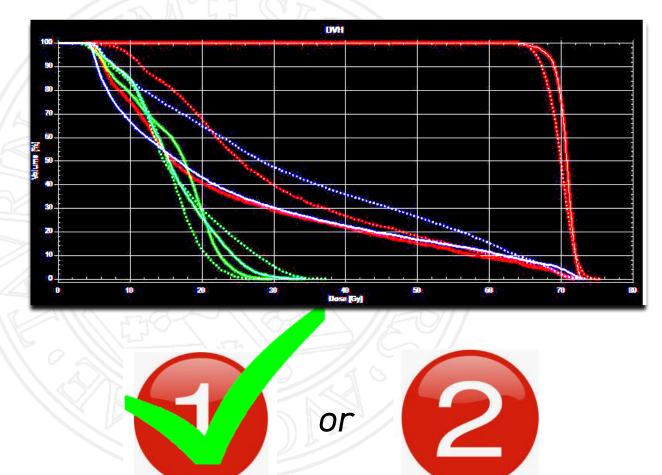


Paolo Tini¹, Stefania Comi^m, Dario Zeriniⁿ, Gianluca Ametrano^o, Valentina Borzillo^o, Lidia Strigari^{p,1}, Silvia Strolin^{p,1}, Alessandro Savini^q, Antonino Romeo^r, Sonia Reccanello^s, Imad Abu Rumeileh^t, Nunzia Ciscognetti^u, Flavia Guerrisi^v, Gabriella Balestra^w, Umberto Ricardi^x, Ben Heijmen^c

Validation Which is the better plan?





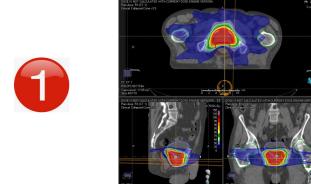


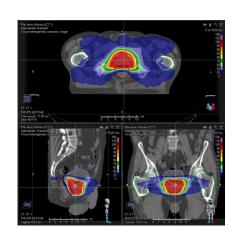


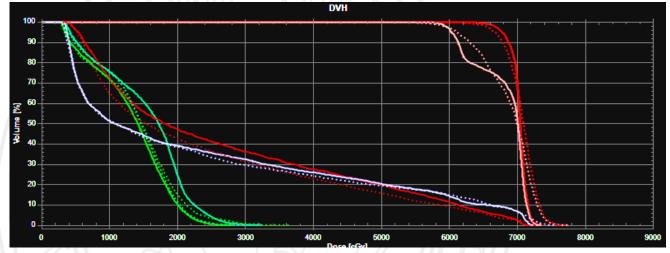




Validation Which is the better plan?







or

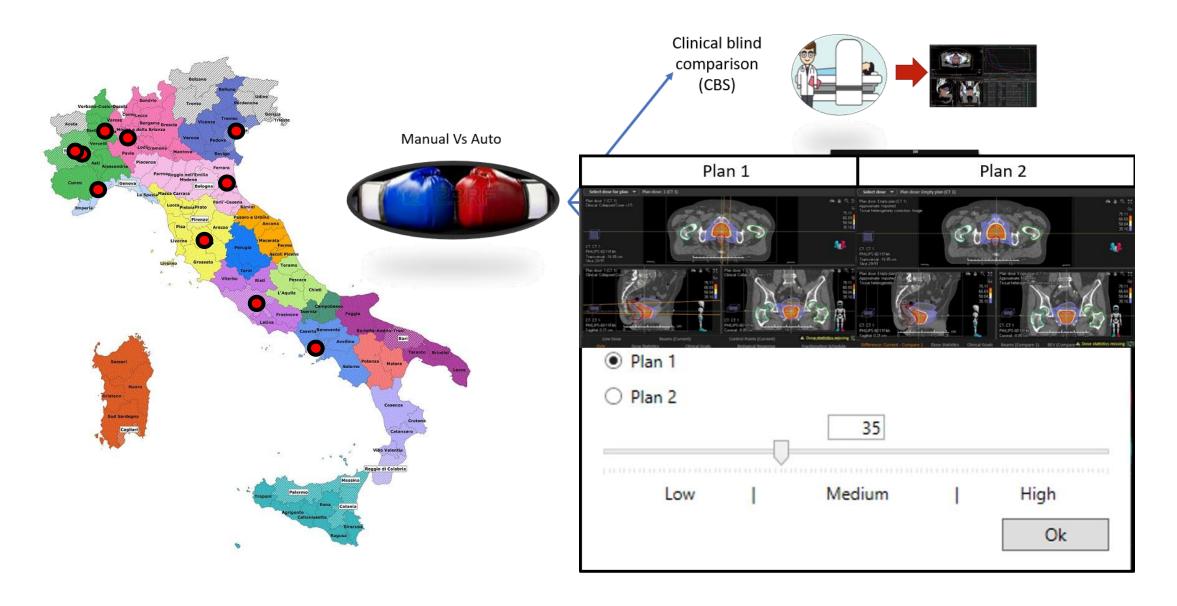








Multi-centre prostate

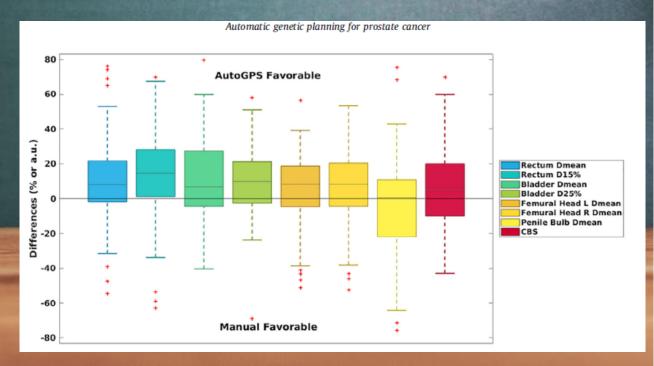


GPS in clinical practice – Prostate

Clinical Blind Score

AutoGPS Favorable Patient 1 Patient 2 Patient 3 Patient 3 Patient 5 Patient 5 Patient 7 Patient 7 Patient 8 Patient 9 Patient 10 L Centres

Dosimetric OAR parameters



GPS in clinical practice - prostate

· Goal: Validate a single-configuration genetic autoplanning system (GPS) for prostate VMAT across 10 centres.

Key results:

- 91% of autoGPS plans were clinically acceptable.
- 69% of acceptable plans were preferred over manual ones.
- Better rectum/bladder sparing with similar PTV coverage.

· Variability:

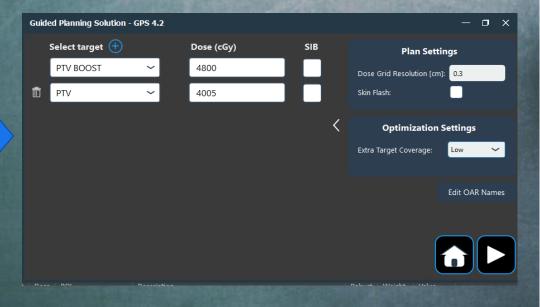
- Inter-centre differences linked to protocol mismatch and manual plan quality.
- GPS could help standardize plan quality across centres.
- · Centre H issue: Solved by minor configuration adjustment to match local coverage criteria.

Strengths & implications:

- Largest multi-centre autoplanning validation to date.
- Shows potential to reduce workload, improve quality, and avoid centre-specific tuning.
- · Limitations: Small sample per centre and QA checks done only in one centre

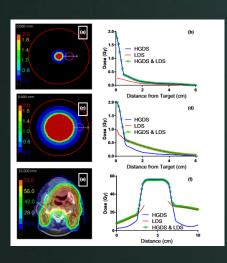
GPS 4.2

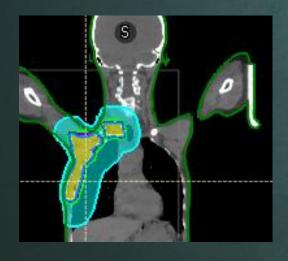




GPS 4.2

Benchmark dose





A method for *a priori* estimation of best feasible DVH for organs-at-risk: Validation for head and neck VMAT planning

Saeed Ahmed

Department of Physics, University of South Florida, Tampa, FL 33612, USA

Benjamin Nelms

Canis Lupus LLC, Merrimac, WI 53561, USA

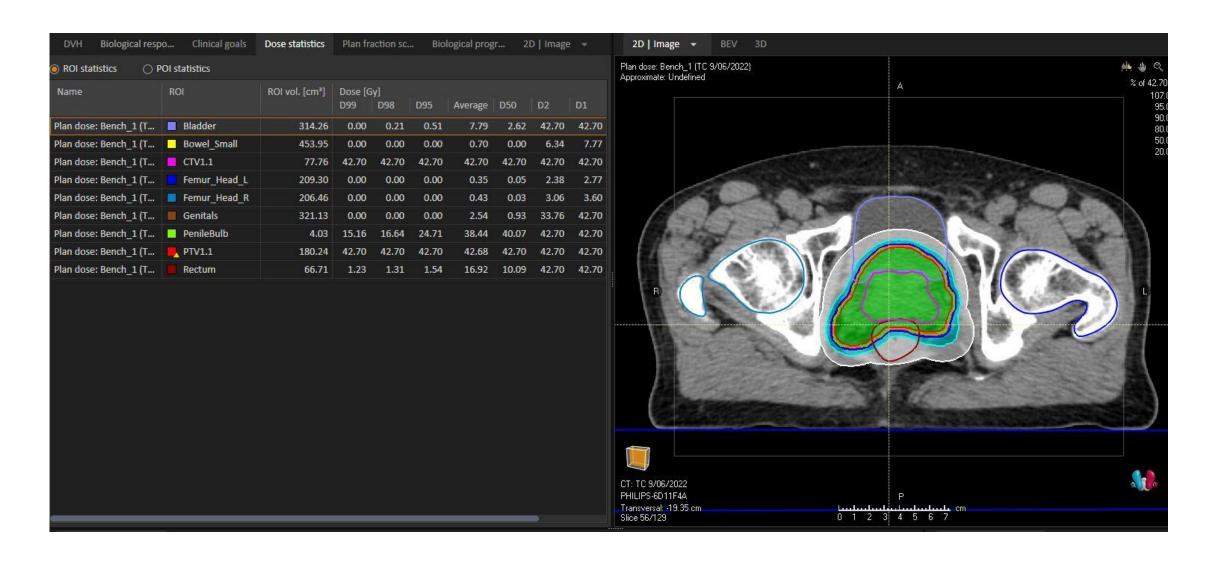
Dawn Gintz, Jimmy Caudell, Geoffrey Zhang, Eduardo G. Moros, and Vladimir Feygelman^{a)}
Department of Radiation Oncology, Moffitt Cancer Center, Tampa, FL 33612, USA

(Received 27 March 2017; revised 24 July 2017; accepted for publication 24 July 2017; published 31 August 2017)

Patient specific optimization function



Benchmark dose



- 1. Generation of a D_0 dose grid of the required resolution, the values are given by the **dose prescribed** to the target(s) within the target(s) and zero outside.
- 2. Generation of a **D1** dose grid, the DO dose is evolved applying to each axial plane a series of convolutions through symmetrical kernels that take into account the Low-Gradient Dose Spread component both at mid-range and far-range.

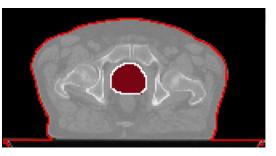
 $D1[y][x,z] = (D_o[y][x,z] * LGDS-Mid_E[x,z]) * LGDS-Far_E[x,z]$

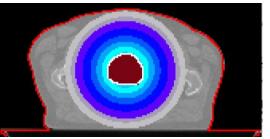
3. Generation of a **D2** dose grid, the DO dose is evolved outside the target(s) assigning to each voxel external to the target(s) the value given by the maximum diffuse value searched among all the nearby target voxels:

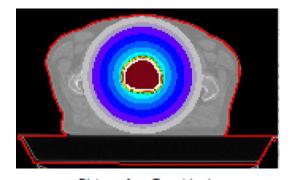
 $D2[x,y,x] = max(x_T,y_T,z_T) \{ Do[x_T,y_T,z_T] \times HGDS_E[r,r_{rad}] \}$

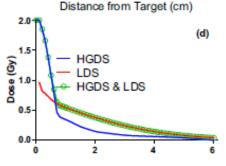
Where $HGDS_{E}[\mathbf{r},\mathbf{r}_{rad}]$ is the High-Gradient Dose Spread function for a given beam energy E, \mathbf{r} is the physical distance between the voxel to be filled [x,y,z] and the target voxel $[x_{T},y_{T},z_{T}]$, and \mathbf{r}_{rad} is the corresponding radiological distance calculated as the line integral of the density extracted from the CT associated with the study.

4. Combination of doses Do, D1 e D2 into a dose D defined as:
D[x,y,z] = max (D1 [x,y,z], D2 [x,y,z], D3 [x,y,z])



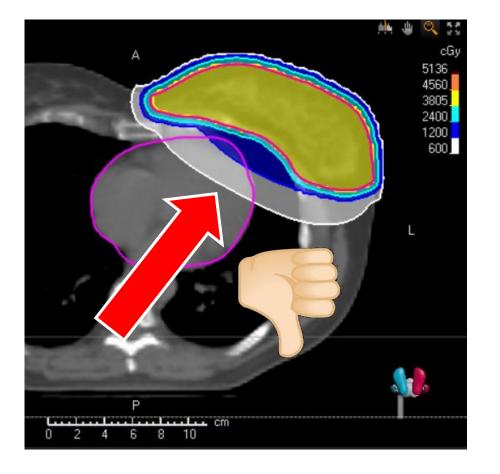




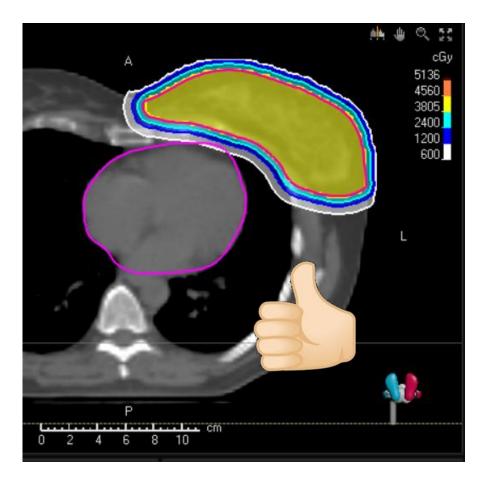


Benchmark dose

Standard parameters



Optimized parameters





Physica Medica 123 (2024) 103394



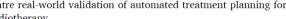
Contents lists available at ScienceDirect

Physica Medica journal homepage: www.elsevier.com/locate/ejmp



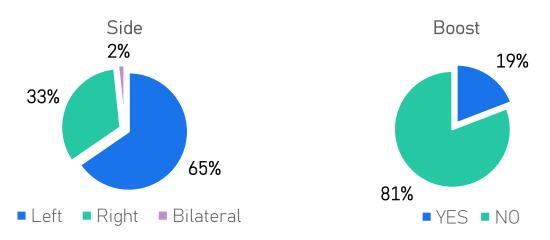
Original paper

Multi-centre real-world validation of automated treatment planning for breast radiotherapy

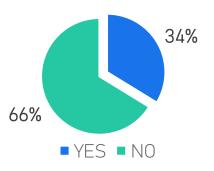


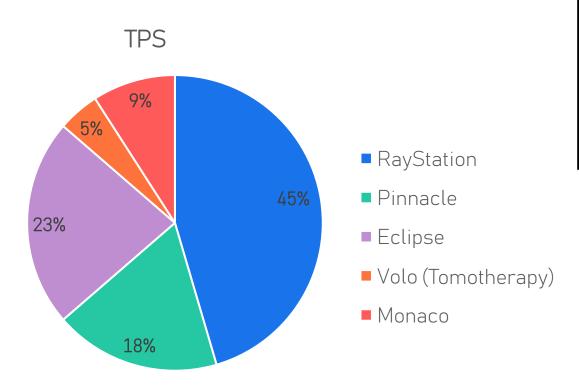
- C. Fiandra a,*, S. Zara b, V. Richetto C, L. Rossi d, M.G. Leonardi e, P. Ferrari f, M. Marrocco S,
- E. Gino h, S. Cora , G. Loi J, F. Rosica k, S. Ren Kaiser , E. Verdolino M, L. Strigari N, Romeo ,
- L. Placidi^p, S. Comi^q, G. De Otto^r, A. Roggio^s, A. Di Dio^c, L. Reversi^t, E. Pierpaoli^u,
- E. Infusino^v, E. Coeli^w, T. Licciardello^x, A. Ciarmatori^y, R. Caivano^z, A. Poggiu^{aa},
- N. Ciscognetti bb, U. Ricardi a, B. Heijmen d

- 24 centres
- 10 patients / centre
- 240 patients



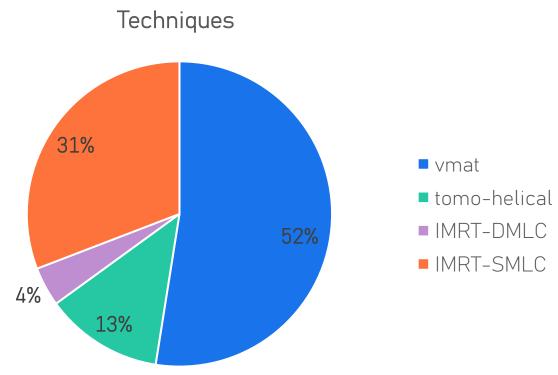
Supraclavicular LNF



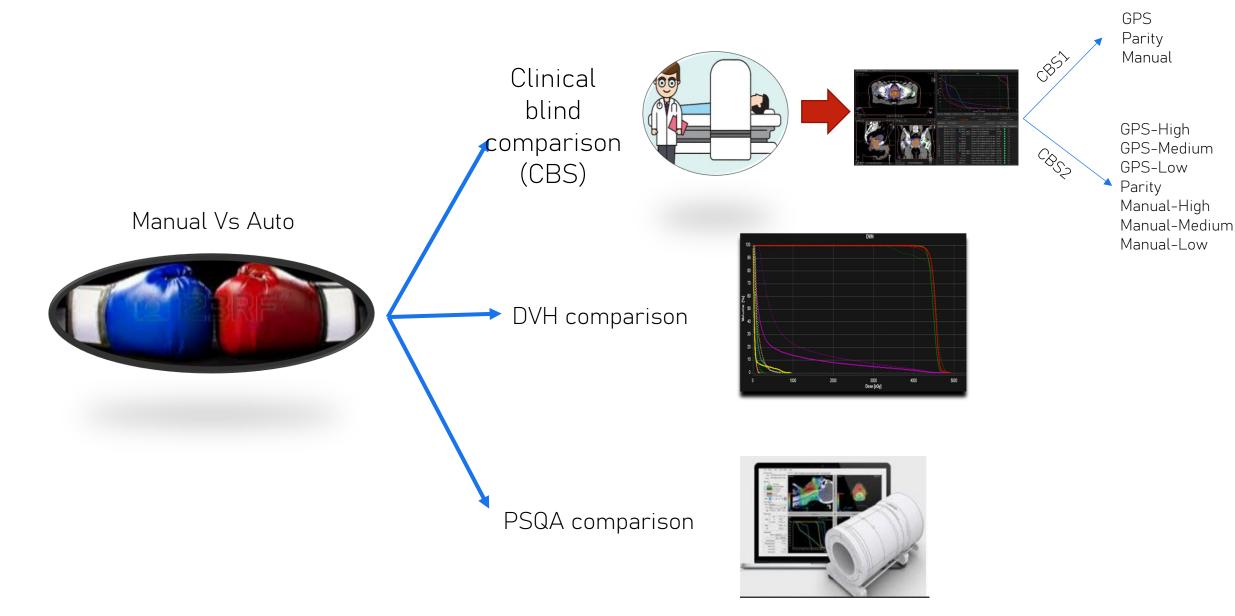




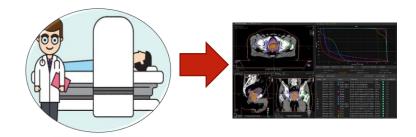
- 24 centres
- 10 patients / centre
- 240 patients



Study design



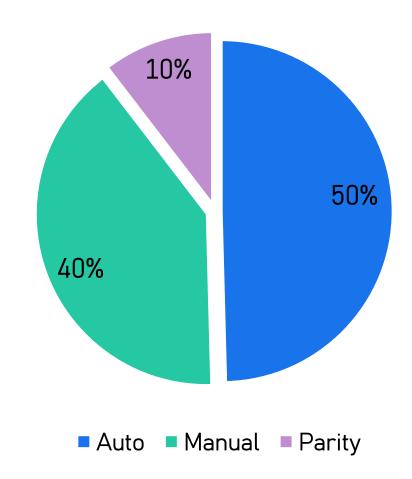
Results: CBS1



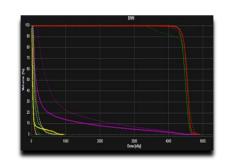
Clinical preferences

All plans were clinically acceptable

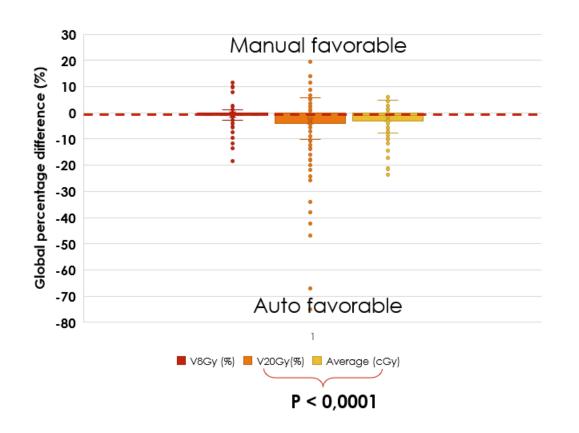
No centre specific configuration



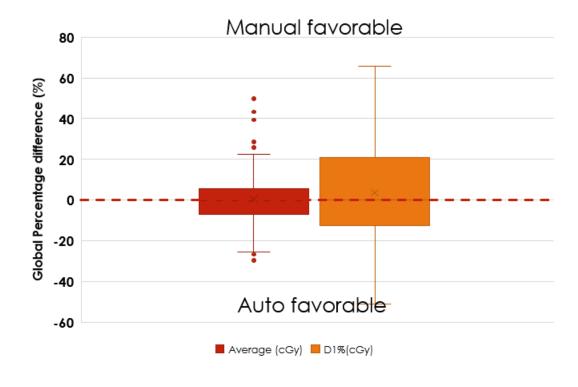
Results: DVH comparison OARs



Heart

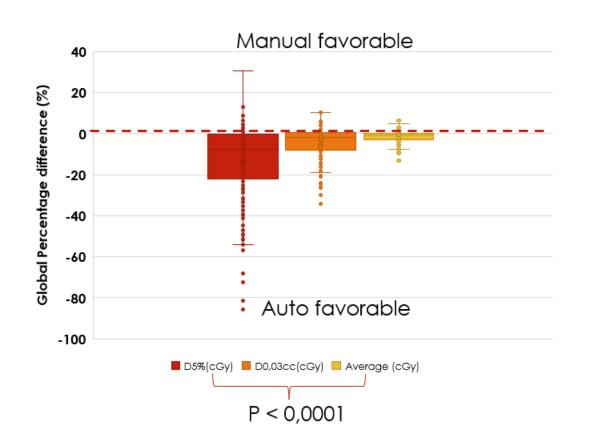


Left anterior descending coronary artery

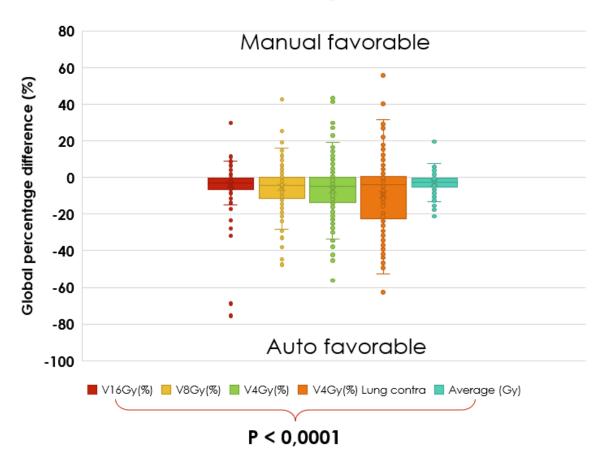


Results: DVH comparison OARs

Contralateral breast

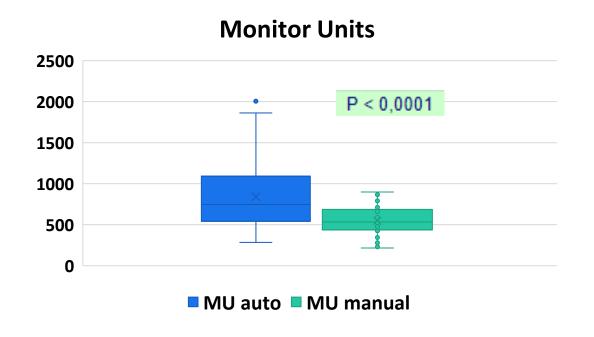


Lungs

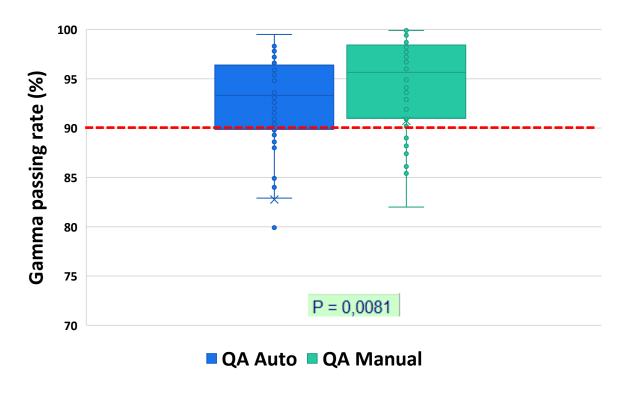


Results: PSQA comparison

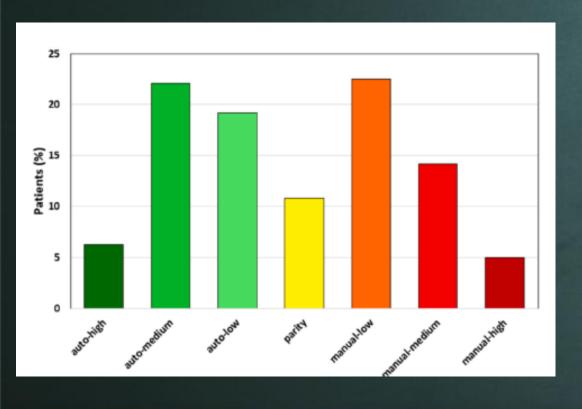


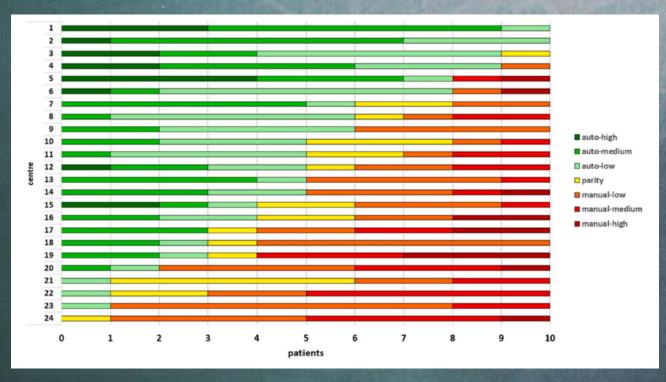






GPS in clinical practice - breast





Conclusion multicentre breast

<u>Purpose:</u> Evaluate a multicentre protocol for breast cancer radiotherapy to ensure reproducibility and plan quality across institutions.

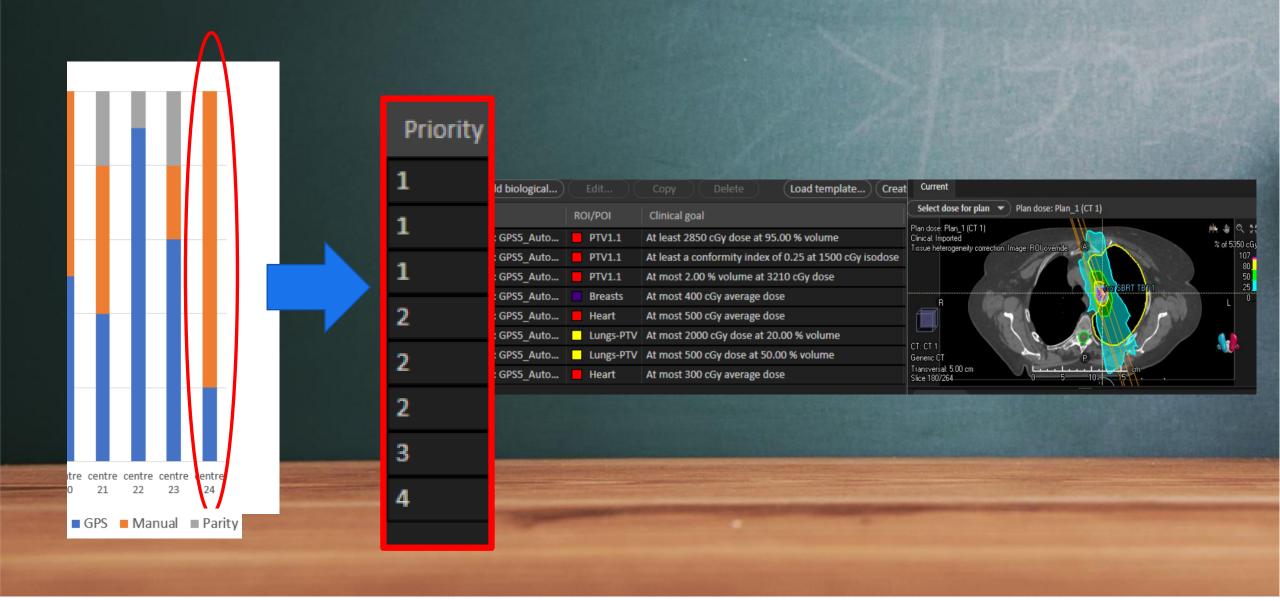
<u>Main Findings:</u> Good consistency in plan quality among centres despite different TPS and planners. Dose constraints for target coverage and OAR sparing were largely met. Variability observed in some dosimetric parameters, but all within clinically acceptable ranges. Inter-centre Variability: Small differences attributed to planner experience, contouring variability, and TPS optimization strategies demonstrates feasibility of harmonizing breast cancer RT planning across centres.

<u>Strengths:</u> Multicentre design provides strong evidence for protocol reproducibility. Highlights value of shared guidelines and centralized review for quality assurance.

<u>Limitations:</u> limited number of patients per centre may not capture all sources of variability. Further standardization of contouring and plan optimization may further reduce variability.

<u>Implications:</u> Results support implementation of shared planning protocols for clinical trials and cooperative studies. Central review and feedback loops are important to maintain consistency over time.

GPS in clinical practice - SBRT



GPS in clinical practice - SBRT





Manual Planning

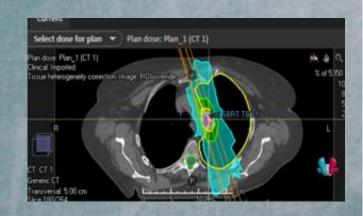
20 lung SBRT cases (10 peripheral: 60 Gy/5 fx, 10 central: 60 Gy/8 fx). VMAT (2–4 coplanar arcs, 6 MV FFF), Eclipse v15.6. D2% \leq 72 Gy, D98% \geq 54 Gy, ITV Dmean \geq 65 Gy.

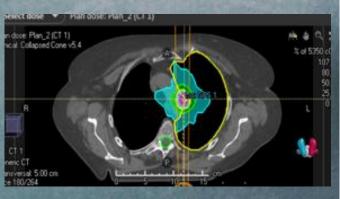
AAA algorithm, 1.25 mm grid.

Automatic Planning (AP)

Guided Planning System (GPS) in RayStation v12A.

Same machine, arcs, isocenter as MP; CCC algorithm, 1.2 mm grid.



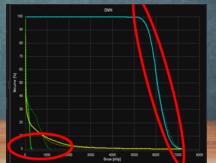


Quality Assurance

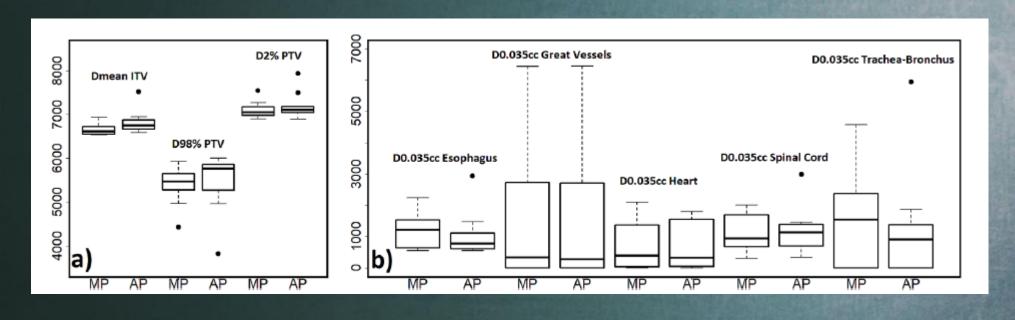
SRS MapCHECK + StereoPHAN phantom, 2%/2 mm gamma (≥ 90% pass).

Plan Comparison

DVH metrics, clinician blind review, gamma pass rate, MUs.
Plan quality metrics: Conformation Number, Homogeneity Index, Gradient Index.



GPS in clinical practice - SBRT



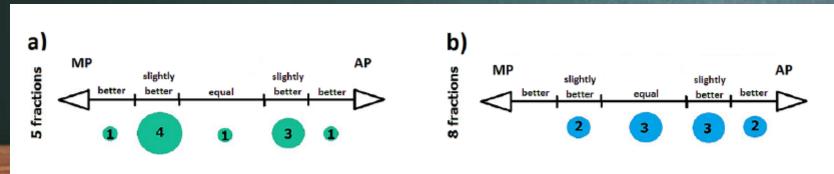


Fig. 3. Clinician's blind comparison results for (a) 5 and (b) 8 fractions scheme.

Conclusion GPS - SBRT

- Automatic planning (AP) matched or outperformed manual planning (MP) in plan quality and consistency for lung SBRT.
- Clinician review confirmed AP plans were clinically acceptable in most cases (equal or better in 65%).
- AP reduces inter-planner variability and saves time, serving as a strong starting point for planning. AP plans still **require** review and possible refinement before delivery.

 The center now uses GPS autoplanning routinely for lung SBRT and is validating it for other tumor sites

Automation in RaySearch Summary

- GPS may be considered as an advanced starting point for planning
 - reduces inter-planner variability, saves time, harmonize optimization strategy serving as a strong starting point but require review and possible refinement before delivery

GPS autoplanning routinely used in 30 centres across Italian country





