



Radiotherapy Tumor Control Cumulative Probability 3d Integral Equation with Tcp Improved Model

Francisco Casesnoves

PhD Engineering, MSc Physics-Mathematics, Physician. Independent Research Scientist. International Association of Advanced Materials, Sweden. UniScience Global Scientific Member, Wyoming, USA.

ARTICLE INFO	ABSTRACT
Published Online: 25 May 2023	A triple-variable Integral Equation for Tumor Control Cumulative Probability (TCCP) is developed with the addition of the important radiotherapy Treatment-Time parameter. Based on a previous double TCCP Integral Equation depending on Biological Model parameters (α , β), the $N_{\text{Effective}}$ model is implemented in Tumor Control Probability (TCP) equation. Results comprise a TCCP Integral Equation model with three essential parameters for Radiation Therapy precise determination of TCCP. Solution shows the method to obtain a 3D Integral Equation of first kind. Results applications could involve better optimization for TCCP in radiotherapy biological models. Radiotherapy Treatment Planning Optimization (TPO) and Radioprotection consequences are explained.
Corresponding Author: DR F CASESNOVES PHD	
KEYWORDS: Pareto-Multiobjective Optimization (PMO), Mathematical Methods (MM), Biological Models (BM), Radiation Therapy (RT), Initial Tumor Clonogenes Number Population (N_0), Effective Tumor Population Clonogenes Number ($N_{\text{Effective}}$), Linear Quadratic Model (LQM), Integral Equation (IE), Tumor Control Probability (TCP), Normal Tissue Complications Probability (NTCP), Biological Effective model (BED), Tumor Control Cumulative Probability (TCCP), Radiation Photon-Dose (RPD), Nonlinear Optimization, Radiotherapy Treatment Planning Optimization (TPO), Software Engineering Methods, Radiation Photon-Dose.	

I. INTRODUCTION

Previously, [20,89], the TCCP Integral Equation was developed and determined with analytical result based on Erf functions. The model set for that double integral, [20,89], based on standard Biological Model parameters (α , β), is improved with the implementation of a variant of the $N_{\text{Effective}}$ model with Treatment-Time variable [90]. That is, a developed modification from [90], based on [20,24,25,83,88,89].

Therefore, the innovation of this study is the Integral TCCP Equation set on with essential TCP Biological Models parameters. In other words, TCCP depending on tumor radiobiological parameters (α , β), Treatment-Time variable, $T_{K(\text{delay})}$, and $T_{\text{Potential}}$ ones. In routinary RT treatment schedule, the usual approximate 30 days for hyperfractionation dose-delivery varies due to unexpected delays, such as weekends, patient circumstances, side effects of radiation etc. In Hyperfractionated RT treatment schedule

times could also adjust for several circumstances. Therefore, the practical utility of this 3D Integral Equation TCCP model is to optimize RT treatment for the interval of $T_{\text{Treatment}}$ time. TPO with biological models could get improvements from the calculations presented.

Consequently, [20,89,90], a more explicit integral equation to improve the radiotherapy treatment dose delivery schedule towards increased patient survival expectation is got. In short, a 3D TCCP Integral Equation model is set for upcoming analytic/numerical determination for improvements in Biomodels RT treatment. Applications for radiotherapy TPO RT are explained in brief.

II. MATHEMATICAL METHOD

The initial probability mathematical algorithm, [20,24,25,83,88,89], to be set in the model is the exponential Linear Quadratic Model for $N_{\text{Effective}}$ clonogenes and the TCP basic statistical ones [22-26,82,83]. Instead, the

“Radiotherapy Tumor Control Cumulative Probability 3d Integral Equation with Tcp Improved Model”

Poisson TCP model, an approximation with Binomial TCP distribution was proven in [20,89]. In this section, cumulative probability mathematical concepts are also detailed in brief.

This study explains further 3D algebraic model variations that are set into the 3D Gaussian convolution integral equation for TCCP cumulative prediction. $N_{\text{Effective}}$ clonogenes model reads,

$$N_{\text{EFFECTIVE}} = N_0 \times 2^{\left[\frac{t-T_K}{T_P} \right]};$$

(1)

Where,

$N_{\text{Effective}}$: Effective N_0 surviving number of tumor clonogens.

N_0 : Surviving number of tumor clonogens .

t: Total radiation treatment parameter.

T_K : time after the start of the first fraction when clonogen proliferation starts parameter (in other publications T_{Delay}).

T_P : Average cell doubling time parameter.

From model [90, Equation (2), Chapter 12] and corroboration with other authors, on [20,24,25,83,88,89], it is guessed the following variant [Casesnoves, 2022],

With Binomial approximation

such as $N_0 \cong 1$,

$P(\alpha, \beta, t) = (1 - \dots$

$\dots e^{[-(\alpha D + \beta K D^2) + A]})$);

(2)

where

P: TCP binomial approximation.

N_s : Initial number of tumor clonogens .

N_o : Surviving number of tumor clonogens .

t : Total RT treatment time, (1).

α : Clonogen radiosensitivity parameter

β : Clonogen radiosensitivity parameter

D : Total radiation dose delivered (could be hyperfractionated)

K : Lea-Catcheside function-factor K, [64]

Before implementing within integral equation, it is defined Factor A as follows,

The A factor,

$$A = \text{Log}(2) \times \left(\frac{t - T_K}{T_P} \right)$$

such as with binomial approximation, $N_0 \cong 1$,

$P(\alpha, \beta, t) = (1 - \dots$

$\dots e^{[-(\alpha D + \beta K D^2) + A]})$);

(3)

where

t: Total radiation treatment parameter.

T_K : time after the start of the first fraction when clonogen

proliferation starts parameter.

T_P : Average cell doubling time parameter.

The integral-algebraic method can be done to get an integral equation of first kind that could be resolved analytically or numerically [89]. The results may constitute an improvement for BED models in general, TCP, TCCP, and NTCP.

The mathematical concept of cumulative probability integral convolution

The cumulative probability is an statistical mathematical concept that involves two convoluted probability functions. Namely, an statistical distribution, frequently a Gaussian, and the given specific probability function subject to the particular problem. The convolution has two kinds of parameters, the probability function itself ones, and the statistical distribution with all types of parameters, those that belong to the probability function and the ones that will give the explicit variables of the integral equation solution. Then, solutions got to get going towards an analytic and/or numerical result.

At the integrand, therefore, there is an intersection of probabilities. That is, the probability of the problem function with the probability of the Gaussian distribution. Because of the integral bounds, that intersection is performed for all the selected integral intervals. In plain language, any integral is a summatory in **R** or **C** spaces. This probability functions intersection within the integral give the cumulative total probability of the problem as follows,

Cumulative Probability Integral Equation... =

$$\dots = \int \text{Prob}(\text{specific}) \cap \text{Prob}(\text{Gaussian}) \dots$$

$$\dots \int \text{Total Prob}(\text{specific function subject to} \dots$$

$$\dots \text{Gaussian distribution}) = \dots$$

$$\dots = \text{Total Prob explicit function} \dots$$

$$\dots \text{with its proper parameters};$$

(Algorithm 1)

III. 3D INTEGRAL EQUATION DEVELOPMENT RESULTS

TCCP can be calculated with a 3D standard Gaussian convolution [20,89]. Algebraic changes can be set as in [20,89] . Therefore, the integral equation for TCCP results,

Binomial Approx,
 such as $N_0 \cong 1$,
 $P(\alpha, \beta, t) = (1 - \dots$
 $\dots e^{[-(\alpha D + \beta K D^2) + A]});$
 $TCCP(\bar{\alpha}, \bar{\beta}, \bar{t}) = \int_{t_1}^{t_2} \int_{\beta_1}^{\beta_2} \int_{\alpha_1}^{\alpha_2} \frac{1}{2\pi\sigma^2} \dots$
 $\dots [1 - P(\alpha, \beta, t)] \times \dots$
 $\dots \times e^{\left[\frac{-1}{2\sigma^2} [(t-\bar{t})^2 + (\beta-\bar{\beta})^2 + (\alpha-\bar{\alpha})^2] \right]} \dots$
 $\dots d\alpha d\beta dt ;$
 with,
 $\sigma = \sqrt{\sigma_\alpha^2 + \sigma_\beta^2 + \sigma_t^2} ;$

(4, Enhanced in Appendix)

where

N_0 : Surviving number of tumor clonogens .

t : Total radiation treatment parameter.

α : Clonogen radiosensitivity parameter.

β : Clonogen radiosensitivity parameter.

D : Total radiation dose delivered.

K : Lea-Catcheside function-factor K, [64].

A : Factor for $T_{Treatment}$ from (2). Adimensional.

σ_α : Standard Deviation for α .

σ_β : Standard Deviation for β .

σ_t : Standard Deviation for $T_{Treatment}$.

σ : Overall normalized Standard Deviation.

This integral equation [Casesnoves, 2023], based on a variation from authors in [90, Equation (2), Chapter 12], can be determined analytically with similar algebraic method than [89] .

IV. RADIOTHERAPY PHYSICS APPLICATIONS

Table 1 developed/modified from previous publications, shows a resume of radiotherapy 3D Integral Equation applications for RT treatment based on biological models. Medical physics principal applications for radiotherapy TPO are explained briefly.

Table 1.- Brief of radiotherapy and radioprotection applications derived for Equations 1-4..

RADIOTHERAPY OPTIMIZATION APPLICATIONS FOR 3D TCCP INTEGRAL EQUATION MODEL		
APPLICATION	FIELD	ADDITIONAL
Biological Models TCP TCCP Improvements	Patient Treatment Precision	More Quality Life and Radioprotection
Post-RT Treatment Survival time	Optimization Time Schedule	Increase of Survival Time
Biological Models Research	Improvements	Improvements LINAC Software And Imaging guided TR Treatment
NTCP Models	Possible applications also	Decrease of Side-Effects at OARs

V. DISCUSSION AND CONCLUSIONS

The objective of the study was to extend the previous TCCP 2D Integral Equation to a 3D further one with the implementation of $N_{Effective}$ model parameters. The resulting three-variables integral equation of first kind was set with a Gaussian convolution to determine the statistical cumulative probability as in [20,89].

The formula developments, Equations 1-4, are also based on a binomial approximation for TCP probability function of the improved model. Algorithm, therefore, is set for next determination as was done in [89]. Mathematical probability concepts for cumulative convoluted probability are explained, Section II. An inconvenient of the equation is the increased number of variables from [89], two, to three ones. Advantages could be more precision and adaptation for a better biological model optimization.

Grosso modo, an advanced formulation for TCCP integral equation was set for further calculations in TPO. Applications for optimal RT planning and increased patient survival time and radioprotection emerge form results.

REFERENCES

1. Casesnoves F (2022). Radiotherapy Wedge Filter AAA Model 18 Mev- Dose Delivery 3D Simulations with Several Software Systems for Medical Physics Applications. Applications. Biomed J Sci & Tech Res 40(5). DOI: 10.26717/BJSTR.2022.46.007337.
2. Casesnoves F (2016) . Mathematical Exact 3D Integral Equation Determination for Radiotherapy Wedge Filter Convolution Factor with Algorithms and Numerical Simulations. Journal of Numerical Analysis and Applied Mathematics 1(2): 39-59. ISSN Online: 2381-7704.
3. Casesnoves F (2015) . Radiotherapy Conformal Wedge Computational Simulations, Optimization Algorithms, and Exact Limit Angle Approach. International Journal of Scientific Research in Science, Engineering and Technology (IJSRSET) 1(2): 353-362. Print ISSN : 2395-1990. Online ISSN : 2394-4099.
4. Casesnoves F (2019) . Improvements in Simulations for Radiotherapy Wedge Filter dose and AAA-Convolution Factor Algorithms. International Journal of Scientific Research in Science, Engineering and Technology (IJSRSET) 6(4): 194-219. Print ISSN: 2395-1990 . Online ISSN : 2394-4099.
5. Casesnoves F (2011) . Exact/Approximated Geometrical Determinations of IMRT Photon Pencil-Beam Path Through Alloy Static Wedges in Radiotherapy Using Anisotropic Analytic Algorithm (AAA). Peer-reviewed ASME

“Radiotherapy Tumor Control Cumulative Probability 3d Integral Equation with Tcp Improved Model”

- Conference Paper. ASME 2011 International Mechanical Eng Congress. Denver. USA. IMECE2011-65435.
6. Casesnoves F (2012) . Geometrical Determinations of Limit angle (LA) related to maximum Pencil-Beam Divergence Angle in Radiotherapy Wedges. Peer-reviewed ASME Conference Paper. ASME 2012 International Mechanical Eng Congress. Houston. USA. IMECE2012-86638.
 7. Casesnoves F (2013) . A Conformal Radiotherapy Wedge Filter Design. Computational and Mathematical Model/Simulation’ . Peer-Reviewed Poster IEEE (Institute for Electrical and Electronics Engineers), Northeast Bioengineering Conference. Syracuse New York, USA. April 6th, 2013. Peer-Reviewed Poster Session on 6th April 2013. Sessions 1 and 3 with Poster Number 35. Page 15 of Conference Booklet Printed.
 8. Casesnoves F (2014) . Mathematical and Geometrical Formulation/Analysis for Beam Limit Divergence Angle in Radiotherapy Wedges. Peer-Reviewed International Engineering Article. International Journal of Engineering and Innovative Technology (IJEIT) . 3(7). ISSN: 2277-3754 . ISO 9001:2008 Certified.
 9. Casesnoves F (2014) . Geometrical determinations of IMRT photon pencil-beam path in radiotherapy wedges and limit divergence angle with the Anisotropic Analytic Algorithm (AAA) Casesnoves, F. Peer- Reviewed scientific paper, both Print and online. International Journal of Cancer Therapy and Oncology 2 (3): 02031. DOI:10.14319/IJCTO.0203.1. Corpus ID: 460308.
 10. Casesnoves F (2014) . Radiotherapy Conformal Wedge Computational Simulations and Nonlinear Optimization Algorithms. Peer-reviewed Article, Special Double-Blind Peer-reviewed paper by International Scientific Board with contributed talk. Official Proceedings of Bio- and Medical Informatics and Cybernetics: BMIC 2014 in the context of the 18th Multi-conference on Systemics, Cybernetics and Informatics: WMSCI 2014 July 15 - 18, 2014, Orlando, Florida, USA. ISBN: 978-1-941763-03-2 (Collection). ISBN: 978-1-941763-10-0 (Volume II) .
 11. Casesnoves F (2007) . Large-Scale Matlab Optimization Toolbox (MOT) Computing Methods in Radiotherapy Inverse Treatment Planning’. High Performance Computing Meeting. Nottingham University. Conference Poster.
 12. Casesnoves F (2008) . A Computational Radiotherapy Optimization Method for Inverse Planning with Static Wedges. High Performance Computing Conference. Nottingham University. Conference Poster.
 13. Casesnoves F (2015) . Radiotherapy Conformal Wedge Computational Simulations, Optimization Algorithms, and Exact Limit Angle Approach. International Journal of Scientific Research in Science, Engineering and Technology 1(2). Print ISSN : 2395-1990, Online ISSN : 2394-4099.
 14. Casesnoves F (2015) . Radiotherapy Standard/Conformal Wedge IMRT-Beamlet Divergence Angle Limit Exact Method, Mathematical Formulation, and Bioengineering Applications. International Article-Poster. Published in Proceedings of Conference. 41st Annual Northeast Bioengineering Conference. Rensselaer Polytechnic Institute. Troy, New York USA, April, p. 17-19. DOI:10.1109/NEBEC.2015.7117152 . Corpus ID: 30285689.
 15. Casesnoves F (2015) . Radiotherapy Standard/Conformal Wedge IMRT-Beamlet Divergence Angle Limit Exact Method, Mathematical Formulation, and Bioengineering Applications. IEEE (Institute for Electrical and Electronics Engineers), International Article-Poster. <http://ieeexplore.ieee.org/stamp/stamp.jsp?tp=&arnumber=7117152>.
 16. Casesnoves F (2015) . Abstract-Journal. ‘Radiotherapy Standard/ Conformal Wedge IMRT-Beamlet Divergence Angle Limit Exact Method, Mathematical Formulation. International Conference on Significant Advances in Biomedical Engineering. 252nd OMICS International Conference 5(1). Francisco Casesnoves, J Bioengineer & Biomedical Sci 2015, 5:1. <http://dx.doi.org/10.4172/2155-9538.S1.003> .
 17. Casesnoves, F (2001) . Determination of absorbed doses in common radio diagnostic explorations. 5th National Meeting of Medical Physics. Madrid, Spain. September 1985. treatment Planning’.
 18. Casesnoves, F (2001). Master Thesis in Medical Physics. Eastern Finland University. Radiotherapy Department of Kuopio University Hospital and Radiotherapy Physics Grouversity-Kuopio. Defense approved in 2001. Library of Eastern finland University. Finland.
 19. Casesnoves F (2013). A Conformal Radiotherapy Wedge Filter Design. Computational and Mathematical Model/Simulation’. Peer-Reviewed Poster IEEE (Institute for Electrical and Electronics Engineers), Northeast Bioengineering Conference. Syracuse New York, USA. Presented in the Peer-

“Radiotherapy Tumor Control Cumulative Probability 3d Integral Equation with Tcp Improved Model”

- Reviewed Poster Session on 6th April 2013. Sessions 1 and 3 with Poster Number 35. Page 15 of Conference Booklet. April 6th, 2013.
20. Casesnoves F (2022). Radiotherapy Biological Tumor Control Probability Integral Equation Model with Analytic Determination. *International Journal of Mathematics and Computer Research* 10(8): 2840-2846. DOI: <https://doi.org/10.47191/ijmcr/v10i10.01>.
 21. Casesnoves F (2022). Radiotherapy Wedge Filter AAA Model 3D Simulations For 18 Mev 5 cm-Depth Dose with Medical Physics Applications”, *International Journal of Scientific Research in Computer Science, Engineering and Information Technology (IJSRCSEIT)* 8(1): 261-274. ISSN: 2456-3307 (www.ijsrcseit.com). DOI: <https://doi.org/10.32628/CSEIT228141>.
 22. Walsh S (2011) . Radiobiological modelling in Radiation Oncology. PhD Thesis. School of Physics. National University of Galway. <http://hdl.handle.net/10379/3027> .
 23. Chapman D, Nahum, A (2015) . Radiotherapy Treatment Planning, Linear- Quadratic Radiobiology. CRC Press. ISBN 9780367866433 .
 24. Mayles, W, Nahum A (2015). Rosenwald, J. Editors. Handbook of Radiotherapy Physics. Second Edition. CRC Press. ISBN 9780367192075 . International Standard Book Number-13: 978-1-4987-2146-2 .
 25. Nahum, A, Webb, S (1993) . A model for calculating tumour control probability in radiotherapy including the effects of inhomogeneous distributions of dose and clonogenic cell density. *Physics in Medicine and Biology*; v. 38(6); p. 653-666 . ISSN 0031-9155 .
 26. Haydaroglu, A, Ozyigit G (2013) . Principles and Practice of Modern Radiotherapy Techniques in Breast Cancer. Springer. DOI:10.1007/978-1-4614-5116-7 .
 27. Casesnoves, F (2019-20) . Die numerische Reuleaux-Methode Rechnerische und dynamische Grundlagen mit Anwendungen (Erster Teil). ISBN-13 : 978-620-0-89560-8, ISBN-10: 6200895600. Publishing House: Scienza Scripts. 2019-20.
 28. Ulmer W, Harder, D (1997) . Corrected Tables of the Area Integral I(z) for the Triple Gaussian Pencil Beam Model. *Z Med Phys* 7: 192-193. DOI: [https://doi.org/10.1016/S0939-3889\(15\)70255-2](https://doi.org/10.1016/S0939-3889(15)70255-2).
 29. Ulmer W, Harder, D (1995) A triple Gaussian pencil beam model for photon beam treatment planning. *Med. Phys* 5: 25-30. DOI :10.1016/S0939-3889(15)70758-0.
 30. Ulmer W, Harder D (1996). Applications of a triple Gaussian pencil beam model for photon beam treatment planning. *Med Phys* 6 : 68-74. [https://doi.org/10.1016/S0939-3889\(15\)70784-1](https://doi.org/10.1016/S0939-3889(15)70784-1).
 31. Ma, C, Lomax, T (2013). Proton and Carbon Ion Therapy. CRC Press. DOI: <https://doi.org/10.1201/b13070>.
 32. Censor, Y, Zenios, S (1997). Parallel Optimization: Theory, Algorithms and Applications’. UOP. DOI:10.12694/SCPE.V3I4.207. Corpus ID: 19584334.
 33. Ulmer, W, Pyyry, J, Kaissl, W (2005). A 3D photon superposition/ convolution algorithm and its foundation on results of Monte Carlo calculations. *Phys Med Biol*, p. 50. DOI: 10.1088/0031-9155/50/8/010.
 34. Ulmer, W, Harder, D (1997). Applications of the triple Gaussian Photon Pencil Beam Model to irregular Fields, dynamical Collimators and circular Fields. *Phys Med Biol*. DOI: <https://doi.org/10.1023/B:JORA.0000015192.56164.a5>.
 35. Haddad K, Anjak O, Yousef B (2019). Neutron and high energy photon fluence estimation in CLINAC using gold activation foils. *Reports of practical oncology and radiotherapy* 24: 41-46. DOI: 10.1016/j.rpor.2018.08.009.
 36. Sievinen J, Waldemar U, Kaissl W. AAA Photon Dose Calculation Model in Eclipse™. Varian Medical Systems Report. Rad #7170A.
 37. Vagena E, Stoulos S, Manolopoulou M (2016). GEANT4 Simulations on Medical LINAC operation at 18MV: experimental validation based on activation foils. *Radiation Physics and Chemistry*. DOI: 10.1016/j.radphyschem.2015.11.030.
 38. Ethics for Researchers (2013). EU Commission. Directorate-General for Research and Innovation. Science in society/Capacities FP7. <https://data.europa.eu/doi/10.2777/7491>.
 39. Casesnoves F (1981). Surgical Pathology I course class notes and clinical practice of Surgical Pathology Madrid Clinical Hospital [Professor Surgeon Dr Santiago Tamames Escobar]. 4th academic year course for graduation in Medicine and Surgery. Lessons and practice Breast Cancer Surgical and Medical Treatment. 1980-1981. Madrid Complutense University.
 40. Tamames Escobar, S (2000). Cirugia/ Surgery: Aparato Digestivo. Aparato Circulatorio. Aparato Respiratorio/ Digestive System. Circulatory System. Respiratory System (Spanish Edition). ISBN 10: 8479034955. ISBN 13: 9788479034955.

41. Formenti, S; Sandra Demaria, S (2013). Combining Radiotherapy and Cancer Immunotherapy: A Paradigm Shift Silvia C. Formenti, Sandra Demaria. *J Natl Cancer Inst* 105: 256-265. DOI: 10.1093/jnci/djs629.
42. Numrich R, (2010). The computational energy spectrum of a program as it executes. *Journal of Supercomputing* 52. DOI:10.1007/s11227-009-0273-x.
43. European Commission, Directorate-General for Research (2021). Unit L3. Governance and Ethics. European Research Area. Science and Society.
44. ALLEA (2017). The European Code of Conduct for Research Integrity, Revised Edn.; ALLEA: Berlin Barndenburg Academy of Sciences.
45. Good Research Practice (2017) Swedish Research Council. ISBN 978-91- 7307-354-7.
46. Ulmer W, Schaffner, B (2011). Foundation of an analytical proton beamlet model for inclusion in a general proton dose calculation system. *Radiation Physics and Chemistry* 80: 378-389. DOI: 10.1016/j.radphyschem.2010.10.006.
47. Sharma, S (2008). Beam Modification Devices in Radiotherapy. Lecture at Radiotherapy Department, PGIMER. India.
48. Barrett, A, Colls (2009). Practical Radiotherapy Planning. Fourth Edition. Hodder Arnold. ISBN 9780340927731.
49. Ahnesjö A, Saxner M, A Trepp (1992). A pencil beam model for photon dose calculations. *Med Phys*, pp. 263- 273. DOI:10.1118/1.596856.
50. Brahime A (2000). Development of Radiation Therapy Optimization. *Acta Oncologica* 39(5). DOI: 10.1080/028418600750013267.
51. Bortfeld T, Hong T, Craft D, Carlsson F (2008) . Multicriteria Optimization in Intensity-Modulated Radiation Therapy Treatment Planning for Locally Advanced Cancer of the Pancreatic Head. *International Journal of Radiation Oncology and Biology Physics* 72(4). DOI: 10.1016/j.ijrobp.2008.07.015.
52. Brown, B, and cols (2014) . Clinician-led improvement in cancer care (CLICC) - testing a multifaceted implementation strategy to increase evidence-based prostate cancer care: phased randomised controlled trial - study protocol. *Implementation Science* 9: 64. DOI: <https://doi.org/10.1186/1748-5908-9-64>.
53. Bortifield, T (2006). IMRT: a review and preview. *Phys Med Biol* 51(2006): R363–R379. DOI: 10.1088/0031-9155/51/13/R21.
54. Censor, Y (1996). Mathematical Optimization for the Inverse problem of Intensity-Modulated Radiation Therapy. Laboratory Report, Department of Mathematics, University of Haifa, Israel.
55. Capizzello A, Tsekeris PG, Pakos EE, Papathanasopoulou V, Pitouli EJ (2006). ‘Adjuvant Chemo-Radiotherapy in Patients with Gastric Cancer. *Indian Journal of Cancer* 43(4). ISSN: 019-509X.
56. Tamer Dawod, EM Abdelrazek, Mostafa Elnaggar, Rehab Omar (2014). Dose Validation of Physical Wedged symmetric Fields in Artiste Linear Accelerator. *International Journal of Medical Physics, Clinical Engineering and Radiation Oncology* 3: 201-209. DOI: 10.4236/ijmpcero.2014.34026.
57. Do SY, David A, Bush Jerry D Slater (2010). Comorbidity-Adjusted Survival in Early-Stage Lung Cancer Patients Treated with Hypofractionated Proton Therapy. *Journal of Oncology*. DOI: 10.1155/2010/251208.
58. Ehgott M, Burjony M. (1999). Radiation Therapy Planning by Multicriteria Optimization. Department of Engineering Science. University of Auckland. New Zealand. Conference Paper.
59. Ezzel, G (1996). Genetic and geometric optimization of three-dimensional radiation therapy treatment planning. *Med Phys* 23: 293- 305. DOI: 10.1118/1.597660.
60. Effective Health Care, (2008). Number 13. Comparative Efectiveness of Therapies for Clinically Localized Prostate cancer. Bookshelf ID: NBK554842.
61. Hansen, P (1998). Rank-deficient and discrete ill-posed problems: numerical aspects of linear inversion’. *SIAM monographs on mathematical modelling and computation*. ISBN-13: 978-0898714036.
62. Hashemiparast, S, Fallahgoul H (2011). Modified Gauss quadrature for ill-posed integral transform. *International Journal of Mathematics and Computation* 13(11). ISSN: 0974-570X.
63. Isa, N (2014). Evidence based radiation oncology with existing technology. *Reports of practical oncology and radiotherapy* 19: 259-266. DOI: 10.1016/j.rpor.2013.09.002
64. Johansson KA, Mattsson S, Brahme A, Turesson I (2003) Radiation Therapy Dose Delivery’. *Acta Oncologica* 42(2): 2003. DOI:10.1080/02841860310004922.
65. Khanna P, Blais N, Gaudreau PO, Corrales-Rodriguez L (2016). Immunotherapy Comes of

“Radiotherapy Tumor Control Cumulative Probability 3d Integral Equation with Tcp Improved Model”

- Age in Lung Cancer, *Clinical Lung Cancer*. DOI: 10.1016/j.clcc.2016.06.006.
66. Kufer KH, Hamacher HW, Bortfeld T (2000). A multicriteria optimisation approach for inverse radiotherapy planning. University of Kaiserslautern, Germany. DOI: 10.1007/978-3-642-59758-9_10.
67. Kirsch A (1996). An introduction to the Mathematical Theory of Inverse Problems. Springer Applied Mathematical Sciences. Series E- ISSN2196-968X.
68. Luenberger, D (1989). Linear and Nonlinear Programming (2nd Edn.). Addison-Wesley. ISBN-13: 978-3030854492.
69. Moczko, J, Roszak, A (2006). Application of Mathematical Modeling in Survival Time Prediction for Females with Advanced Cervical cancer treated Radio-chemotherapy. *Computational Methods in science and Technology* 12(2). DOI: 10.12921/cmst.2006.12.02.143-147
70. Ragaz, J, Ivo A Olivotto, John J Spinelli, Norman Phillips, Stewart M Jackson, et al. (2005). Regional Radiation Therapy in Patients with High-risk Breast Cancer Receiving Adjuvant Chemotherapy: 20-Year Results of the Columbia Randomized Trial'. *Journal of National Cancer Institute* 97(2). DOI: 10.1093/jnci/djh297.
71. Steuer R (1986). Multiple Criteria Optimization: Theory, Computation and Application. Wiley. <https://doi.org/10.1002/oca.4660100109>.
72. Spirov SV, Chui CS (1998). A gradient inverse planning algorithm with dose-volume constraints. *Med Phys* 25: 321-323. DOI: 10.1118/1.598202.
73. Das I, and colls (1997). Patterns of dose variability in radiation prescription of breast cancer. *Radiotherapy and Oncology* 44: 83-89. DOI: 10.1016/s0167-8140(97)00054-6
74. Casesnoves, F (2018). Practical Radiotherapy TPO course and practice with Cyberknife. Robotic simulations for breathing movements during radiotherapy treatment. Sigulda Radiotherapy Cyberknife Center. Latvia. Riga National Health Oncology Hospital Varian LINACs TPO practice/lessons several Varian LINACs. Riga Technical University Bioengineering Training-Course Nonlinear Life. August 2018.
75. Casesnoves, F. (2022). Radiotherapy Linear Quadratic Bio Model 3D Wedge Filter Dose Simulations for AAA Photon-Model [18 Mev, Z= 5,15 cm] with Mathematical Method System. *Biomed J Sci & Tech Res* 46(2)-2022. BJSTR. MS.ID.007337. DOI: 10.26717/BJSTR.2022.46.007337.
76. Casesnoves, F (1985). Masters in philosophy Thesis at Medical Physics Department. Protection of the Patient in Routinary Radiological Explorations. Experimental Low Energies RX Dosimetry. Medicine Faculty. Madrid Complutense University. 1984-85.
77. Casesnoves, F (1983-5). Ionization Chamber Low Energies Experimental Measurements for M-640 General Electric RX Tube with Radcheck ionization camera, Radcheck Beam Kilovoltmeter and TLD dosimeters. Radiology Department practice and measurements. Madrid Central Defense Hospital. Medical Physics Department. Masters in philosophy Thesis. Medicine Faculty. Complutense University. Madrid.
78. Casesnoves, F (1985). Determination of Absorbed Doses in Routinary Radiological Explorations. Medical Physics Conference organized by Medical Physics Society Proceedings Printed. San Lorenzo del Escorial. Madrid. September 1985.
79. Greening, J (1985). Fundamentals of Radiation Dosimetry. Taylor and Francis. Second Edition. 1985. DOI: <https://doi.org/10.1201/9780203755198>.
80. International Commission of Radiation Protection (1977). Bulletin 26th. The International Commission on Radiological Protection. Recommendations of the International Commission on Radiological Protection. Pergamon Press. Copyright © 1977 The International Commission on Radiological Protection.
81. Stanton, P; Colls (1996) . Cell kinetics in vivo of human breast cancer. *British Journal of Surgery* 1996,83,98-102. DOI: <https://doi.org/10.1002/bjs.1800830130>.
82. Hedman M, Bjork-Eriksson T, Brodin O, Toma-Dasu I (2013). Predictive value of modelled tumour control probability based on individual measurements of in vitro radiosensitivity and potential doubling time. *Br J Radiol* 2013;86: 20130015. DOI:10.1259/bjr.20130015.
83. Fowler, J. 21 years of Biologically Effective Dose. *The British Journal of Radiology*, 83 (2010), 554–568.
84. Marcu, L, and al (2018). Radiotherapy and Clinical Radiobiology of Head and Neck Cancer. Series in Medical Physics and Biomedical Engineering. CRC Press. 2018.
85. Casesnoves, F (2022). Radiotherapy 3D Isodose Simulations for Wedge Filter 18 Mev-Dose [z = 5,15 cm] with AAA Model with Breast Cancer Applications. *International Journal on Research*

“Radiotherapy Tumor Control Cumulative Probability 3d Integral Equation with Tcp Improved Model”

Methodologies in Physics and Chemistry (IJRPC)
ISSN: 2349-7963 Volume: 9 Issue: 2. 2022.

86. Garden, A; Beadle, B; Gunn, G (2018). Radiotherapy for Head and Neck Cancers. Fifth Edition. Wolters Kluwer. 2018.
87. Casesnoves, F (2023). Radiotherapy Genetic Algorithm Pareto-Multiobjective Optimization of Biological Effective Dose and Clonogens Models for Head and Neck Tumor Advanced Treatment. International Journal of Mathematics and Computer Research. ISSN: 2320-7167. Volume 11 Issue 01 January 2023, Page no. – 3156-3177. DOI: 10.47191/ijmcr/v11i1.08.
88. Casesnoves, F (2023). Radiotherapy effective clonogens model graphical optimization approaching linear quadratic method for head and neck tumors. International Journal of Molecular Biology and Biochemistry. ISSN Print: 2664-6501. ISSN Online: 2664-651X. Impact Factor: RJIF 5.4. IJMBB 2023; 5(1): 33-40.
89. Casesnoves, F (2022). Radiotherapy Complete Mathematical Demonstration for Biological Tumor Control Cumulative Probability Integral Equation Model with Applications. International Journal of Mathematics and Computer Research ISSN: 2320-7167. Volume 10 Issue 10 October 2022, Page no. – 2916-2924. DOI: 10.47191/ijmcr/v10i10.01.
90. Bentzen, S; and alter. (2008). Radiation Oncology Advances. ISBN-13: 978-0387-36743-9. Springer.

VI. SCIENTIFIC ETHIC STANDARDS

Model is a modification from [90] authors, based also on [20,24,25,83,88,89] techniques. RT applications methods for these publications were created by Dr Casesnoves in 2021-2. Methods from [20,89] were created by Dr Francisco Casesnoves in 3rd November 2016, and Interior Optimization Methods in 2019. BED model setting in Algorithms and programming were developed by Dr Casesnoves from previously published BED models. This article has previous papers information, from [1-21], whose inclusion is essential to make the contribution understandable. This study was carried out, and their contents are done according to the International Scientific Community and European Union Technology and Science Ethics [38,43-45]. References [38,43,44,45]: ‘European Textbook on Ethics in Research’. European Commission, Directorate-General for Research. Unit L3. Governance and Ethics. European Research Area. Science and Society. EUR 24452 EN. And based on ‘The European Code of Conduct for Research Integrity’. Revised Edition. ALLEA. 2017. This research was completely done by the author, the computational-software, calculations, images, mathematical propositions and statements, reference citations, and text is original for the author. When a

mathematical statement, algorithm, proposition or theorem is presented, demonstration is always included. When a formula is presented, all parameters are detailed or referred. If any results inconsistency is found after publication, it is clarified in subsequent contributions. When a citation such as [Casesnoves, ‘year’] is set, it is exclusively to clarify intellectual property at current times, without intention to brag. The article is exclusively scientific, without any commercial, institutional, academic, religious, religious-similar, non-scientific theories, personal opinions, political ideas, or economical influences. When anything is taken from a source, it is adequately recognized. Ideas and some text expressions/sentences from previous publications were emphasized due to a clarification aim [38, 43-45].

VII. AUTHOR’S BIOGRAPHY



Dr Francisco Casesnoves earned the Engineering and Natural Sciences PhD by Tallinn University of Technology (started thesis in 2016, thesis Defence/PhD earned in December 2018, official graduate Diploma 2019). He works as independent research scientist in computational-engineering/physics. Dr Casesnoves earned MSc-BSc, Physics/Applied-Mathematics (Public Eastern-Finland-University, MSc Thesis in Radiotherapy Treatment Planning Optimization, which was developed after graduation in a series of Radiation Therapy Optimization-Modelling publications [2007-present]). Dr Casesnoves earned Graduate-with-MPhil, in Medicine and Surgery [1983] (Madrid University Medicine School, MPhil in Radioprotection Low Energies Dosimetry [1985]). He studied always in public-educational institutions, was football player 1972-78 (defender and midfielder) and as Physician, supports healthy life and all sports activities. Casesnoves resigned definitely to his original nationality in 2020 for ideological reasons, democratic-republican ideology, and ethical-professional reasons, and does not belong to Spain Kingdom anymore. His constant service to the International Scientific Community and Estonian technological progress (2016-present) commenced in 1985 with publications in Medical Physics, with further specialization in optimization methods in 1997 at Finland—at the moment approximately 100 recognized

“Radiotherapy Tumor Control Cumulative Probability 3d Integral Equation with Tcp Improved Model”

publications with approximately 75 DOI papers. His main branch is Computational-mathematical Nonlinear/Inverse Methods Optimization. Casesnoves best-achievements are the Numerical Reuleaux Method in dynamics and nonlinear-optimization [books 2019-2023], The series of Radiotherapy Improvements for AAA superposition-convolution model, the Graphical and Interior Optimization Methods [2016-8], the new Computational Dissection-Anatomical Method, [2020], invention of Forensic Robotics [2020-2021], and Molecular Effect Model for High Temperature Superconductors [2020]. Dr Casesnoves scientific service since 2016 to the Free and Independent Republic of Estonia for technological development (and also at Riga technical University, Power Electrical and Electronics Department) is

about 50 physics-engineering articles, three books series, and 1 industrial radiotherapy project associated to Europe Union EIT Health Program (Tartu University, 2017). Recently, a new book in Interior Optimization Mathematical Methods with Electronics Applications was published (2023).

APPENDIX

Binomial Approx,

such as $N_0 \cong 1$,

$$P(\alpha, \beta, t) = (1 - e^{[-(\alpha D + \beta K D^2) + A]}) ;$$

$$\begin{aligned} \text{TCCP}(\bar{\alpha}, \bar{\beta}, \bar{t}) &= \int_{t_1}^{t_2} \int_{\beta_1}^{\beta_2} \int_{\alpha_1}^{\alpha_2} \frac{1}{2\pi\sigma^2} \times [1 - P(\alpha, \beta, t)] \times \dots \\ &\dots \times e^{\left[\frac{-1}{2\sigma^2} \left[(t-\bar{t})^2 + (\beta-\bar{\beta})^2 + (\alpha-\bar{\alpha})^2 \right] \right]} d\alpha d\beta dt ; \end{aligned}$$

with,

$$\sigma = \sqrt{\sigma_\alpha^2 + \sigma_\beta^2 + \sigma_t^2} ;$$

INTEGRAL EQUATION MODEL (4) ENHANCED

[Casesnoves, 2023]