ISSN: 2574 -1241



Radiotherapy Linear Quadratic Bio Model 3D Wedge Filter Dose Simulations for AAA Photon-Model [18 Mev, Z=5,15 cm] with Mathematical Method System

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ARTICLE INFO

Received: September 12, 2022 Published: September 22, 2022

Citation: Francisco Casesnoves. 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.

ABSTRACT

Continuing with Anisotropic Analytic Model (AAA) for Wedge Filters (WF) photon-dose delivery series, this contribution moves ahead with implementation of those AAA studies of WF formulation/simulations into Biological Models-significantly developed by other authors in the literature. Results comprise the WF dose delivery 3D simulations for approximate tumor clonogenes population survival rate. Solutions of previous 3D photon-dose delivery numerical/imaging simulations for AAA model 18 Mev photon-beam, depth-doses [z=5,15 cm], habitually breast tumors depth-dose magnitudes, and 15° WF are set into fundamental Linear Quadratic Bio model equation. Developments of Linear Quadratic Bio models are mathematically/statistically presented—focused on methodology and imaging software, accuracy is not the main objective at this stage. Results for 2D variation of parameter $[\alpha]$ related to N0 values are proven. A new statistical/probabilistic modification for the Linear Quadratic Bio model for software adjustment is presented. Software and programming methods for BM were achieved. Results of 3D Graphical Software for this Bio model implementations are shown. Findings involve the mathematical-computational method explanations to work up and implement a photon-dose model into a final survival-rate biological one. Applications on Clinical Radiotherapy Physics and Radiation Oncology for breast cancer treatment are described in brief.

Abbreviations: MM: Mathematical Methods; BM: Biological Models; LQM: Linear Quadratic Model; IE: Integral Equation; TCP: Tumor Control Probability; TCCP: Tumor Control Cumulative Probability; RPD: Radiation Photon-Dose; Nonlinear Optimization, Series Approximations, Radiotherapy Treatment Planning Optimization; SSD: Source- Surface Distance ; Software Engineering Methods, Radiation Photon-Dose, AEF: Attenuation Exponential Factor; Nonlinear Optimization; WF: Wedge Filter; AAA: Anisotropic Analytic Model; FF: Fluence Factor; OF: Omega Factor; TPO: Treatment Planning Optimization; BT: Breast Tumor

Brief Introduction and Objectives

Following the Radiation Therapy research series, [1-20], this stepping-up study objective is the implementation of AAA extensive previous publications results into Biological Models (BM) TPO parameters. Development for 3D Graphical Optimization solutions from recent study [19] to obtain Tumor Clonogenes Survival Rate NS constitutes the purpose of this research. Photon-beam Biological Models are based mainly on exponential functions with radiobiological parameters, namely [α , β], whose magnitudes are determined in general with in vitro experimental [21-24]. The Photon-beam Model applied for getting these objectives is AAA Model [1-20]. Second practical purpose of the article is to set Photon-dose delivery magnitudes, modified by Wedge Filters (WF), for dosedepths useful in Breast Cancer TPO. That is, [z= 5,15 cm, for 18 MeV] are usual dose delivery depths in breast tumors, modified to conform its shape with WF. Thus, Radiation Dose is implemented within the BM exponential-equation to determine the TPO dose optimized by the proper radiosensitivity characteristics of the tumor [21-24]. In consequence, the photon-dose magnitude can be also used for statistical methods to determine probabilistic parameters such as Tumor Control Probability with distributions like Poisson or Binomial. Hence, TCP can be got though convolution of TCP with Gaussians depending on radiobiological parameters [α , β]. Breast cancer BM parameters have been experimentally determined in the literature because the incidence/prevalence in developed countries of this tumor increase gradually [25]. The research results present two main strands. First is a group of BM new algorithms for NS Rate determination. Secondly 3D Graphical Optimization and Numerical Calculations of NS Rate from these equations at delivery dose depths [z= 5,15 cm, for 18 MeV]. Results magnitudes coincide with other contributions datasets [21-24]. Additionally, a polynomial fit numerical equation graph for variation of parameter $[\alpha]$ related to N0 values [data for software from [23], Table 44.2)] was obtained, Figure 1. Succintly, 3D Graphical Optimization methods with imaging processing and BM based on NS Rate are presented specifically for breast cancer radiotherapy TPO applications.

Mathematical Models and Methods

Section is divided into several sub-sections. In brief, from Bio model formulations and new model [19], to techniques for implementation of photon-dose delivery AAA Model into Linear Quadratic Bio model. The basic AAA WF dose delivery formulas and numerical data for [z=5, 15 cm] depths are reminded from previous publications [refs]. The statistical-probabilistic approximations to develop a Tumor Control Probability Integral Equation Model are explained [19,21-24].

Linear Quadratic Model Mathematical Statistical Model [CAS-ESNOVES, 2022]

Bio models' equations are not very complicated, and based usually on exponential functions, statistical distributions, and two radio sensitivity key parameters, namely [α and β biological modelling parameters]. An Integral Equation Model (IEM), based on new Linear Quadratic Model and Statistical Binomial Distribution approximation was published in recent contributions [19]. The simplest Linear Quadratic model equation set in [19,21-24] reads. Elementary Biological Model for clonogenes survival population $\mathrm{N}_{\mathrm{s}_{\mathrm{s}}}$

$$N_s = N_o \times e^{-[\alpha D + \beta D^2]}$$
.....(1)

Where,

N_s: Initial number of tumors clonogens

 $N_{\ensuremath{\text{o}}\xspace}^{\ensuremath{\text{:}}}$ Surviving number of tumors clonogens α : Clonogen radio-sensitivity parameter

 $\boldsymbol{\beta} :$ Clonogen radiosensitivity parameter D: Total radiation dose delivered

Parameters [α and β] have interrelated each other [β/α] and to N_0 magnitude. Parameter $[\alpha]$ is related to N_0 magnitude [ref], although an average value is generally implemented. Both [α and β] measurements are taken in vitro [19,21-24]. Equation (1) is more complicated in practice for a number of constraints. Firstly, the dose D is hypofractionated in clinical practice [19,21-24]. Secondly, a Lea-Catcheside function-factor K [21], has to be introduced. However, fractional dose factors, d, and number of fractions n, [21-24], are omitted in (1) for simplification, and are not relevant for the mathematical method development. In the standard BM research, [19,21-24], the quotient [σ/β] is generally considered constant. In this study, however, alpha and betha radiosensitive parameters are set independently. The reason is avoiding too many approximations. By using 100% percentages and 1% rates for $N_{o}s$, the model can be better implemented in Statistical Distributions to obtain TCP formulation, [19], with Poisson, Binomial, and Normal-Gaussian distributions. By using this modification, [Casesnoves, 2022], several numerical inconvenient are sorted. This variant, using percentages and rates of N0 implies to make calculations setting parameters $[\alpha, \alpha]$ mainly, and β] values in function of N₀, for example [23, Table 44.2]. Figure 1 shows an original numerical equation from these values [23]. Thus, [19], the modified BM equation reads.

Percentage and rate[% or 1],

$$N_{s}[\%] = N_{0}[100\%] \times e^{-[\alpha D + \beta K D^{2}]};$$

or in rate,
 $N_{s}[1\%] = N_{0}[1\% = 1] \times e^{-[\alpha D + \beta D^{2}]};$(2)
[Casesnoves 2022];

where

N_c: Initial number of tumors clonogens

 $N_{\ensuremath{o}\xspace}^{}$: Surviving number of tumors clonogens α : Clonogen radio-sensitivity parameter

 $\boldsymbol{\beta} :$ Clonogen radiosensitivity parameter D: Total radiation dose delivered

K: Lea-Catcheside function-factor K, [21]

rates for N0 is conditioned for the magnitude changes of [α , β] related N0 [23], Table 44.2], Figure 1.

This approximation, Equation (2), facilitates further calculations and evolved models. However, setting percentages and/or



Figure 1: Numerical equation graph for variation of parameter [α] related to N0 values [data for software from [23], Table 44.2, (Mayles, W ; Nahum, A)]. Model [green] is the straight line, numerical experimental data [red] match almost perfectly the model line. Errors and Residual Norms are almost null. However, it can be considered an approximation because numerical data published is subject to several constraints [23]. The increment of [α] parameter when N0 rises is sharply demonstrated.

Basic Formulation and Data Referred

The large number of basic formulas reminder from [1-20], are described in Table 1 to explain the first implementation into 3D programs to get finally the Survival Rate NS data. Exclusively for illustration, main implemented Photon-Dose Equations from [1-20]. Also from [19,21-24], dose-depth [z=5], numerical data from

[19] appropriate for breast cancer irradiation is shown Tables 2 & 3. Depth Dose Equations for [z=15 cm] can be found at [1-8]. Those all [z=5, 15 cm] cover the approximate TPO range from a small mammary glands size to a bigger one. Essential algorithms reminders are presented in Equations 3,4. The most important reminder algorithms from [1-20], for photon-dose implementation into the BMread.

Table 1: Most AAA references formulation along radiotherapy series studies with foundations and corrections by other authors. All data for [z=5 cm] dose delivery parameters published at [20].

Radiotherapy Wedge Filter AAA Model Photon-Dose Formulation Indicated						
Formulation algorithm	Article Reference	Application	Complementary			
Foundation AAA algorithm for pho- ton- dose in water, integral equation model	Original from Ulmer and Harder [28,29,31,32].	It was the base for AAA model with heterogeneities for TPO	In [29] the Integral Operator and Equation for water dose delivery with WF with analytic solution is shown			
AAA algorithm WF applications for photon-dose in water, integral equation model	Original from Ulmer and Harder [28,29,31,32].	The corrected dose- delivery for AAA with WF is useful for TPO.	Principal integral equation model with compact Gaussian integral operator.			
AAA algorithm applications with tissue heterogeneities for pho- ton-dose, integral equation model Original from Ulmer and Harder and Sievinen [28,29,31,32,34].		Tissue heterogeneities were im- plemented within foundation AAA model.	Principal WF integral equation model with compact Gaussian integral operator.			
2D and 3D Graphical Simulations for AAA WF dose delivery in water [z=15 cm, z=5 cm] .	Extensively published in [1], and [2-8,20]. Presented also in several SIAM conferences.	Useful for 3D WF dose delivery corrected with Omega Factor.	Dataset for 2D/3D/4D simulations.			

2D and 3D Graphical Simulations Computational Method with Matlab, GNU-Octave, and Freemat. For AAA WF dose delivery in water [z=15 cm, z=5 cm].Extensively published in [1], and [2-8,20]. Presented in several SIAN conferences.		Useful for 3D WF dose delivery pro- gramming corrected with Omega Factor.	Applications to software engineer- ing in radiotherapy and model simulations.	
Omega Factor Algorithm and I(z) corrections.	It is at [1,27], and in [1-8,20] the reader can find previous calculation in references series.	For analytic geometry photon-dose correction of the foundation WF algorithm in [29]	Very important I(z) correction for dose delivery precision.	
Delivery dose integral equation with Omega Factor	It is at [1], and formulas with analytic geometry calculations development [2-8].	Omega Factor corrects dose deliv- ery equation appropriately.	Correction proven in papers series.	
Delivery dose integral equation analytic solution with Omega Factor development [2-8,20].		Omega Factor corrects dose delivery equation analytic solution appropriately.	Mathematical Method to set Omega Factor for exact dose.	
Biological Integral Equation model analytic method and parameters [α , β].	At [19] BM, and at [21-24,20] pa- rameters [α, β] magnitudes.	At initial stages, for Tumor Control Probability determination after inserting photon- dose magnitudes.	It is based on convolution of Bino- mial probability with Gaussian	

Table 2: AAA model main Photon-Beam Spectra parameters for 3D Graphical simulations. Dataset is large, it is recommended to consult references [1-8, 20,26-29].

Numerical Dosimetry Dataset						
Radiation parameter	Experimental references and simulation selected data	Approximated magnitude	Aaa model data			
SSD [cm]	Simulation dataset [2-8,20 28,29]	100 cm	[1-8, 20,26-29]			
Fluence Rate [N x cm-2 x s-1] N: number of photons	Φ0 ε Experimental Data with Foil- ers [33,36]	Φ0 ε [4.16 x 109 ,2.82 x 1010]	[1-8, 20,26-29]			
Field Size [cm2] WF Angle	WF Angle=15° 12 x 12 [cm]	WF Angle=15° 12 x 12 [cm]	[1-8, 20,26-29]			
Omega Factor Angles ϕ 1=30° ϕ 2=30° [approximately] WF angle α	[1-8, 20,26-29]	φ1=30° φ2=30°WF Angle=15°	[1-8, 20,26-29]			
Dose Rate [Gy/s]	Dose Rate [Gy/s] According to [33,36]		[1-8, 20,26-29]			
Standard Dose Protocol BreastTumor and parameters [α , β] Forimplementation at LQM		About 40-50 Gy parameters [α = 0.38 Gy-1, β = 0.026 Gy-2] [in vitro experimental, 21-24,37, data variation according to authors not significant in Magnitude order]	[1-8, 20,26-29]			

Table 3: From [72], AAA model main Photon-Beam computational equations parameters for 3D Graphical simulations [z= 5 cm], [1-8,20].

Numerical Method For Computational Programming Implementation (18 Mev,z=5cm) with [Ω]F				
Does summary element 3D (Omega Factor) DΩ[i=1,,3](x,y,z)				
$D\Omega[i=1](x,y,z)=1.1*10-2 \exp(2.3809e-05-0.0782*x)* [Erf[x+12.5994)/0.124795]-Erf[x-12.6066)/0.124795]]* [Erf[(y+12.6)/0.124795]-Er-f[(y-12.6)/0.126795]]$				
$D\Omega[i=2](x,y,z)=0.4847*10-2 \exp(0.0145-0.0782*x)* [Erf[x+12.2282)/3.083817]-Erf[x-12.9718)/3.083817]]* [Erf[(y+12.6)/3.083817]-Erf[(y-12.6)/3.083817]] = 0.4847*10-2 \exp(0.0145-0.0782*x)* [Erf[(x+12.2282)/3.083817]-Erf[x-12.9718)/3.083817]]* [Erf[(y+12.6)/3.083817]-Erf[(y-12.6)/3.083817]] = 0.4847*10-2 \exp(0.0145-0.0782*x)* [Erf[(x+12.2282)/3.083817]-Erf[(x-12.9718)/3.083817]]* [Erf[(y+12.6)/3.083817]-Erf[(y-12.6)/3.083817]] = 0.4847*10-2 \exp(0.0145-0.0782*x)* [Erf[(x+12.2282)/3.083817]-Erf[(x-12.9718)/3.083817]]* [Erf[(y+12.6)/3.083817]] = 0.4847*10-2 \exp(0.0145-0.0782*x)* [Erf[(y-12.6)/3.083817]] = 0.4847*10-2 \exp(0.0145+0.0782*x)* [Erf[(y-12.6)/3.0847*x]] = 0.4847*10-2 \exp(0.0145+0.0782*x)* [Erf[(y-12.6)/3.0847*x]] = 0.4847*10-2 \exp(0.0145*x)* [Erf[(y-12.6)/3.0847*x]] = 0.4847*x]$				
$D\Omega[i=3](x,y,z)=0.3739*10-2 \exp(6.0734e-04-0.0782x)* [Erf[x+12.5845)/0.630286]-Erf[x-12.6155)/0.630286]]* [Erf[(y+12.6)/0.630286]-Erf[(y-12.6)/0.630286]]* [Erf[(y+12.6)/0.630286]]* [Erf[(y+12.6)/0.6302$				
$D\Omega[1=3](x,y,z)=0.3/39^{\circ}10-2 \exp[(6.0/34e-04-0.0/82x]^{\circ} [Err[x+12.5845]/0.630286]-Err[x-12.6155]/0.630286]]^{\circ} [Err[(y+12.6)/0.630286]-Er-f[(y-12.6)/0.630286]]$				

 $D[X,Y,Z] = \frac{I(z)}{4(1+z/F)^2} \times \dots$ $\dots \times \int_{-a}^{a} \int_{-b}^{b} \Phi_{\Omega} \times \dots$ $\dots \times \sum_{K=1}^{K=3} \frac{C_K}{\pi \sigma_K^2} \times e^{[-[(x-u)^2 + (y-v)^2/\sigma_K^2(Z)]]} dudv;$ with mod ified fluence in $[\Omega]_F$, $\Phi_{\Omega}(u,v,z) = \Phi_0(u,v,z) \times \dots$ $\dots \times e^{[-u_w X[L + \frac{Cu}{F+z} \times \left(\frac{\sin \alpha}{\cos(\alpha + \phi)}\right)] \times [\Omega]_F]} \dots \dots (3)$ where I(z) is the area integral of the dose over a plane perpendicular to the z-axis at depth z, normalized to one incident electron, σ (z) is the depth-dependent mean square radial displacement, x, y, z are the coordinates of the dose-delivery point at beam-output coordinates system, u, v, z, are the coordinates of photon-fluence at depth z, a, b, are the field-size magnitudes at depth z, CK are optimization parameters resulting for a triple Gaussian function setting, Φ 0 is the photon fluence source for WF; a'(z) =a(l + z/F), b'(z) =b(1 + z/F) are the half side lengths projected into depth z, with F as

the source-surface distance (SSD). Parameters μ w (WF material parameter), α (WF angle) and φ (photon-beam divergence angle) are defined at [1-8,26-29]. The WF manufacturing angle is α . C is the distance in z-coordinate from source to WF surface, L is half-length of WF for y coordinate. In general, Fluence modification is $\Phi U = \Phi 0$ / $(1 + z/F)^2$ according to [28,29]. Then source fluence is $\Phi 0$ modified primarily for the dose- delivery depth z and secondly by the WF parameters [1-8,26-29]. And the solution, exact, complete, geometrically corrected with Omega Factor, and analytical of this integral equation [Casesnoves, 2015, April, Philadelphia], reads.

$$\begin{split} D(x, y, z) &= \frac{I(z)A\Phi_0}{4(1+z/F)^2} X \sum_{K=1}^{K=3} [\sigma_K^2(z)S^2 - 2SX]...X \\ ...X[[Erf\left(\frac{y+b'}{\sigma_K(z)}\right)] - [Erf\left(\frac{y+b'}{\sigma_K(z)}\right)]]...X \\ ...X[[Erf\left(\frac{x+a'+\sigma_K^2(z)S^2}{\sigma_K(z)}\right)] - [Erf\left(\frac{x-a'-\sigma_K^2(z)S^2}{\sigma_K(z)}\right)]]; \\ Where S &= S([\Omega]_F), and A = A([\Omega]_F); \\ A([\Omega]_F) &= X e^{[-\mu_w X[L\times\left(\frac{\sin \alpha}{\cos(\alpha+\phi}\right)]]X[[\Omega]_F]}; \\ S([\Omega]_F) &= X e^{[-\mu_w X[\pm\frac{Cu}{F+z}\times\left(\frac{\sin \alpha}{\cos(\alpha+\phi}\right)]]X[[\Omega]_F]}; \\ [\Omega]_F &= [1 + \frac{\tan^2\phi_2}{1+\tan^2\phi_1}]^{1/2}; \dots (4) \end{split}$$

where A and S parameters are defined at formulas according to a series of parameters. I(z) is the area integral of the dose over a plane perpendicular to the z-axis at depth z, normalized to one incident electron, σ (z) is the depth- dependent mean square radial displacement, x, y, z are the coordinates of the dose-delivery point at beam-output coordinates system, u, v, z, are the coordinates of photon-fluence at depth z, a, b, are the field-size magnitudes at depth z, CK are optimization parameters resulting for a triple Gaussian function setting, ΦU is the photon fluence modified for WF, a'(z) =a(l + z/F), b'(z) =b(1 + z/F) are the half side lengths projected into depth z, with F as the source-surface distance (SSD). Parameters μ w (WF material parameter), α (WF angle) and φ (photon-beam divergence angle) are defined at [1-8, 26-29]. The WF manufacturing angle is α . C is the distance in z-coordinate from source to WF surface, L is half-length of WF for y coordinate. In general, Fluence modification is $\Phi U = \Phi 0 / (1 + z/F)^2$ according to [28,29], and $\Phi U = \Phi 0 / (1 + z/F)^2$ according to [28,29]. Then source fluence is Φ_0 modified primarily for the dose-delivery depth z and secondly by the WF parameters [1-8, 26-29]. Parameters µw (WF material parameter), α (WF angle) and ϕ (photon-beam divergence angle) are defined at [1-7,9-15,17,30,31]. Development of Equations 1-4 are extensively presented in [1-8,28,29]. The angles Φ_1 and Φ_2 are the geometrical angles for Omega Factor computation defined [1-8]. Constants are given following the previous equations. This formula is an Omega Factor correction from the equations of classical AAA model [28,29,32,33]. The S and A variables depend on Omega Factor and were mathematically developed in previous numerical

computation programming and 3D Graphical simulations [1-20].

Bio Model and Photon-Dose Parameters from Photon-Dose Model to Linear Quadratic Model Methodology

Table 2 presents BM parameters for complete determination of SR of the BM based on [19,21-24] algorithms, IEM and algebraic calculations. Table 3 shows AAA photon-dose parameters calculations for first optimization step, based on from [1-20]. The most important parameters to be implemented within programming are: approximate Photon Fluence Rate [N x cm⁻² x s⁻¹], and approximate Delivery Dose per second [Gy x s⁻¹]. These data are calculated with experimental dosimetry from the literature for 18 Mev [27,34]. 18 Mev Photon Beam Spectrum dataset is taken form experimental literature [27,34]. Once the Photon-Dose delivery is determined, the magnitude is implemented within BM. Thus, the numerical development/design of the 3D simulations programs is rather difficult. Table 2 also shows approximations that were applied for all calculations at software. Table 2 AAA model main Photon-Beam Spectra parameters for 3D Graphical simulations. Dataset is large, it is recommended to consult references [1-8,20,26-29,34,35]. The first calculation and program implementation are the AAA photon-dose (3-4). The second part involves implementation of photon-dose into BM algorithm for calculation of NS Rate by using (1-2) algorithms [36,37]. The patterns design of Matlab programs are rather arduus to obtain acceptable results approximations.

Breast Tumor Radiotherapy TPO Bio Model Specifications

Breast tumors (BT) incidence and prevalence has increased significantly in recent decades [19,20,25,38,39], especially in developed countries. Epidemiology statistics [25], by year 2013 show these approximated data,

- 1) Breast cancer: about 11% of all tumors.
- **2)** Europe shows an approximate incidence of 90 per 100,000 women in Western Europe [25].
- **3)** Developed world zones [25]: higher (greater than 80 per 100,000).

The most frequent statistical anatomical place for BT growth is the superior-external quadrant of mammary glands. In that case, it is also frequent when the tumor is palpable, [25,38,39] to find ganglia tumoral invasion at corresponding axilla. Therefore, TPO depth parameter z=5 cm is appropriate in general for small mammary glands, while z=15 cm approximation is convenient for bigger mammary glands and/or internal tumor invasion. That is the TPO simulations justification for simulation parameters in this study [25,38,39]. Constant medical exploration, and self-exploration constitutes the first simple method for early detection [25,38,39]. Total dose protocol for TPO in breast cancer is about [40,41] Gy, with variations from country, institution to country, institutions, Cancer Care Centers [25,38,39].

Integral Equation TCP and TCCP Model [CASESNOVES 2022]

Following the previous mathematical development, Tumor Control Cumulative Probability (TCCP) was determined in a new Integral Equation Model (IEM) [19] Equation (5). The usual Poisson Distribution for the TCP, [21-24], was approximated to a Binomial Distribution to obtain an analytic solution of the TCCP integral equation published in [21]. This method starting point was based on previous BM contributions [21-24]. As detailed in [19], the IEM for TCCP reads.

$$TCP(\alpha,\beta) = \frac{1}{2\pi\sigma^2} \dots$$
$$\dots \int_{\beta_1}^{\beta_2} \int_{\alpha_1}^{\alpha_2} [1 - P(\alpha,\beta)] \times \dots$$
$$\dots e^{\left[-\frac{1}{2\sigma^2}\left[(\alpha - \overline{\alpha})^2 + (\beta - \overline{\beta})^2\right]\right]} \dots$$
$$\dots d_{\alpha} d_{\beta} = I_1 - I_2 \dots \dots (5)$$

Where

 α : Clonogen radiosensitivity integral parameter β : Clonogen ra-

diosensitivity integral parameter

σ: Approximated standard deviation both for α, β parameters I₁: First Gaussian integral, Normal Distribution

 I_2 : Second Probability Function Gaussian convolution with Binomial Distribution α_1 : Inferior 2D IE limit

 $\alpha_2\!\!:$ Superior 2D IE limit $\beta_1\!\!:$ Inferior 2D IE limit $\beta_2\!\!:$ Superior 2D IE limit

P (α, β): Binomial Probability function for Convolution

All the development of this IEM and analytic determination with Erf functions is explained in [19]. Analytical solution of this IEM at [19].

Software-Programming Method

The programming method(s) used for this study are based in a high number of previous papers [1-20]. For BM, new advances and programs were required. Table 4 shows the techniques and software improvements to obtain better calculations, 3D processing images, error determinations, and get applied exactly the new/ modified BM model formulas/algorithms.

Table 4: AAA model main	parameters for 3	3D Graphical	simulations	[1-20].	Program	software	parts	numerically	calculated for
Equations 1-4 with new corre	ections. The data	from Tables 1	-3 was imple	mentee	d.		-	-	

Brief of Software Method(S) In Matlab System					
Computational difficulty	Programming technical solution	Comments			
Implementation of Photon- Dose AAA model into BM	Precision required to check and verify correct order and exactitude of every loop, pattern, and array	REASON Programming involves double optimization: Firstly Photon-dose AAA model			
Optimization of implementation of Photon- Dose AAA model into BM for fast running time	Precision required to organize and optimize correct order and exactitude of every imaging processing subroutine	REASON Programming involves double optimization: Photon-dose AAA model and then optimal photon- dose implementation into BM			
Imaging processing method for final BM	Matrices size fit for better image and minimum running time	Several subroutines can be used.			
Imaging processing method for final BM parameter intervals	To get good-quality 3D BM images is necessary optimize, select, and order the best parameters combination linked to imaging processing tools	The exact and approximated parameters magnitudes implementation are rather laborious			
Programming patterns to change axes intervals	Setting axes for inverse probability magnitude	These patterns are rather arduous			

3D Numerical and Imaging Simulations Results

1-6 for [z= 5,15 cm, 18 Mev]. The Numerical results are presented in Table 5 (Figure 2-7).

The 3D Graphical Optimization of NS Rate is shown in Figures



Figure 2: Matlab NS Rate simulation 3D image for 18 Mev photon-beam at 5 cm depth-dose with Omega Factor and I(z) corrected. The delivery time shot is 1 second. Matrices for Image Processing have about [300 x 300] elements. Imaging Processing Method 1. At figure, inset, axes interval modifications explained.



Figure 3: Matlab NS Rate simulation 3D image for 18 Mev photon-beam, 5 shots, at 5 cm depth-dose with Omega Factor and I(z) corrected. The delivery time shot is 1 second. Matrices for Image Processing have about [300 x 300] elements. Imaging Processing Method 1.



Figure 4: Matlab NS Rate simulation 3D image for 18 Mev photon-beam, 10 shots, at 5 cm depth-dose with Omega Factor and I(z) corrected. The delivery time shot is 1 second. Matrices for Image Processing have about [300 x 300] elements. Imaging Processing Method 1.



Figure 5: Matlab NS Rate simulation 3D image for 18 Mev photon-beam, 1 shot, at 15 cm depth-dose with Omega Factor and I(z) corrected. The delivery time shot is 1 second. Matrices for Image Processing have about [300 x 300] elements. Imaging Processing Method 1.



Figure 6: Matlab NS Rate simulation 3D image for 18 Mev photon-beam, 15 shots, at 5 cm depth-dose with Omega Factor and I(z) corrected. The delivery time shot is 1 second. Matrices for Image Processing have about [300 x 300] elements. Imaging Processing Method 1.



Figure 7: Matlab NS Rate simulation 3D image for 18 Mev photon-beam, 10 shots, at 5 cm depth-dose with Omega Factor and I(z) corrected. The delivery time shot is 1 second. Matrices for Image Processing have about [300 x 300] elements. Imaging Processing Method 1. Inset, explanations for Maximum and Minimum NS Rates.

Clinical Radiotherapy Medical Physics Applications

Table 5 shows a resume of breast tumor medical physics prin-

cipal applications for radiotherapy TPO. Those prospective applications are also useful for radiotherapy research/applications on breast tumors and other types of cancer (Table 6).

 Table 5: Brief of Numerical results for Survival Rate. Results agree with literature, e. g., [Figure 10.1 and others, [22]. Matlab 3D image processing data.

Numerical Results for NS Rate Breast Tumors Radiotherapy Biological Model							
Shot time [simulated; 1 shot = 1 second]	Depth z=5 cm MaximumDepth z=5 cm Minimum[NS Rate x100] [thin part[NS Rate x100] [broadWF]part WF]		Depth z=15 cm Maximum [NS Rate x100] [thin part WF]	Depth z=15 cm Minimum [NS Rate x100] [broad part WF]			
1 s	99%	93%	99%	95%			
5 s	95%	70%	96%	77%			
10 s	90%	48%	93%	58%			
15 s	85%	32%	90%	44%			

Table 6: Some radiotherapy Breast Cancer TPO Medical Physics study applications.

Clinical Medical Physics TPO Breast Tumor Applications						
Туре	Clinical Research		Mixed	Comments		
Treatment planning optimization	TPO precise for breast tumors with BMs	TPO Modelling BMs developments	Clinical improvements with BMs after research	Inverse planning system set up on BMs		
LINAC OPTIMIZATION	Optimization of photon- dose for BMs	LINACs BMs Usage for IMRT, IMPT	Exploration of new possibilities	Manufacturing adaptation of LINACs fro BMs		
Theoretical new models	Dosimetry improvements in accuracy according to radiobiology experimental	From tumor survival clinical statistics advances in BMs	New BMs research soured, both theory and clinical experimental tials	BMs got experimental evidence to be set on TPO		

Discussion and Conclusion

The objective of the research was to apply algorithms from [20] to get NS acceptable 3D and numerical simulations for breast cancer rutinary photon dose depths [z= 5, 15 cm]. The photon dose parameters were presented extensively in publications series [1-20]. The results are a series of 3D Graphical Optimization sets with NS Rates for those parameters. Numerical and Graphical data obtained is agree closely the experimental of literature [21-24]. Biomodel parameter NS Rate agrees to experimental dataset in the literature [64-67] with approximations. The 2D polynomial approximation for $[\alpha]$ variation related to N0 magnitude results be fine. Advantages of this method are accuracy in 3D determination of NS and easy method from the presented models to calculate Clonogenes Survival Rates for breast tumors. Inconvenient, for example, are the experimental variation of [α] parameter related to N_o, Figure 1. The results are calculated for AAA model in water without tissue heterogeneities corrections. However, mammary gland overall density is close to water one. Grosso modo, new BMs algorithms were presented with formulation, numerical and 3D Graphical Optimization results. Appropriate applications for breast cancer TPO come up from all the solutions obtained.

Scientific Ethics Standards

IEM was developed by Dr F Casesnoves on $15^{\rm th}\,March\,2022.\,All$

initial equations were developed from previous researchers' contributions [21-24]. The NS initial integral formulas were published in [21-14]. From those equations, all the mathematical development is original from the author [1,20]. This article has previous papers mathematical techniques, [1-18], whose use was essential to make IE model analytic solution, approximations, and NS Rate and Probability Function. Table 3 is understanding reader's essential from [20], and Equations 12-4 also. The complete mathematical development will be carefully reviewed/checked in subsequent publications. The number of Dr Casesnoves publications at references is intended also for reader's learning. When editors permit, references are ordered in subsequent rationale arrangement. This study was carried out, and their contents are done according to the European Union Technology and Science Ethics. Reference, 'European Textbook on Ethics in Research' [37,42-44]. European Commission, Directorate-General for Research. Unit L3. Governance and Ethics. European Research Area. Science and Society. EUR 24452 EN [42-4460-63]. And based on 'The European Code of Conduct for Research Integrity'. Revised Edition. ALLEA. 2017 [37,42-44]. This research was completely done by the author, the calculations, images, mathematical propositions and statements, reference citations, and text is original from the author. When a mathematical statement, proposition or theorem is presented, demonstration is always included. If any results inconsistency is found after publication, it is clarified in subsequent contributions. The article is exclusively scientific, without any commercial, institutional, academic, religious, religious-similar, non-scientific theories, personal opinions, friends and/or relatives' favors, 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 [37,42-72].

Author's Biography

Dr Francisco Casesnoves earned the Engineering and Natural Sciences PhD by Talllinn University of Technology (started thesis in 2016, thesis Defence/PhD earned in December 2018, official graduate Diploma 2019). Dr Casesnoves is European Union and Internationally qualified as Doctor in Engineering to supervise PhD Theses, Master Theses, and Bachelor Theses in science and engineering. 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, ethical professional reasons, anti-state monarchy corruption positions, 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 publications with approximately 65 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 2019-2020], 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 PhD thesis is an Estonian scientific service to European Social Fund and several EU Research Projects. 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 40 physics-engineering articles, two books' series, and 1 industrial radiotherapy project associated to Europe Union EIT Health Program (Tartu University, 2017).

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ISSN: 2574-1241

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DOI: 10.26717/BJSTR.2022.46.007337

Francisco Casesnoves. Biomed J Sci & Tech Res

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