



RPT-TEC

Sept 2022

MIRDcalc and MIRDfit: MIRD tools for pharmokinetic modelling and dosimetry

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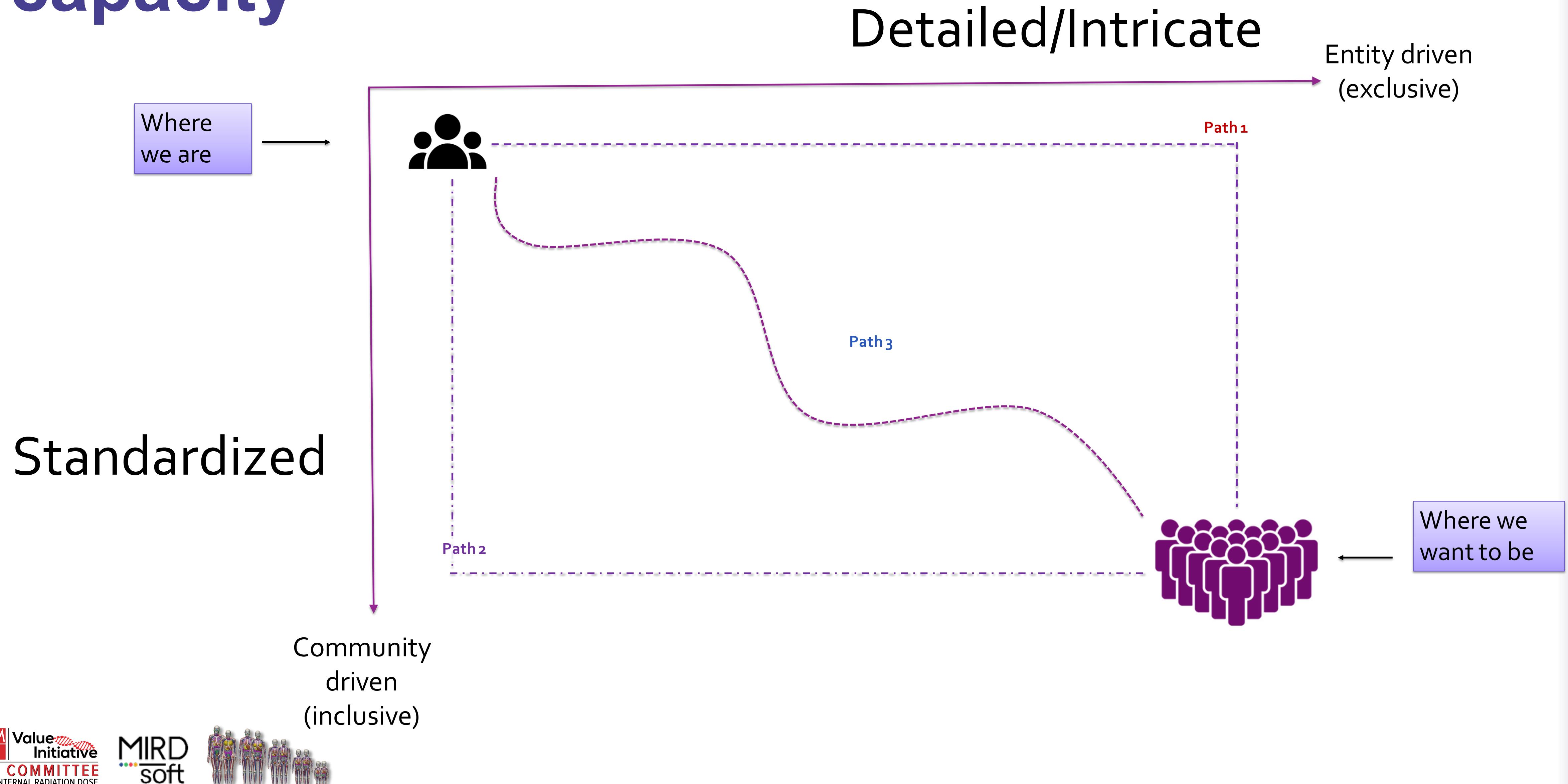
Presentation overview

- Topics

- Standardization and error propagation via MIRDsoft.org tools
 1. MIRDcalc: TIAC->dosimetry
 2. MIRDfit: data points -> TIAC
- Perspectives on organ level dosimetry
- Perspectives on standardization
- Closing remarks



Current status of personalized dosimetry capacity



Introduction to MIRDsoft.org

- The SNMMI MIRD committee initiative to make a suite of free dosimetry-supporting software tools
- Collaborative project
 - MIRDcalc grant
 - NIBIB - U01, Bolch (UF)/Kesner(MSK)
 - MIRDcell grant
 - NCI - R01, Howell (Rutgers)
- Project hosted on www.MIRDsoft.org
 - Software dissemination
 - Community platform



Introduction to MIRDsoft.org



- Available now
 - **MIRDcalc v1 – organ-level internal dosimetry**
 - MIRDcell v3 – cell-level dosimetry
- Coming soon
 - **MIRDfit – biodistribution fitting/statistical analysis**
 - MIRDct – CT dosimetry
 - MIRDmc – voxel & mesh Monte Carlo dosimetry



MIRDcalc screenshot

MIRD SCHEMA ORGAN LEVEL DOSIMETRY SPREADSHEET

Biodistribution Model INPUT

Element	Isotope	Sex	Phantom
Ho	Lu-176	Female	57 kg (interp)
Ir	Lu-177	Male	58 kg (interp)
La			59 kg (interp)

subject ID: LU-177 test patient

Dosimetry Estimate OUTPUT

Input parameters:

Phantom	58 kg (interp)	♀	% injection accounted for	32%	W_R
Isotope	Lu-177		Input S value uncertainty	20%	γ
Halflife	1.5953E+02 [hours]		# organs with nonzero TIACs	7	β
Subject ID	Lu-177 test patient		Input isotope/organ UID	MIW	α

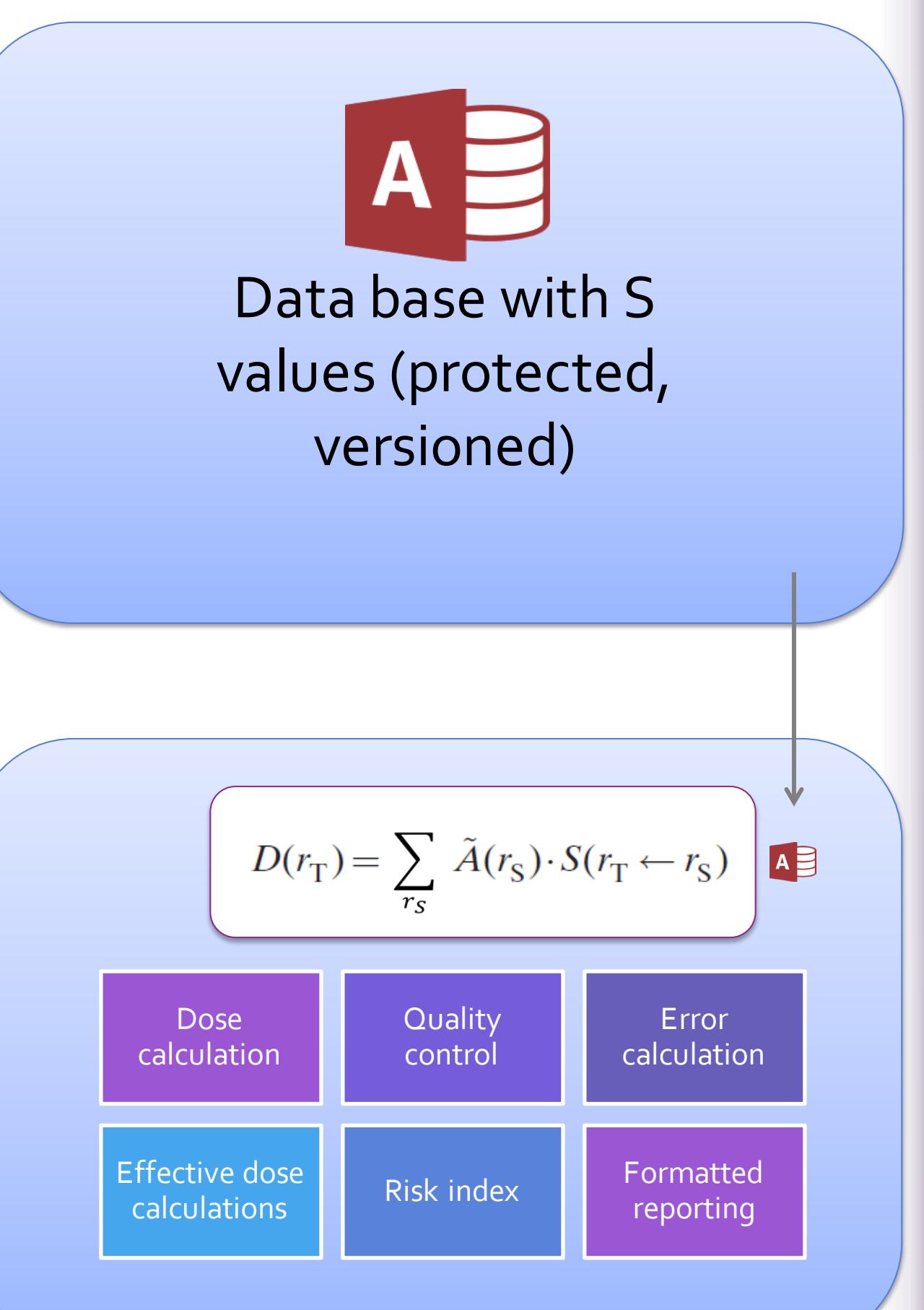
Estimated dosimetry (absorbed dose) - 37/50 displayed here

Organ	Abs Dose [mGy / MBq]	Uncertainty (SD) [mGy / MBq]	MIRD calc	Detr. Weighted & Effective Dose * ^a [mSv / MBq]	σ [mSv / MBq]
Adipose tissue	9.81E-02	1.83E-02	EDW	9.52E-02	4.85E-03
Adrenals	9.84E-02	1.58E-02	E Effective Dose	9.57E-02	4.50E-03
Bone - endosteal cells	9.11E-02	9.38E-03			
Bone marrow - red (act)	9.74E-02	1.10E-02			
Brain	1.35E+03				
Breast tissue	4.44E+02				
Colon - ICRP133	3.04E+00				
Esophagus	8.86E-02				
Extrathoracic region	4.04E-01				
Eye lens	4.00E-01				
Gallbladder wall	8.11E+00				
Heart wall	2.80E+02				
Kidneys	3.41E+02				
Liver	2.02E+03				
Lungs					
Major blood vessels					
Muscle					
Oral mucosa					
Pancreas					
Rest of blood	0.238	0%	0.11		
Rest of parenchym	55.78	24%	1.9		
Salivary glands					
Spleen	1.8	1%	0.005		
Thymus					
Thyroid					
Tumor1_300cc_10c	5.4	2%	0.1		
Tumor2_28cc_50%	0.8	0%	0.1		
Urinary bladder con					
Rest of body					
Rest of body mass: 54.8 Kg					
Organ model (S value) uncertainty	20%				
Waste					
Total TIAC entered into table:	72.93	% theoretical activity accounted:	32%		
Total TIAC required to account for 100% emissions:	230.15				

* Time integrated activity coefficients (TIACs) in units [hours] = time-integrated activity [dis] divided by the administered activity [dis/time]

Internal dosimetry spreadsheet

Ready Scroll Lock

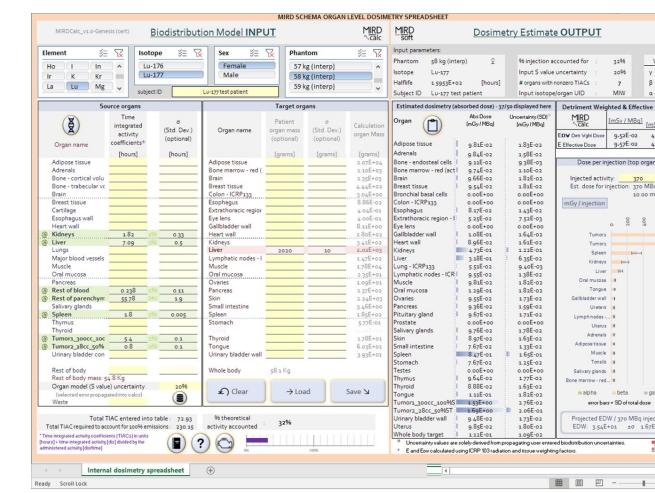




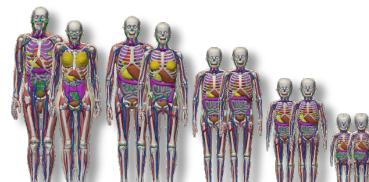
MIRDcalc

● Innovations

- 81 source regions, 48 target regions, 333 isotopes
- Single screen user interface
- Real time processing
- Graphical quality control checks
- Modern ICRP phantoms
 - Well documented
- Spectrum of phantom models (m/f, pediatric to adult, 1 kg steps)
- Dynamic source regions
 - Rest of body
 - Rest of blood
 - Rest of parenchyma



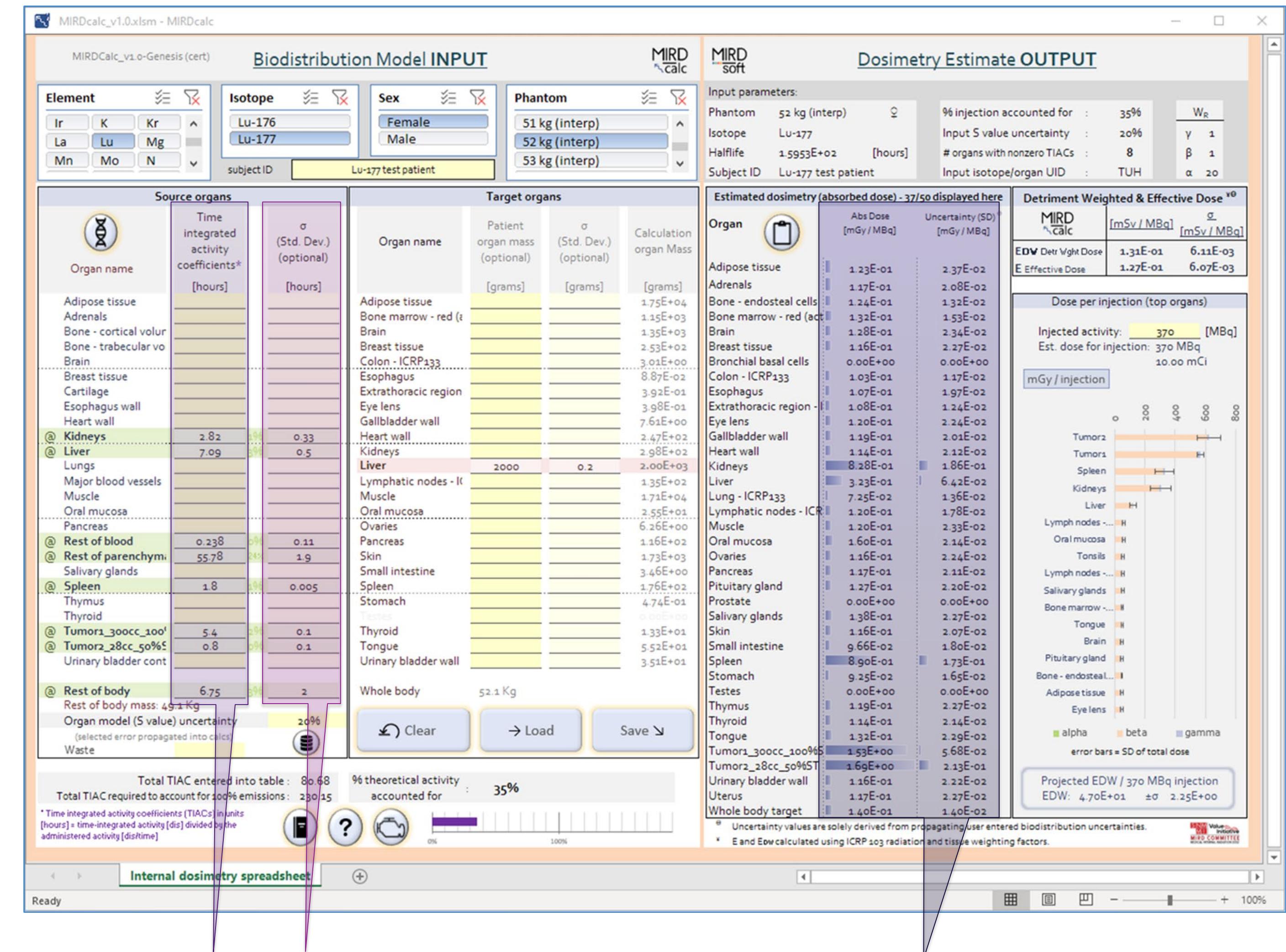
- (New) blood model
- Uncertainty propagation
- Tumor dosimetry module
- Output: Organ dose, effective dose, detriment weighted dose, risk index
- Thorough case documentation
 - Highly detailed output text files
 - Default screen capture
- Command line execution
 - Supports batch processing, possibly 3rd party
- And more...





MIRDcalc dosimetry software

- MIRDcalc has capacity to propagate uncertainty



TIAC

 σ_{TIAC} $Absorbed\ dose \pm \sigma_{Dose}$

Uncertainty estimation in time-activity curve fitting

- Pharmacokinetic modelling used to transform raw measurements into biodistribution input for dosimetry calculations
 - Time integrated activity coefficients (TIACs – area under the curve) used to define sources of radiation
 - Uncertainty in time point measurements can mathematically propagate into uncertainty in TIACs
- Existing software
 - SAAM-II
 - NUKFIT
- Existing guidance document
 - EANM Procedural Guidelines for Uncertainty Analysis





Molecular radiotherapy: The NUKFIT software for calculating the time-integrated activity coefficient

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[European Journal of Nuclear Medicine and Molecular Imaging \(2018\) 45:2466–2474](https://doi.org/10.1007/s00259-018-4136-7)
<https://doi.org/10.1007/s00259-018-4136-7>

GUIDELINES



EANM practical guidance on uncertainty analysis for molecular radiotherapy absorbed dose calculations

Jonathan I. Gear¹ · Maurice G. Cox² · Johan Gustafsson³ · Katarina Sjögren Gleisner³ · Iain Murray¹ · Gerhard Glatting⁴ · Mark Konijnenberg⁵ · Glenn D. Flux¹

Received: 9 August 2018 / Accepted: 14 August 2018 / Published online: 14 September 2018
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Abstract
 A framework is proposed for modelling the uncertainty in the measurement processes constituting the dosimetry chain that are involved in internal absorbed dose calculations. The starting point is the basic model for absorbed dose in a site of interest as the product of the cumulated activity and a dose factor. In turn, the cumulated activity is given by the area under a time-activity curve derived from a time sequence of activity values. Each activity value is obtained in terms of a count rate, a calibration factor and a recovery coefficient (a correction for partial volume effects). The method to determine the recovery coefficient and the dose factor, both of which are dependent on the size of the volume of interest (VOI), are described. Consideration is given to propagating estimates of the quantities concerned and their associated uncertainties through the dosimetry chain to obtain an estimate of mean absorbed dose in the VOI and its associated uncertainty. This approach is demonstrated in a clinical example.

Keywords Dosimetry · Uncertainty analysis

Introduction

Internal dosimetry following the administration of radio labelled pharmaceuticals for diagnostic and therapeutic purposes is a well-established technique in nuclear medicine. The European Association of Nuclear Medicine (EANM) is a professional non profit medical association that facilitates communication worldwide among individuals pursuing clinical and research excellence in nuclear medicine. The EANM was founded in 1985. This guidance document is intended to assist practitioners in providing appropriate nuclear medicine care for patients. The rules provided in the document are not inflexible or requirements of practice and are not mandatory. They were also <1% (standard error between 0.4% and 5%). In general, the application of the software is user-friendly and the results are mathematically correct and reproducible. An application of NUKFIT is presented for three different clinical examples.

9



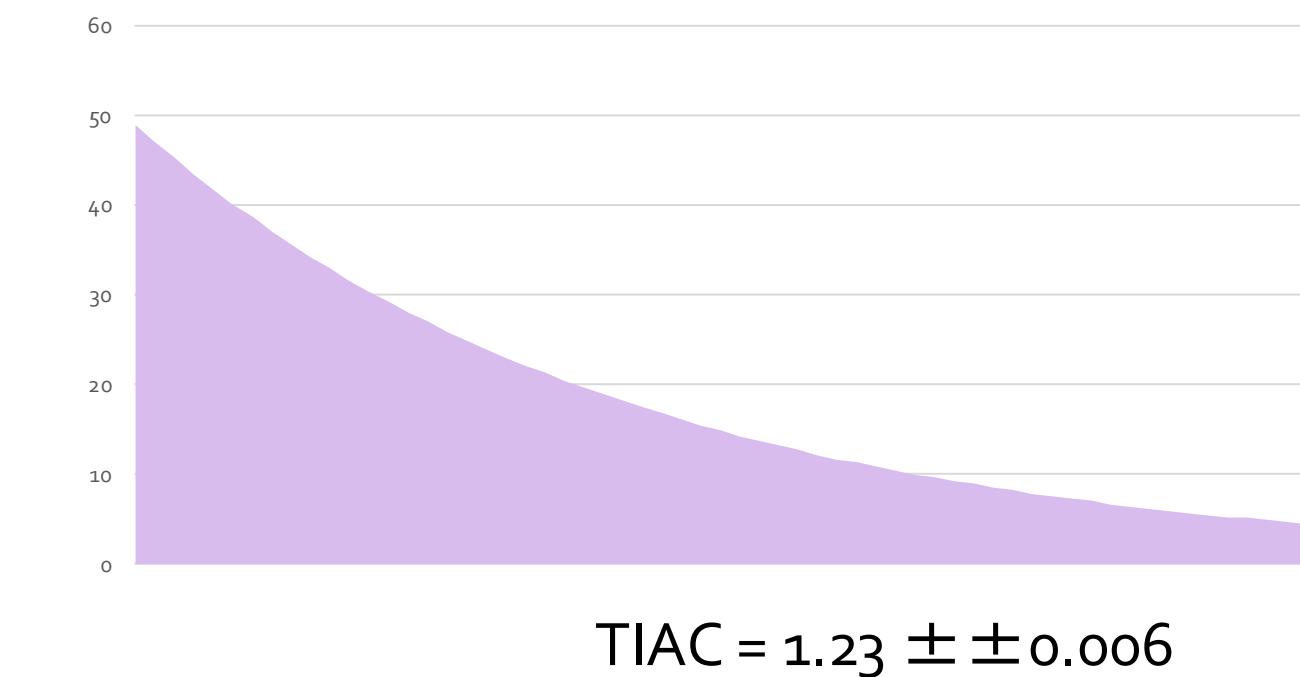
Review of curve fitting procedures

Given calibrated measurements:

1. Select (exponential) modelling equation
2. Curve parameters determined by iterative fitting approach
 - o Least squares approach using non-linear regression to minimize objective function
3. Time-integrated activity coefficient derived from integration
4. Uncertainties are obtained by Gaussian error propagation, taking the variances and covariances of the fit parameters into account
5. Additional statistics can be determined, and used in selection of “best model”
 - o R^2 , Akaike information criterion, Akaike corresponding weights, Weighted residuals

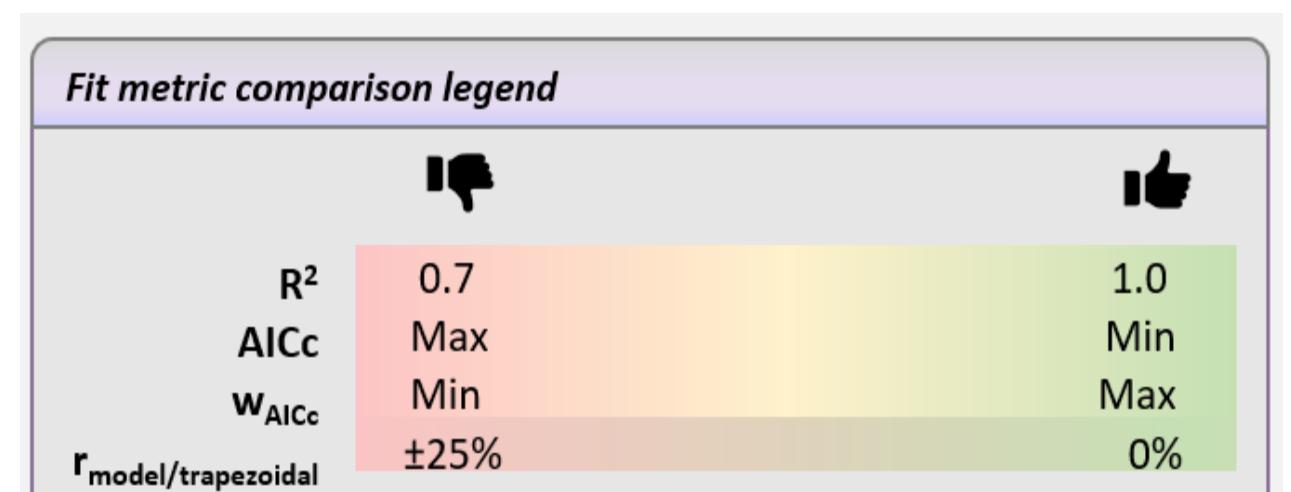
x	f(x)	u(x)
1	33	5
2	22	6
3	15	7
4	10	8
n

$$AUC = \int_0^{\infty} A_1(e^{-(\lambda_1 + \lambda_{phys})t} - e^{-(\lambda_2 + \lambda_{phys})t})dt = \frac{A_1}{\lambda_1 + \lambda_{phys}} - \frac{A_1}{\lambda_2 + \lambda_{phys}}$$



Variance-covariance table

	b0	b1	b2
b0	1.014933	0.009028	0
b1	0.009028	0.000103	0
b2	0	0	1E+16



Review of curve fitting procedures

○ Microsoft solver output

Title: Model5Fit						Correlation Table for Parameters			Breush-Pagan Test for Homoscedasticity of Residuals			
Weight	SD for Y	SD for X's	Observations	Parameters		b0	b1	b2	Breush-Pagan MSS/2	ChiSqcrit(.05)	Check(.05)	
Absolute	N	N	4	3	1	b0	1	0.882848583	0	0.067579832	3.841458821	
						b1	0.882848583	1	0	Ok	0.794893693	
						b2	0	0	1	0.164065213	3.841458821	
ANOVA						Ratio Objects/Parameters			Error Sum of Squares (RSS or SSE)			
Source	Sum Sq.	Degr.	Fred.	Mean Sq.	F calc.	Crit.	F (.05)		n	p	dF	
Model	5299.666346	2		2649.833173	11385.93616	199.5			0.232728617	4	3	Error Me
Residual	0.232728617			1	0.232728617					0.232728617	0.241209772	
Total	5299.899075	3										
Chi Square Statistics						Observation Statistics (CI based on Z Statistic)			R			
ChiSq. Calc	Crit. ChiSq. (.05) Inf.			Crit. ChiSq. (.05) Sup.		Y	Ycalc	Low. Regr. Conf. Band (.95)	Up. Regr. Conf. Band (.95)	R^2	R^2 adj	
0.232728617	0.00393214			3.841458821	0.62950794	22.59288635	22.55376416	20.39588962	24.71163869	0.999978044	0.999956088	
						10.31188916	10.38159246	9.127913392	11.63527152	-0.36426	0.999868264	
						4.813540079	4.778690652	3.339608914	6.21777239	0.287323		
						2.197005332	2.199651397	0.583119636	3.816183158	-0.03406		
R						Cut Point			PRESS			
	Low. CI (.95)	Up. CI (.95)	R^2	R^2 adj					MEP	0.310126598		
	0.999978044	0.9858773	0.999999966	0.999956088	0.999868264				R^2 prediction	0.999765938		
Parameters Statistics (from linearized Variance-Covariance Matrix)						SDEC			SDEC			
Par.	Value	Std. Err.	Z calc.	Crit. Z (.05)	Low. CI (.95)							
b0	48.99751939	1.007438849	48.63572558	1.959963985								
b1	0.396995227	0.010150499	39.1109069	1.959963985								
b2	39.11467473	100000000	3.91147E-07	1.959963985								
Parameters Statistics and Wald Test (from Fisher Information Matrix)						SDEP			SDEP			
Par.	Value	Std. Err.	Z calc.	Crit. Z (.05)	Low. CI (.95)							
b0	48.99751939	1.008125899	48.60257973	1.959963985								
b1	0.396995227	0.01015934	39.07687147	1.959963985								
b2	39.11467473	100000000	3.91147E-07	1.959963985								
Chi^2 Joint Confidence Parameters Statistics						AIC (Akaike Information Criterion)			-5.376706402			
Par.	Value	Low. CI (.95)	Up. CI (.95)	Passed	Prob.	Check						
b0	48.99751939	46.18124073	51.81379805	0	Ok							
b1	0.396995227	0.368619674	0.42537078	0	Ok							
b2	39.11467473	-279548309.2	279548387.4	1	No							
Variance-Covariance Table for Parameters						AICc (Adjusted Akaike Information Criterion)			65535			
	b0	b1	b2									
	1.014933034	0.009028016	0									
	0.009028016	0.000103033	0									
	0	0	1E+16									
Hessian Variance-Covariance Table for Parameters						HQ (Hannan Quinn Criterion)			-9.416900842			
	b0	b1	b2									
	1.016317828	0.009043782	0									
	0.009043782	0.000103212	0									
	0	0	1E+16									
Correlation Table for Parameters						HQc (Adjusted Hannan Quinn Criterion)						
	b0	b1	b2									
	1.333333333	No										
Durbin Watson Statistic for Autocorrelation of Residuals						BIC (SBC-Schwarz Bayesian Criterion)			-7.217823318			
Note: DW tests do not apply directly to residuals from regression through the origin.						BICc (SBCc-Adjusted Schwarz Bayesian Criterion)						
Durbin Watson Bounds Test						Models Comparison using Information Criterion (Replace IC)						
	DW	DWcritInf(.05)	DWcritSup(.05)	4-DWcritSup(.05)			IC	IC				
							Min IC	65535	65535	65535		
Box Pierce Ljung Q' Test for Autocorrelation Of High Order Autoregressive Order							Weight Sum	2	1	1		
							Probability	0.5	0.5			
	Q'	ChiSqcrit(.05)	Check (.05)				Evidence Ratio	1	1			
	1	3.828275665	5.991464547	Ok								
Residuals Runs Test						Log-Likelihood Section						
	Res. +	Res. -	Runs	Low Crit. Runs (.05)	Up. Crit. Runs (.05)							
	2	2	4	2	4							
				No	1							
Extra Sum F Test for Significance of Extra Parameters in Nested Models (Replace Residual Mean Square and dF)						Log-Likelihood Function			-2.759989654			
Reduced Model Residual Sum Of Squares						Log-Likelihood Tests for Nested Models (Replace RSS and NPar values)						
							RSS	NPar	Test Value	dF	Crit. Ch	
							0.232728617		3		Lagrange Multipl	
							Unrestricted Model		3		Likelihood Ratio	
							Ratio	1	0	#NUM!	Wald Statistic	
F Test for Models Comparison (Replace Residual Mean Square and dF)						F Test for Models Comparison (Replace Residual Mean Square and dF)						
							Residual Sum of Squares #1	dF #1	Residual Sum of Squares #2			
							0.232728617	1	0.232728617	1	647.7890115	
Reduced Model dF						Extra Sum F Test for Significance of Extra Parameters in Nested Models (Replace Residual Mean Square and dF)						
							Full Model					
							0.23272861					

MIRDfit interface

S_{NM}
MI Value Initiative
MIRD COMMITTEE MEDICAL INTERNAL RADIATION DOSE

MIRD MIRD fit soft + New case Documentation

Study setup

Biodistribution input

Trapezoidal integration

Regression-based integration

Monoexponential (physical decay only)

Monoexponential





MIRDfit interface

Study setup

Subject ID (optional)
89Zr-peptide

Notes (optional)
Adult male reference dosimetry estimate

Select an element

Se	Si	Sm	Sn
Sr	Ta	Tb	Tc
Te	Th	Ti	Tl
Tm	U	V	W
Xe	Y	Yb	Zn
Zr			

Select an isotope

- Zr-85
- Zr-86
- Zr-87
- Zr-88
- Zr-89**
- Zr-89m
- Zr-93
- Zr-95

Select an phantom

- ICRP 00 Newborn female
- ICRP 00 Newborn male
- ICRP 01 year old female
- ICRP 01 year old male
- ICRP 05 year old female
- ICRP 05 year old male
- ICRP 10 year old female
- ICRP 10 year old male
- ICRP 15 year old female
- ICRP 15 year old male
- ICRP Adult Female
- ICRP Adult Male**

Fitter tool

Monoexponential (physical decay only)

$$\%IA(t) = A_1 e^{-\lambda_{phys} t}$$

Parameter A_1 : Value 6.889688, Std. Err. 0.000884

Monoexponential

$$\%IA(t) = A_1 e^{-(\lambda_1 + \lambda_{phys})t}$$

Parameter A_1 : Value 6.889688, Std. Err. 0.000884

Parameter λ_1 : Value 0.18102, Std. Err. 0.0433

Parameter λ_1 : Value 0.32601, Std. Err. 0.00115

Summary output

TIAC	3.87E-01	[h]
TIAC Std. Err.	2.04E-02	[h]
TIAC CV	5.3%	
1	49	
2	27	
3	23	
4	14	
5	10	
6	6	
7	4	
8	2	
9	1	
10	0	

Actions

- Send to shelf
- Fit (re-fit) this model
- Reset/reject this model





MIRDfit interface

Study setup

- Subject ID (optional): ^{99}Zr -peptide
- Notes (optional): Adult male reference dosimetry estimate
- Select an element: Se, Si, Sm, Sn, Sr, Ta, Tb, Te, Th, Ti, Tl, Tm, U, V, Y, Yb, Zn, Zr
- Select an isotope: Zr-85 , Zr-86 , Zr-87 , Zr-88 , Zr-89 , Zr-89m , Zr-90 , Zr-95 , Zr-97
- Select a phantom: ICRP 00 Newborn female, ICRP 00 Newborn male, ICRP 01 year old female, ICRP 01 year old male, ICRP 05 year old female, ICRP 05 year old male, ICRP 10 year old female, ICRP 10 year old male, ICRP 15 year old female, ICRP 15 year old male, ICRP Adult Female, ICRP Adult Male

Biodistribution input

Biodistribution toolbox

Select a source region

- Adrenals
- Alveolar-interstitial (sub lungs)
- Blood (classic ICRP)
- Bone - cortical surface (sub volume)***
- Bone - cortical volume
- Bone - trabecular surfaces (sub volumes)
- Bone - trabecular volumes
- Bone marrow - red (active)*****
- Bone marrow - yellow (inactive)
- Brain
- Breast tissue

Summary output

%IA entries are:

- Biological uptake/clearance
- Effective uptake/clearance

Format for input variances of %IA data (weighting scheme):

- Standard deviation
- Relative standard deviation (coefficient of variation)
- Weight (direct entry)

Time, t [h]	%IA	RSD
4	6.889687881	0.037267081
24	2.07978333	0.086206897
48	0.638076123	0.045454545
72	0.234590016	0.1
120	0.030694386	0.25

Clear time **Clear %IA** **Unshelve**

Select a source region

Source region viewer

Summary output

TIAC, TIAC Std. Err., TIAC CV, R², AIC_c, N obs, F undetectable, Alerts

ICRP Adult Male

Actions

- Send to shelf
- Fit (re-fit) this model
- Reset/reject this model
- Fit (re-fit) this model
- Reset/reject this model

Equation

$$= A_1 e^{-(\lambda_1 + \lambda_{\text{phys}})t}$$

λ_1 : 89688, 0.000884

λ_2 : 18102, 0.0433

λ_3 : 2601, 0.00115

λ_4 : 1.57E+00, 0.431E-02

λ_5 : 2.7%, 0.931

λ_6 : 14.13, 100.0%

λ_7 : 0.34, None.

Graph

Weighted residual vs Time [h]

Time [h]: 0, 50, 100, 150, 200



MIRDfit interface

Trapezoidal fitting toolbox

Nose: (t0 to t1)

- %IA(0) = 0
- %IA(0) = %ID(t0)
- %IA(0) = Organ blood content
- Linear
- Exponential extrapolation (t1, t2)
- Exponential extrapolation (phys to t1)
- Excluded (no extrapolation)

Tail: (tn to tD)

- Exponential extrapolation (phys to inf)
- Exponential extrapolation (tn-1, tn)
- Linear extrapolation to 10 phys half-lives
- Linear extrapolation (tn-1, tn)
- Excluded (no extrapolation)

Summary output

TIAC 5.20E+00 [h]

Nose component: 2.7%
Body component: 63.2%
Tail component: 34.1%

Trapezoidal integration

Nose: (t0 to t1)

- %IA(0) = 0
- %IA(0) = %ID(t0)
- %IA(0) = Organ blood content
- Linear
- Exponential extrapolation (t1, t2)
- Exponential extrapolation (phys to t1)
- Excluded (no extrapolation)

Extrapolation to 10 phys half-lives

- Linear extrapolation (tn-1, tn)
- Excluded (no extrapolation)

Summary output

TIAC 1.67E+00 [h]

Nose component: 16.8%
Body component: 83.2%
Tail component: 0.1%

Regression-based integration

Regression toolbox

Fit all selected (quickfit)

Reset all

Regression model filter

- Biexponential (%IA = %blood at t = 0)
- Biexponential (%IA = 100 at t = 0)
- Biexponential (rapid clearance & slow clearance phases)
- Biexponential (rapid clearance phase & physical decay)
- Biexponential (uptake & clearance phases)
- Monoexponential
- Monoexponential (physical decay only)

Summary output

TIAC 3.87E-01 [h]

TIAC Std. Err.	2.04E-02
TIAC CV	5.3%
R ²	0.1449
AICc	30.27
W stats	0.0%
F model/exponential	0.23
Alerts	None.

Weighted residual

TIAC 4.31E-02 [h]

TIAC Std. Err.	4.31E-02
TIAC CV	2.7%
R ²	0.991
AICc	14.13
W stats	100.0%
F model/exponential	0.34
Alerts	None.

Weighted residual

Monoexponential (physical decay only)

%IA(t) = $A_1 e^{-\lambda_{phys} t}$

Filter step

Parameter A_1
Initial guess 6.889688

Filter results

Parameter A_1
Value 0.34188
Std. Err. 0.018

Summary output

TIAC 1.57E+00 [h]

TIAC Std. Err.	4.31E-02
TIAC CV	2.7%
R ²	0.991
AICc	14.13
W stats	100.0%
F model/exponential	0.34
Alerts	None.

Weighted residual

Monoexponential

%IA(t) = $A_1 e^{-(\lambda_1 + \lambda_{phys})t}$

Filter step

Parameter A_1 λ_1
Initial guess 6.889688 0.000884

Filter results

Parameter A_1 λ_1
Value 0.18102 0.0433
Std. Err. 0.32601 0.00115

Summary output

TIAC 1.57E+00 [h]

TIAC Std. Err.	4.31E-02
TIAC CV	2.7%
R ²	0.991
AICc	14.13
W stats	100.0%
F model/exponential	0.34
Alerts	None.

Weighted residual

Actions

Fit metric comparison legend

	R ² AICc W stats F model/exponential	0.7 Max Min ±25%		1.0 Max Min 0%
--	--	---------------------------	--	-------------------------

Input OK - ready to fit...

Operations

Send to shelf

Fit (re-fit) this model

Reset/reject this model

Send to shelf

Fit (re-fit) this model

Reset/reject this model





MIRDfit interface

Regression toolbox

Fit all selected (quickfit)

Reset all

Regression model filter

- Biexponential (%IA = %blood at t = 0)
- Biexponential (%IA = 100 at t = 0)
- Biexponential (rapid clearance & slow clearance phases)
- Biexponential (rapid clearance phase & physical decay)**
- Biexponential (uptake & clearance phases)
- Monoexponential
- Monoexponential (physical decay only)

Fit metric comparison legend

R ²	0.7	1.0
AICc	Max	Min
W _{AICc}	Min	Max
F _{model/trapezoidal}	±25%	0%

S N M

Monoexponential (physical decay only)

$\%IA(t) = A_1 e^{-\lambda_{phys}t}$

Fit setup

Parameter A_1
Initial/guess: 6.889688

Fit results

Parameter A_1
Value: 0.34188
Std. Err.: 0.018

Summary output

TIAC	3.87E-01	[h]
TIAC Std. Err.	2.04E-02	[h]
TIAC%CV	5.3%	
R ²	0.1449	
AICc	30.27	
W _{AICc}	0.0%	
F _{model/trapezoidal}	0.23	
Alerts	None.	

Actions

Send to shelf

Fit (re-fit) this model

Reset/reject this model

Weighted residual

Time [h]

%IA (effective)

Data ● **Fit** — **Initial guess** - - -

Figure description: This panel shows the fit setup for a monoexponential model with physical decay only. The initial guess for A_1 is 6.889688. The fit results show A_1 is 0.34188 with a standard error of 0.018. The summary output table includes TIAC, TIAC Std. Err., TIAC%CV, R², AICc, W_{AICc}, and F_{model/trapezoidal}. The weighted residual plot shows a decreasing trend over time, and the effective IA plot shows a rapid decay phase followed by a slow decay phase.

Monoexponential

$\%IA(t) = A_1 e^{-(\lambda_1 + \lambda_{phys})t}$

Fit setup

Parameter A_1 λ_1
Initial/guess: 6.889688 0.000884

Fit results

Parameter A_1 λ_1
Value: 8.18102 0.0433
Std. Err.: 0.32601 0.00115

Summary output

TIAC	1.57E+00	[h]
TIAC Std. Err.	4.31E-02	[h]
TIAC%CV	2.7%	
R ²	0.9911	
AICc	14.13	
W _{AICc}	99.9%	
F _{model/trapezoidal}	0.94	
Alerts	None.	

Actions

Send to shelf

Fit (re-fit) this model

Reset/reject this model

Weighted residual

Time [h]

%IA (effective)

Data ● **Fit** — **Initial guess** - - -

Figure description: This panel shows the fit setup for a monoexponential model including both uptake and physical decay. The initial guess for A_1 is 6.889688 and λ_1 is 0.000884. The fit results show A_1 is 8.18102 and λ_1 is 0.0433. The summary output table includes TIAC, TIAC Std. Err., TIAC%CV, R², AICc, W_{AICc}, and F_{model/trapezoidal}. The weighted residual plot shows a decreasing trend over time, and the effective IA plot shows a rapid decay phase followed by a slow decay phase.

Biexponential (rapid clearance phase & physical decay)

$\%IA(t) = A_1 e^{-(\lambda_1 + \lambda_{phys})t} + A_2 e^{-\lambda_{phys}t}$

Fit setup

Parameter A_1 λ_1 A_2
Initial/guess: 3.444844 0.02652 3.444844

Fit results

Parameter A_1 λ_1 A_2
Value: 8.33385 0.04578 0.06577
Std. Err.: 0.33622 0.00152 0.02451

Summary output

TIAC	1.60E+00	[h]
TIAC Std. Err.	4.44E-02	[h]
TIAC%CV	2.8%	
R ²	0.9964	
AICc	29.56	
W _{AICc}	0.0%	
F _{model/trapezoidal}	100.0%	
Alerts	0.96	
None.		

Actions

Send to shelf

Fit (re-fit) this model

Reset/reject this model

Weighted residual

Time [h]

%IA (effective)

Data ● **Fit** — **Initial guess** - - -

Figure description: This panel shows the fit setup for a biexponential model with rapid clearance and physical decay. The initial guess for A_1 is 3.444844, λ_1 is 0.02652, and A_2 is 3.444844. The fit results show A_1 is 8.33385, λ_1 is 0.04578, and A_2 is 0.06577. The summary output table includes TIAC, TIAC Std. Err., TIAC%CV, R², AICc, W_{AICc}, and F_{model/trapezoidal}. The weighted residual plot shows a decreasing trend over time, and the effective IA plot shows a rapid decay phase followed by a slow decay phase.

Send to shelf

Fit (re-fit) this model

Reset/reject this model

Send to shelf

Fit (re-fit) this model

Reset/reject this model

Send to shelf

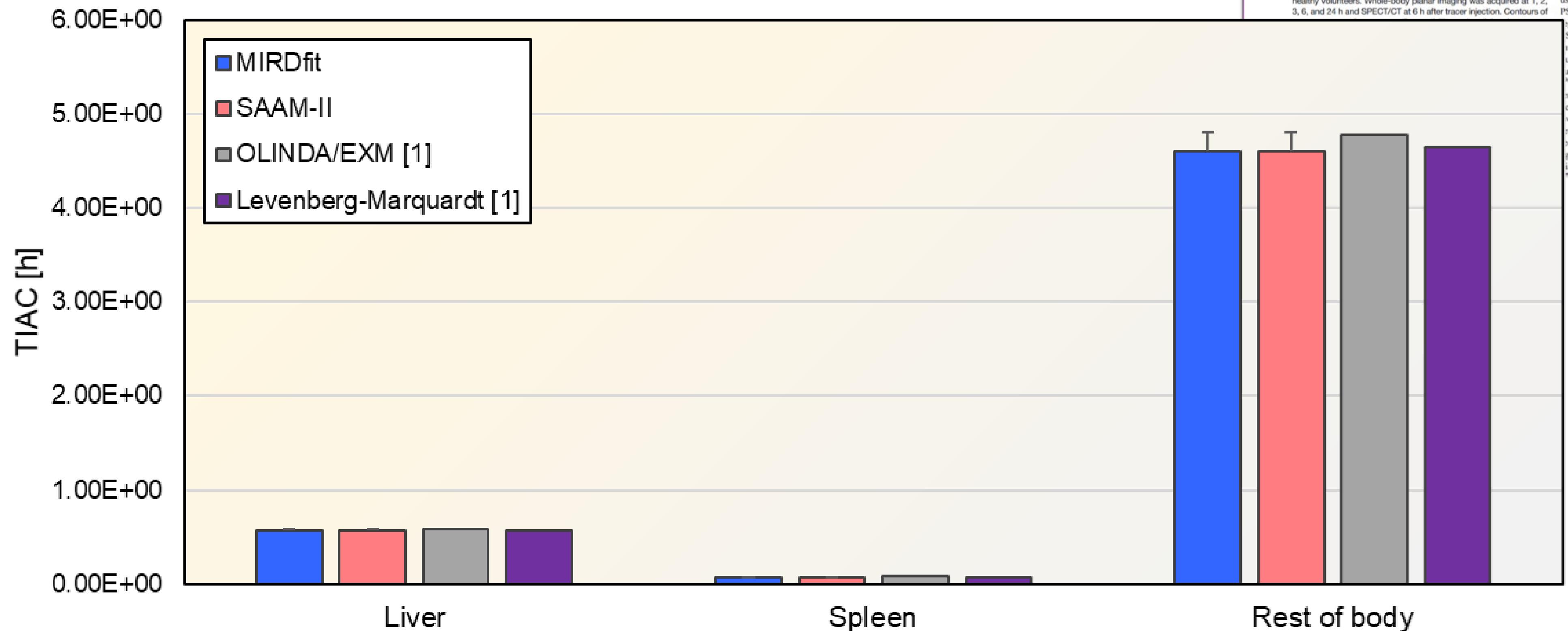
Fit (re-fit) this model

Reset/reject this model

MIRDfit performance

- Within 1% agreement with SAAM-II in testing

Fit results for 99m Tc-PSMA I&S



Radiation Dosimetry of 99m Tc-PSMA I&S: A Single-Center Prospective Study

Szabolcs Urbán¹, Catherine Meyer^{2,3}, Magnus Dahlbom^{2,3}, István Farkas¹, Gábor Sipka¹, Zsuzsanna Besenyi¹, Johannes Czernin^{2,4,5}, Jeremie Calais^{*2-5}, and László Pávics^{*1}

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99m Tc-PSMA I&S is a prostate-specific membrane antigen (PSMA) tracer that can be used for planar and SPECT/CT γ -imaging and radioguided surgery. The primary aim of this study was to estimate the dosimetry of 99m Tc-PSMA I&S using a hybrid method (sequential γ -planar imaging and 1 single SPECT/CT) in healthy volunteers. The secondary aim was to depict the tracer biodistribution and tumor-to-background ratios (TBGs) in patients with prostate cancer (PCa). **Methods:** Dosimetry of 99m Tc-PSMA I&S was investigated in 4 healthy volunteers. Whole-body planar imaging was acquired at 1, 2, 3, 6, and 24 h and SPECT/CT at 6 h after tracer injection. Contours of

regions in which access to PSMA PET imaging is limited, 99m Tc-PSMA imaging can also enable radioguided surgery (RGS) with intraoperative γ -detection. PSMA-targeted RGS can help and urologists to detect PCa lymph node (LN) metastasis during surgery. Different 99m Tc-PSMA compounds have been developed. 99m Tc-Mas3- γ -nal-kSub-KuE (99m Tc PSMA I&S) is a nonpeptidic compound derived from the PSMA I&T precursor that can be obtained with a reliable kit-labeling procedure (2). Previous work has shown the utility of 99m Tc-PSMA I&S for RGS in large retrospective cohorts with improved treatment outcome (3,4). As a required step for further translation and approval by regulatory agencies, the primary objective of this study was to provide the radiation dosimetry analysis of 99m Tc-PSMA I&S in healthy volun-



Presentation overview

- Topics
 - Standardization and error propagation via MIRDsoft.org tools
 1. MIRDcalc: TIAC->dosimetry
 2. MIRDfit: data points -> TIAC
 - Perspectives on organ level dosimetry
 - Perspectives on standardization
 - Closing remarks



Anatomy of a dinner preparation workflow

- Several methods have been proposed

Complexity spectrum



Anatomy of a dosimetry workflow

- Several methods have been proposed

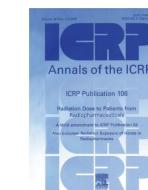
Complexity spectrum



Lookup table.

One size for all

- ICRP 53, ICRP 80, ICRP 106,...
- Regulatory approved package insert for one size fits all radiopharmaceutical therapies



Organ level dosimetry

MIRD formalism

Voxel level dosimetry

(in house or commercial software)

Point kernel

Monte Carlo

Patient specific – derived from patient biodistribution



Perspectives on organ level dosimetry



Lookup table.
One size for all

Organ level dosimetry

- Patient specific
- Integrates millions of points of data (image voxels)
- Can be accomplished with accessible (free) software
- Flexible to accommodate variations in equipment, resources
- Relatively easier to standardize
- Models systems beyond voxels (blood, bone marrow, whole body, tumor)
- Output organ level dose (current standard for contextualizing dosimetry)

Voxel level dosimetry
(in house or commercial software)

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Perspectives on standardization

- Not all dose is equal
 - Equipment, expertise, resources, and values differ across the global healthcare community
 - Community is best served if we account for these differences
- 20th century models of innovation are often siloed. 20th century dosimetry practices have failed to establish footing in clinical workflows
- 21st century brings new opportunities to standardize dosimetry across the community.
- We have examples of simple, standardized dosimetry protocols working in close fields: Y90 microspheres
 - Accessible
 - Treatment has been a gateway for:
 - Innovation in image based dosimetry,
 - Nuclear medical physicists working and billing for dosimetry
 - Provides example of simple models evolving (BSA -> partition)



Perspectives on standardization

- RPT-TEC models should be derived from diverse data from multiple regions/centers if it is to be relevant across regions/centers
- Can follow trial designs to collect data from multiple centers and process with central workflow



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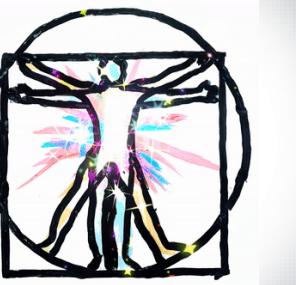


Closing remarks

- MIRDsoft tools are created in an effort to create easy to use, easily accessible, vetted software for dosimetry community.
 - Inspired by existing tools (Olinda, IDAC, Nukfit,)
 - Vetted and endorsed by the SNMMI MIRD committee
 - MIRDcalc and MIRDfit provide (new) option for integrating uncertainty into dosimetry calculations
 - We have the tools, how best to use them (ie input)
 - MIRDcalc pamphlet coming soon (final stages of submission)
 - Pamphlet + benchmark/software comparison compendium with 1000+ cases
 - MIRDcalc not currently FDA approved
-
- MIRDsoft.org and/or MIRDsoft.org model can be used for collecting dosimetry/adverse event data in the future...



Acknowledgements



- **MSKCC MIRDcalc team**

- Lukas Carter
- Juan Ocampo Ramos
- Harry Marquis

- **UF MIRDcalc team**

- Wes Bolch (MIRDcalc co-PI)
- Edmond Olguin
- Justin Brown
- Bonnie President
- Sean Domal
- Yitian Wang
- Laura Dinwinddie
- Robert Dawson
- Cameron Kofler
- Jared Baggett
- Wyatt Smithier

- **Contributing colleagues**

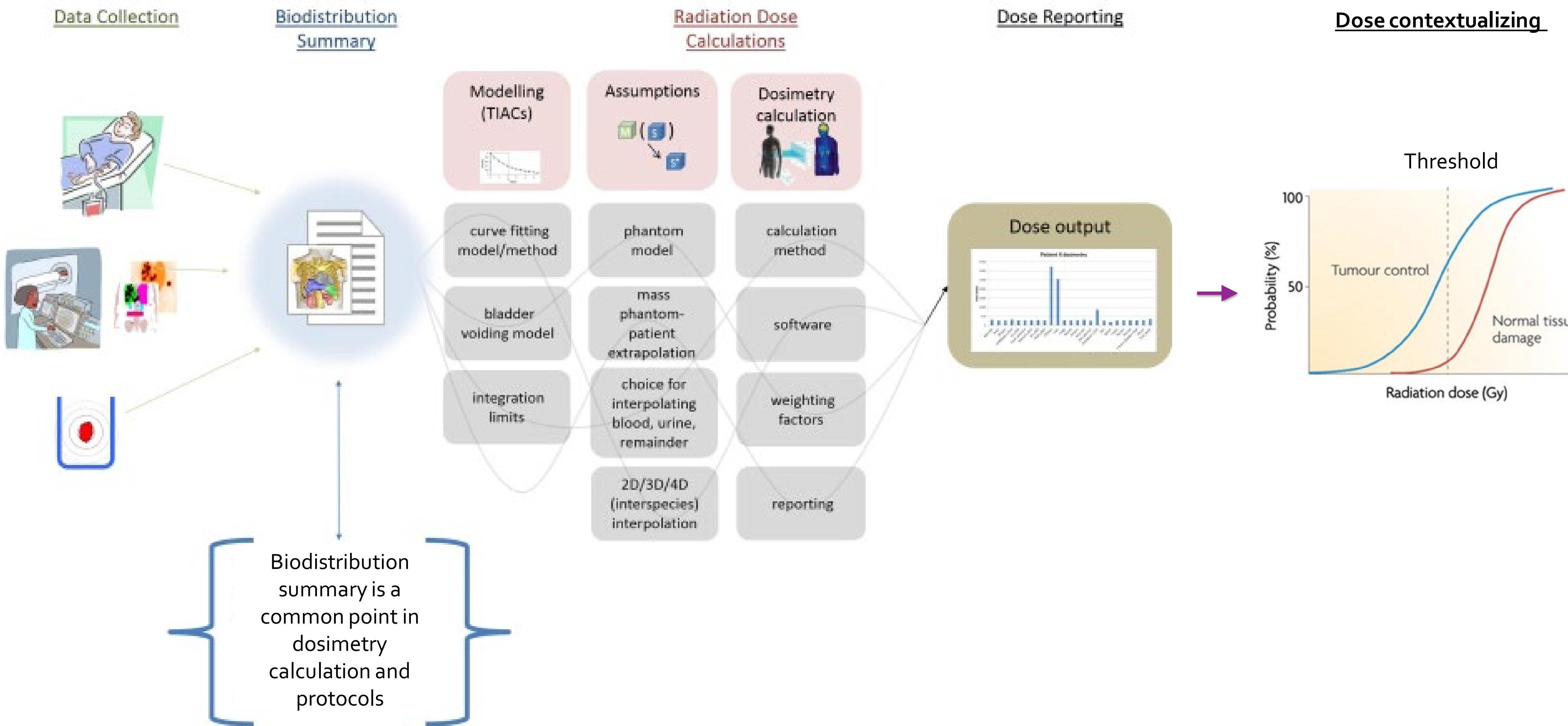
- Pat Zanzonico (MSK)
- John Humm (MSK)
- Joe O'Donoghue (MSK)
- Daniel Lafontaine (MSK)
- Jazmin Schwartz (MSK)
- Simone Krebs (MSK)
- Heiko Schoder (MSK)
- Derek Jokisch (Marion)
- Roger Howell (Rutgers)
- Darrell Fisher (Versant)
- George Sgouros (Hopkins)
- Bill Erwin (MD Anderson)
- Bud Wendt (MD Anderson)

- **SNMMI MIRD committee**

- **Funding:**

- NIH U01 EB028234 (Kesner/Bolch)
- NIH P30 CA008748 (MSK)

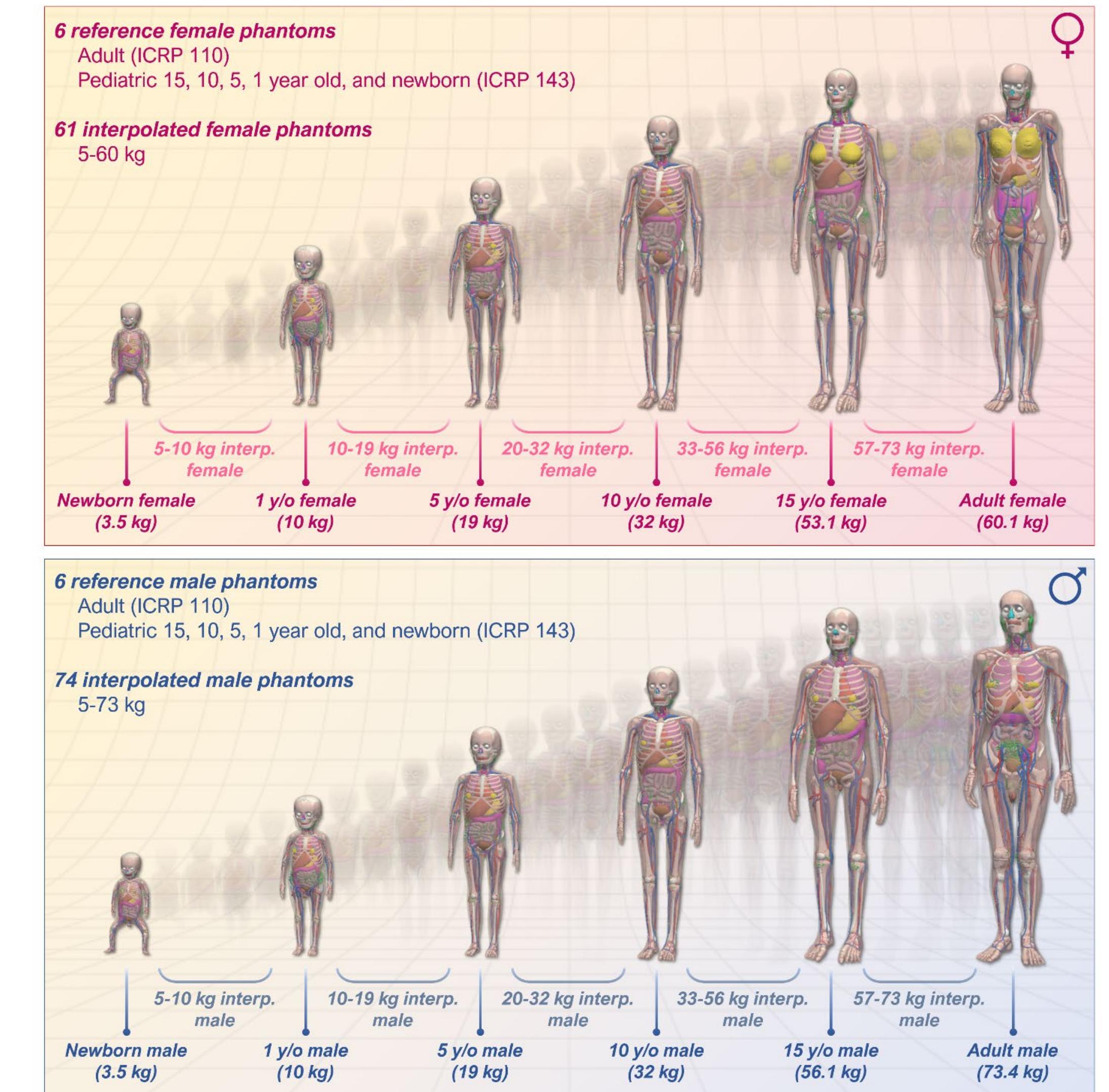
Overview of internal dosimetry protocols





MIRDcalc phantoms

- Phantom models
 - ICRP reference phantoms (reports 110 and 143)
 - Newborn (m/f) year old (m/f)
 - 1 year-old (15/f) year old (m/f)
 - 5 year-old (Ad/f/t) (m/f)
 - Additional source regions generated
 - Heart contents
 - Major blood vessels
 - MIRDcalc interpolation feature
 - Organ masses interpolated linearly relative to whole body mass
 - S values interpolated log-log

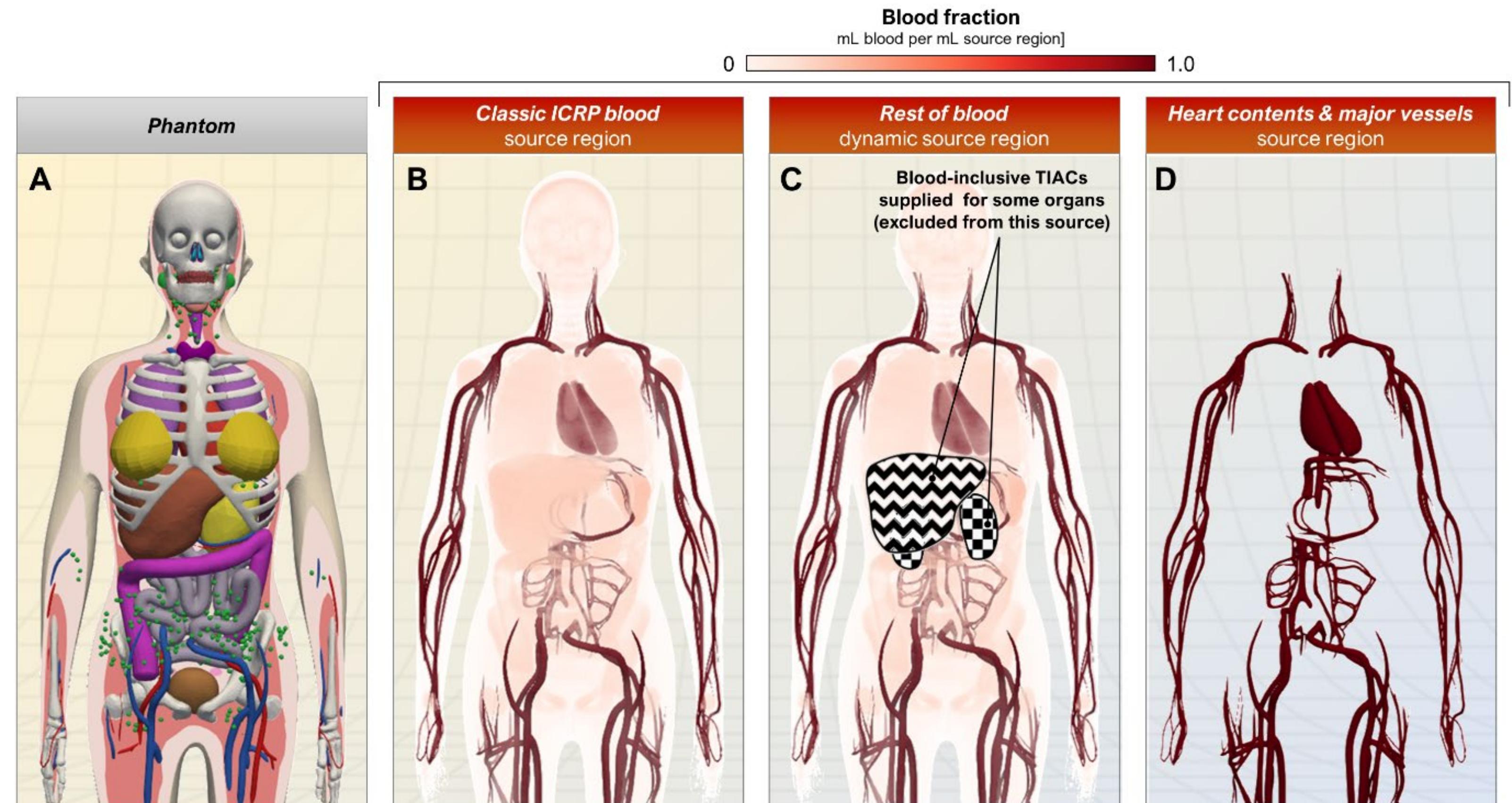


Visualization of MIRDcalc phantom library



MIRDcalc blood model

- User can use multiple blood source regions

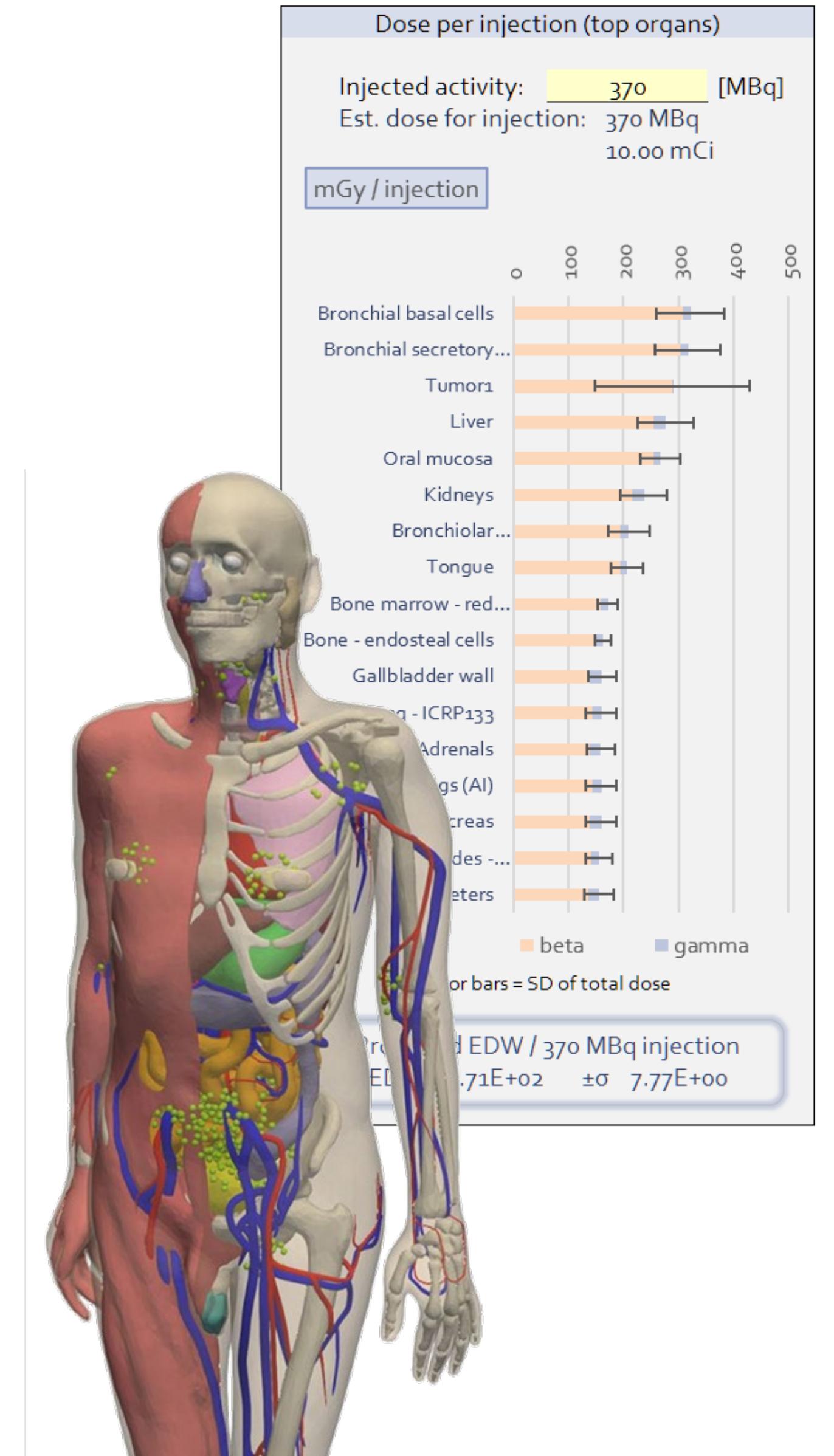


MIRDcalc blood source options



MIRDcalc tumor dosimetry

- MIRDcalc tumor dose model
 - Spherical tumor model
 - Olguin et. al, PMB, 2020
 - Model parameters
 - Sphere volume (optional uncertainty)
 - TIAC (optional uncertainty)
 - Tissue composition (bone/soft tissue)
 - Dosimetry semi-integrated with organ dosimetry
 - Self dose (no cross dose)
 - Integrated TIAC accounting



MIRDcalc validation

- Example comparison with other software
 - Fluorodeoxyglucose (18F-FDG), ICRP 128 distribution

