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Photodynamic Therapy

- First clinical approval 1993 (Bladder Cancer)
- 16 K publication 1993-now containing 'clinical' or 'patient' as keywords
- 70 ongoing clinical trials in oncology (superficial ~45 vs. interstitial ~20)
- Approval based on administered Photosensitizer [mg/kg] and delivered optical energy [J cm⁻²], [J cm⁻¹], or [J]
- Benefit of repeat therapies particular for immune response.

Trans Surface Accessible Illumination

- Large surface area illumination
- Photon density gradient given by $1/\mu_{\text{eff}}$
 - Possible Photosensitizer gradient
- Lack of eloquent structures
 - Exception Bladder
- Good healing









 Curative therapy – First line therapy

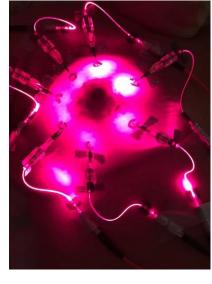
Deep Seated or Solid Target

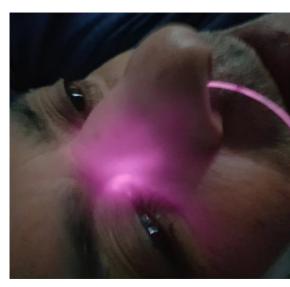
- Source number volume dependent
- Photon density gradient given by $1/\pi r~\mu_{eff}$ or $1/\pi r^2~\mu_{eff}$
- Eloquent structures can be present
- Necrotic tissue removal

- Salvage therapy
 - Exception Prostate, GBM

Interstitial Photodynamic Therapy Current Treatment planning concepts

• Drug dose [mg/kg], Energy delivered [J or J/cm or J/cm²], heuristic Light source placement, spacing determined by tissue and PDT wavelength





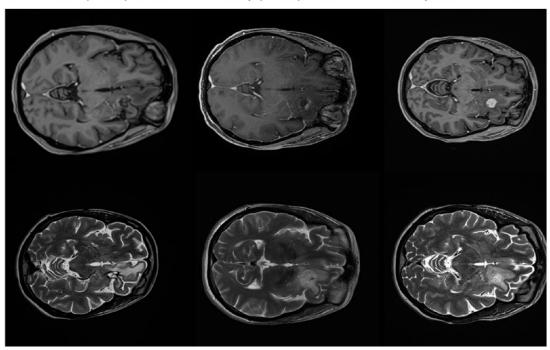


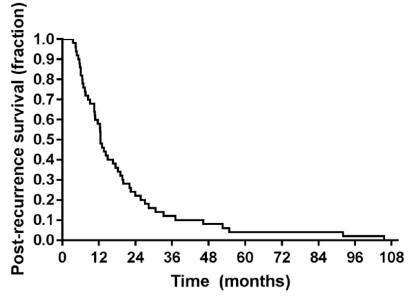


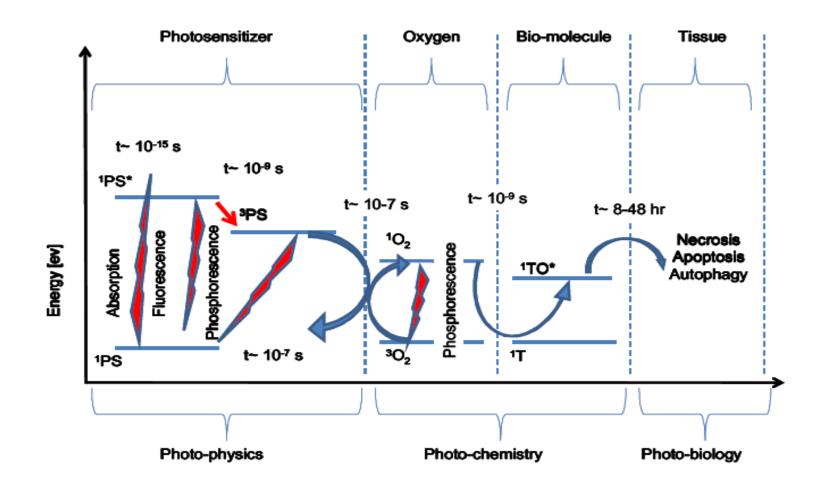
• Validating Light source placement, dose definition and progress monitoring.

Example iPDT for Malignant Glioma

18 months postoperative MRI Early postoperative MRI Preoperatove MRI







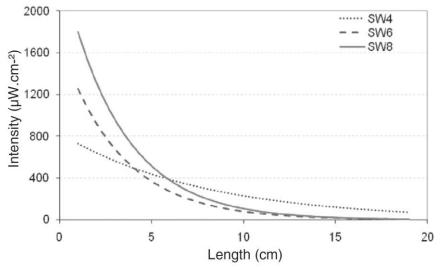
Monitoring

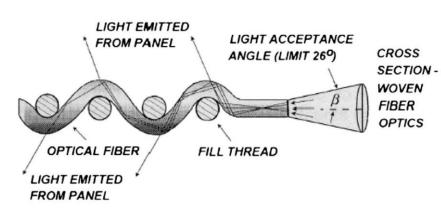
Planning

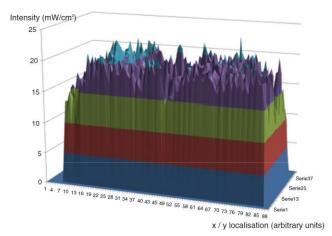
Photodynamic Therapy needs temporal and spatial overlap of Photosensitizer, Oxygen and Photons with appropriate quantum energy

Light delivery Outer surface





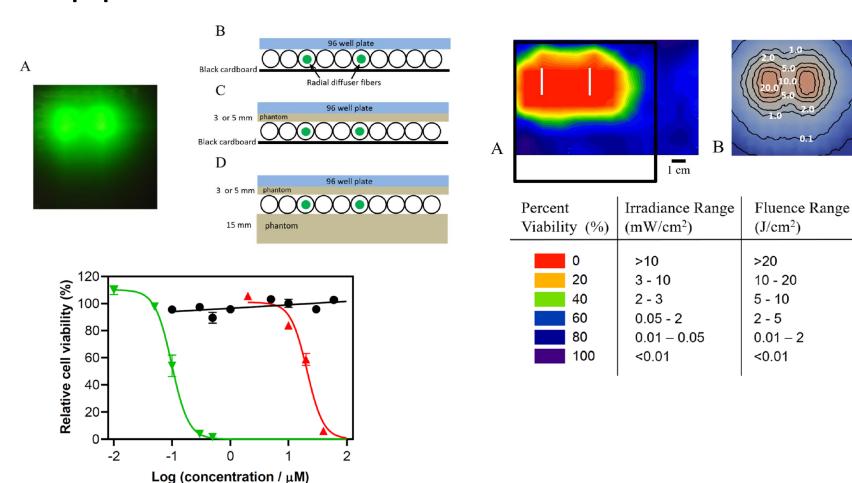








Surgical access allowing surface PDT applications



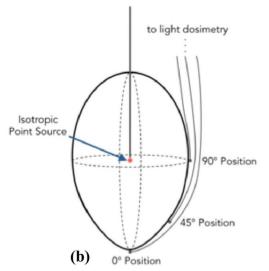
Fluence Dosimetry in Cavity

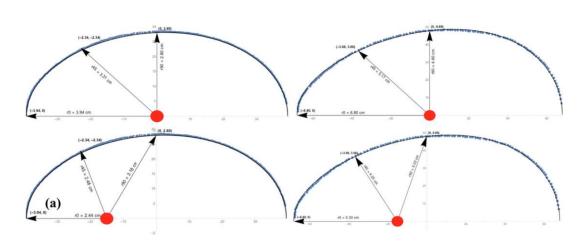
$$\phi_{dir} = \frac{S}{4\pi r^2} A(\theta),$$

$$\int_{0}^{\pi} A(\theta) d\Omega / 4\pi = 1$$

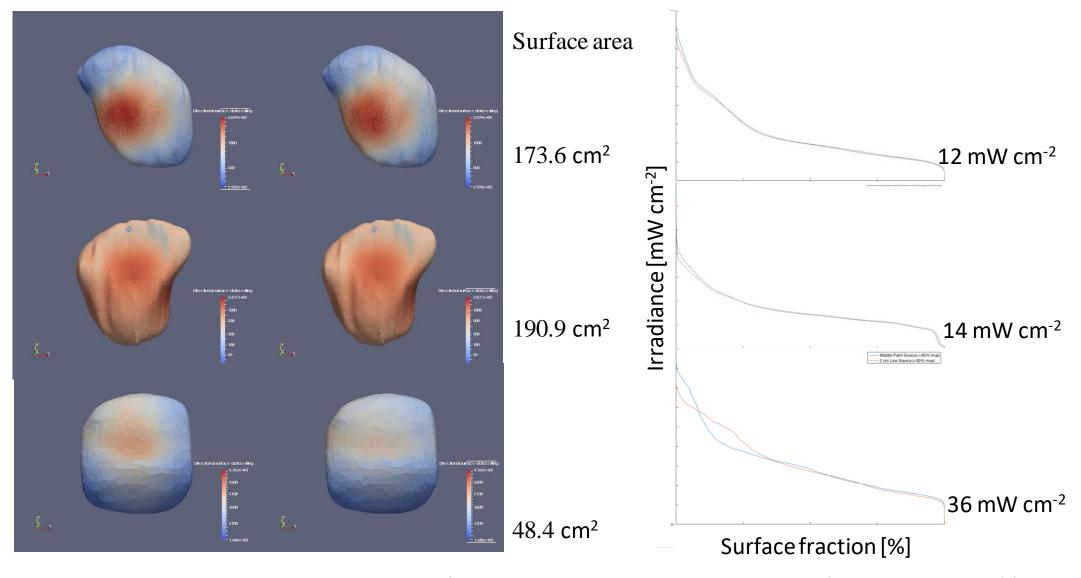
$$\phi_{sc} = \frac{4S}{A_s} \cdot R_d \cdot (1 + R_d(1 - f) + R_d^2(1 - f)^2 + \dots) = \frac{4S}{A_s} \cdot \frac{R_d}{1 - R_d(1 - f)^2}$$





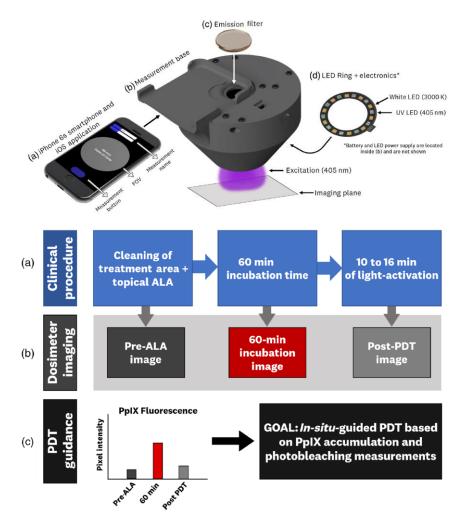


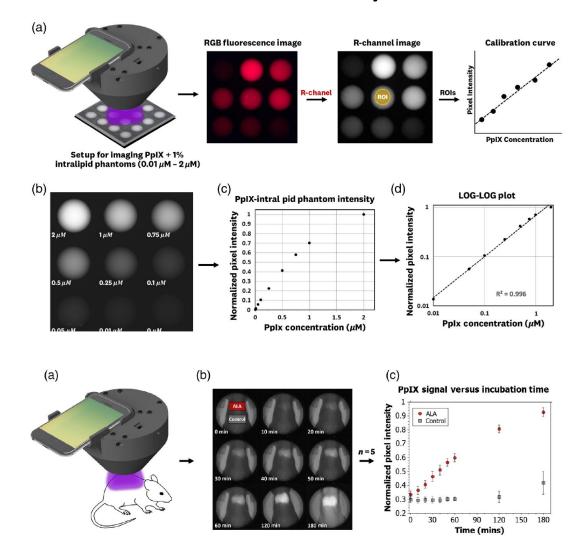
Bladder Surface Irradiance homogeneity



Lilge L, Wu J, Xu Y, et al. Minimal required PDT light dosimetry for Non-Muscle Invasive Bladder cancer. Journal of Biomedical Optics, 25(6), 068001, (2020)

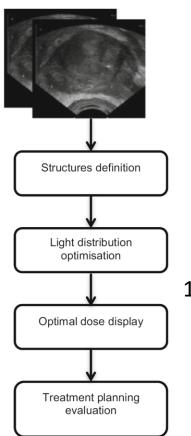
Photosensitizer based Surface Dosimetry





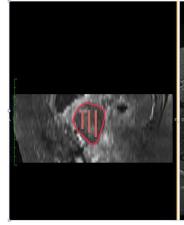
iPDT treatment plan based on Templates and

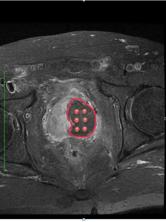
Clinical Imaging



Transrectal ultrasound (TRUS) images acquired in the operating room

Targets and organs at risk (urethra rectum).







150 mW/cm for 1333 sec

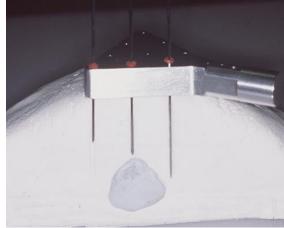
The action model (Based on light penetration model)

 $1/\mu_{eff}$ 3.8 -7.1 mm@ 753 nm

Fibres, positions on the grid and lengths of the diffusing tip

Light distribution





Images Courtesy of Colin Hopper. UCL UK

N. Betrouni et al. Vascular targeted photodynamic therapy with TOOKAD® Soluble (WST11) in localized prostate cancer: efficiency of automatic pre-treatment planningLasers Med Sci (2017) 32:1301–1307 DOI 10.1007/s10103-017-2241-7

Speed

Anatomical Accuracy

Usability

CUDA-MCML (GPLv3) 2010 /Alerstam et al./

MCX (GPLv3) 2009 /Fang, Boas/

MCX-CL (GPLv3) 2012/2018 /Yu, Kaeli, Fang/

FullMonteCUDA (BSD*) 2019 /Young-Schultz et al/

> MMCL (GPLv3) 2020 /Fang, Yan/

MCML (BSD) 1993 /Wang,Jacques,Prahl/

mcxyz (BSD) ?? /Jacques/ tMCimg (BSD) 2002 /Boas et al./

YMC3D (MIT) 2018 /Bjorgan et al./

MOSE (non-free) 2004

MMC/iMMC/DMMC (GPLv3) 2010 /Fang,Yan,/

MCtet (unknown) 2018

SVMC (part of MCX) 2021 /Yan, Fang/ FullMonteSW (BSD*)
2018
/Cassidy et al./

VirtualPhotonics(BSD)
2010
/Cuccia et al./

MCOnline (non-free) 2013

MCmatlab (GPLv3) 2018 /Marti et al./

> ValoMC (MIT) 2019 /Leino et al./

Speed in performing light propagation models

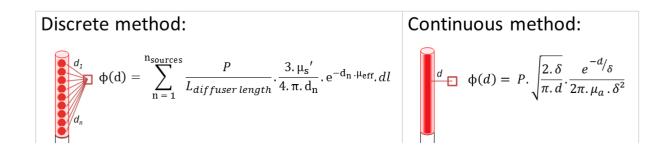


Table 2: Relative deviation computed for different source lengths; the mean is computed for all distances.

	Discrete	e method	Continuous method		
Source length (mm)	Mean (SD)	Max deviation	Mean (SD)	Max deviation	
10	4.94% (2.38%)	8.00%	11.16% (7.18%)	19.10%	
15	4.70% (4.41%)	15.38%	24.31% (6.40%)	32.77%	
20	1.05% (1.23%)	4.24%	27.40% (3.94%)	30.56%	
30	1.70% (1.71%)	5.56%	28.61% (4.21%)	32.65%	
40	0.91% (1.46%)	5.00%	30.12% (5.14%)	37.50%	
50	1.89% (1.34%)	4.49%	29.48% (4.77%)	35.96%	

C. Dupont et al. 5-ALA Photodynamic Therapy in Neurosurgery, Towards the Design of a Treatment Planning System: A Proof of Concept. Innovation and Research in BioMedical engineering, Elsevier Masson, 2016, 10.1016/j.irbm.2016.11.002. hal-01403867

What Dose is Monitored or Planed for ?

• Power or photon density, photosensitizer bleaching.

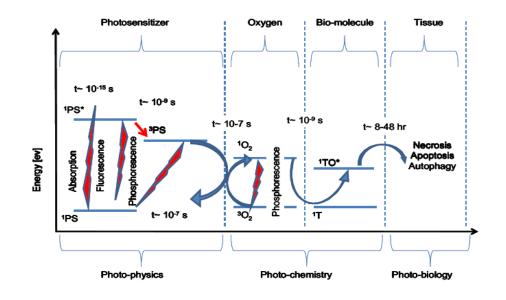
• Outcome determining parameter, direct or indirect reporter.

Direct dose

Triplet dose

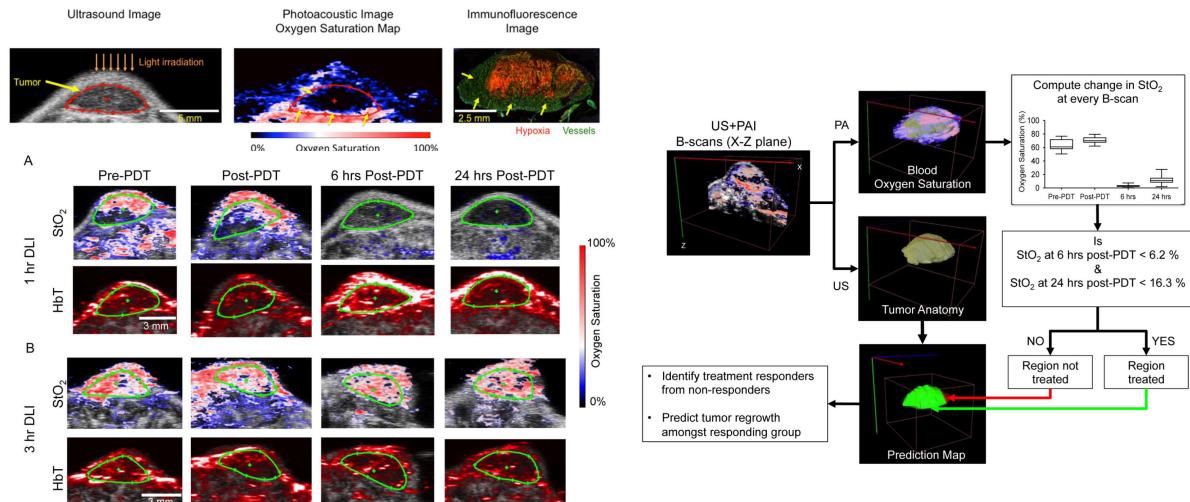
Implicit dose

Explicit dose



Biological response

Imaging based PDT Efficacy monitoring

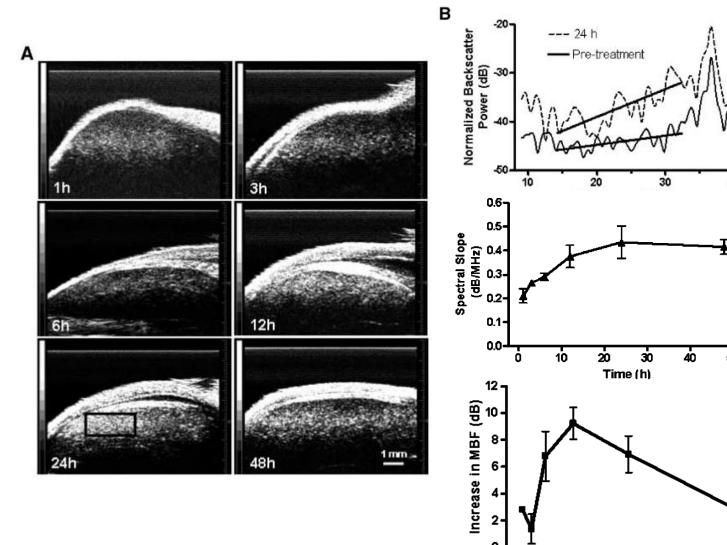


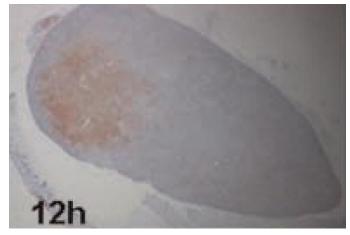
Mallidi S et al. Prediction of Tumor Recurrence and Therapy Monitoring Using Ultrasound-Guided Photoacoustic Imaging Theranostics 2015; 5(3): 289-301. doi:10.7150/thno.10155

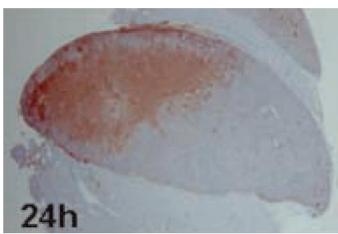
Apoptosis Imaging based PDT Monitoring

10

Time (h)

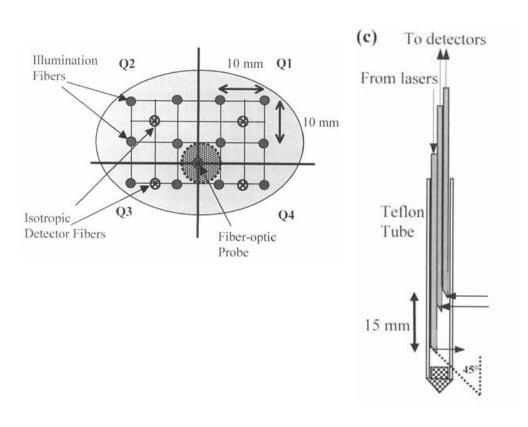


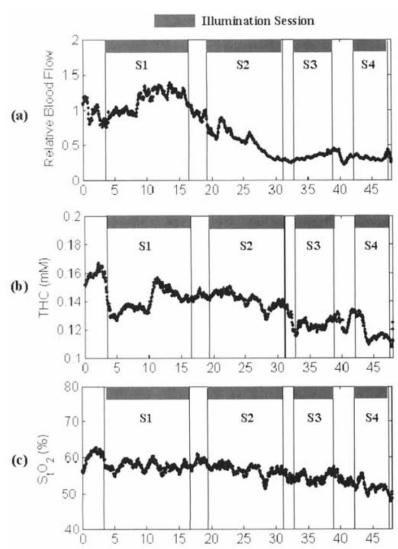




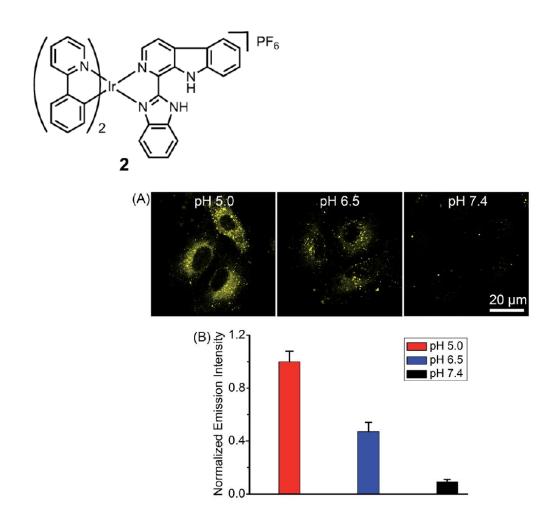
Banihashemi B et al. Ultrasound Imaging of Apoptosis in Tumor Response: Novel Preclinical Monitoring of Photodynamic Therapy Effects. Cancer Res 2008;68(20):8590–6

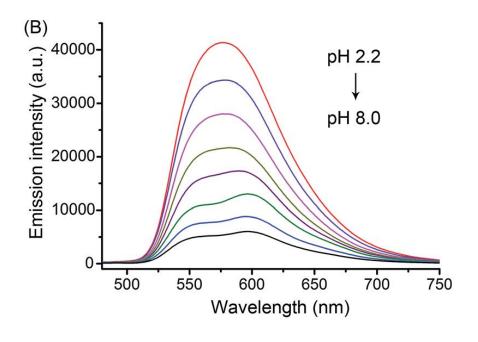
Monitoring Hemodynamic changes online



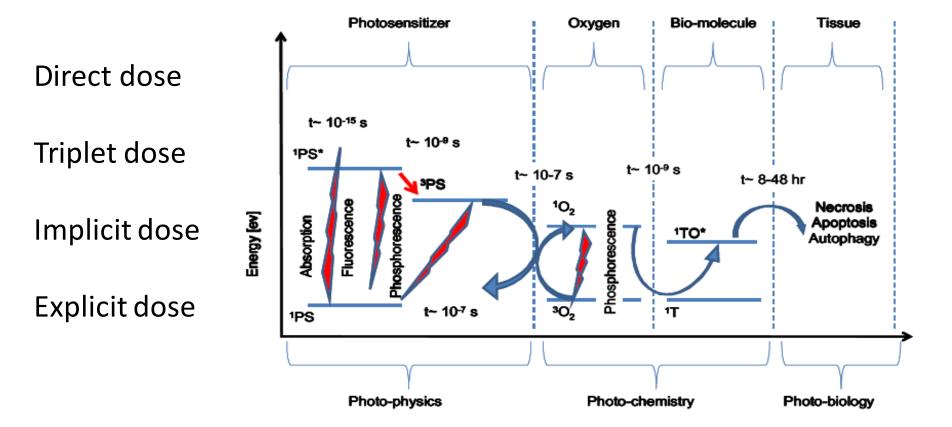


Photosensitizer and Sensor Combination





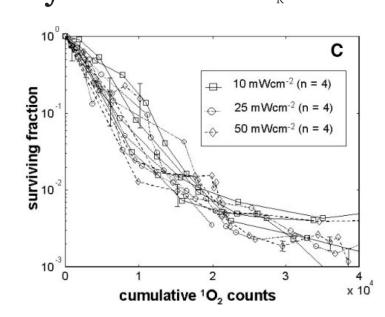
What Dose is Monitored or guides Planning?



Indirect Singlet Oxygen Dose Metric

$$[{}^{1}\mathrm{O}_{2}](t) = \mathrm{N}\sigma[\mathrm{S}_{0}]\Phi_{\mathrm{D}}\frac{\tau_{\mathrm{D}}}{\tau_{\mathrm{T}} - \tau_{\mathrm{D}}}[\exp(-t/\tau_{\mathrm{T}}) - \exp(-t/\tau_{\mathrm{D}})]$$

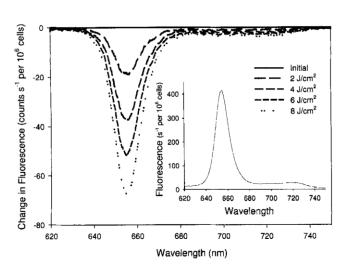
$$\int L_{1270}(t) dt = \frac{N\sigma[S_0]\Phi_{\scriptscriptstyle D}\tau_{\scriptscriptstyle D}}{\tau_{\scriptscriptstyle R}}$$

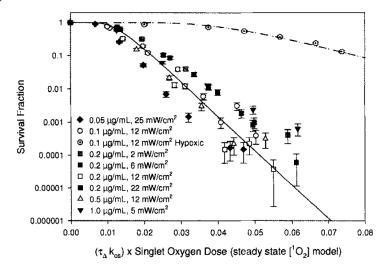


$$[{}^{1}O_{2}](t) = N\sigma[S_{0}]\Phi_{D}\frac{\tau_{D}}{\tau_{T} - \tau_{D}}[\exp(-t/\tau_{T}) - \exp(-t/\tau_{D})] \qquad [{}^{1}O_{2}] = S_{\Delta}\Phi_{t}I_{a(t)}\left(\frac{[{}^{3}O_{2}]}{k_{p}/k_{ot} + [{}^{3}O_{2}]}\right)\left(\frac{1}{k_{d} + k_{oa}[A]}\right)$$

Dose =
$$\frac{1}{\tau_{\Delta}k_{os}\delta}([S_0]_0 - [S_0](T)) \quad \text{for } [S_0] \ll \delta$$

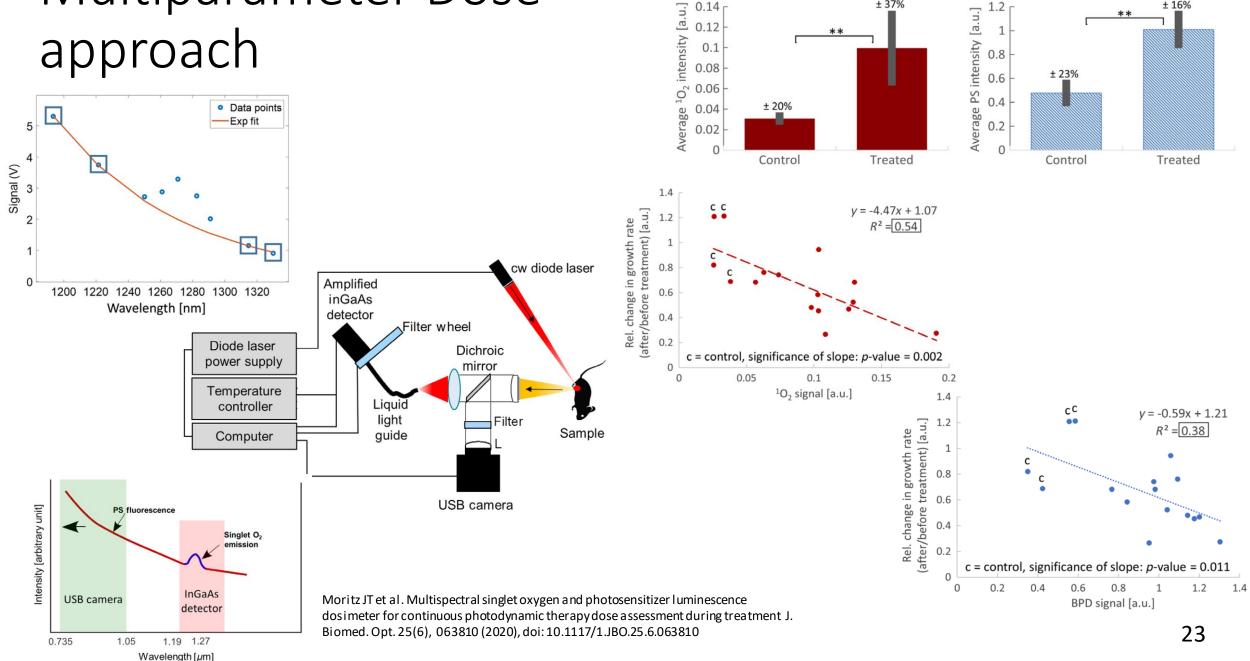
Dose =
$$\frac{1}{\tau_{\Delta}k_{os}} \ln\left(\frac{[S_0]_0}{[S_0](T)}\right)$$
 for $[S_0] \gg \delta$.





Dysart JS et al. Calculation of Singlet Oxygen Dose from Photosensitizer Fluorescence and Photobleaching During mTHPC Photodynamic Therapy of MLL Cells? Photochemistry and Photobiology, 2005, 81: 196-205

Multiparameter Dose approach



0.14

0.12

0.1

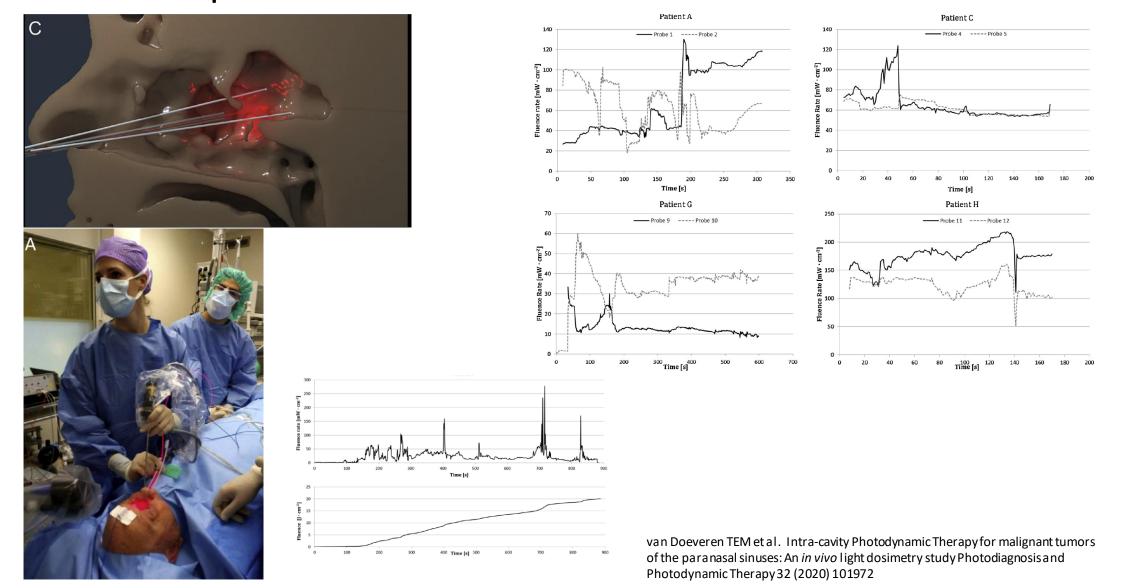
0.08

± 37%

0.8

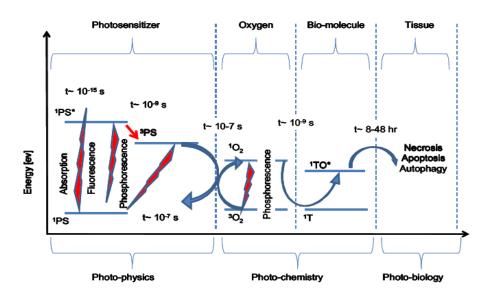
± 16%

Intra-operative fluence rate measurements

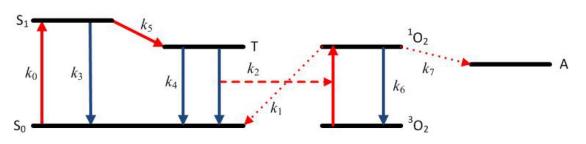


Intermezzo

- Treatment progress monitoring appears feasible
 - Φ (r, t), [PS] (r, Φ), [${}^{3}O_{2}$] (r, Φ), [${}^{1}O_{2}$] (Φ)
 - What is the target PDT dose?
- Was the original plan for PDT
 - Optimal light delivery (outcome)
 - Personalized
 - What is the target PDT dose?



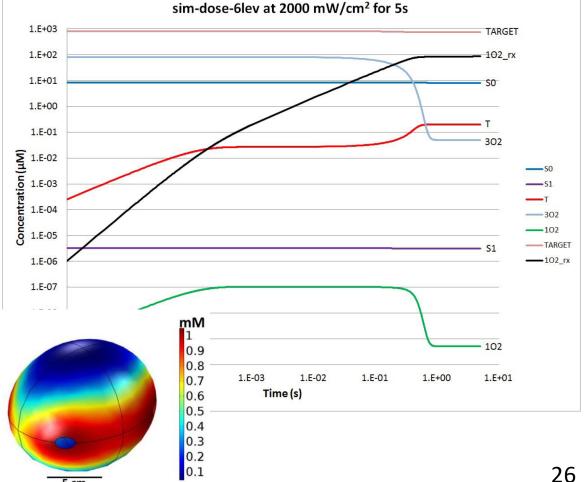
Photochemical Dosimetry



$$\begin{split} \frac{d[S_0]}{dt} &= -k_0[S_0] - k_1[^1O_2]([S_0] + \delta) + k_2[T][^3O_2] + k_3[S_1] + k_4[T] \\ &\frac{d[S_1]}{dt} = -(k_3 + k_5)[S_1] + k_0[S_0] \\ &\frac{d[T]}{dt} = -k_2[T][^3O_2] - k_4[T] + k_5[S_1] \\ &\frac{d[^3O_2]}{dt} = -S_\Delta k_2[T][^3O_2] + k_6[^1O_2] + g\left(1 - \frac{[^3O_2]}{[^3O_2](t=0)}\right) \\ &\frac{d[^1O_2]}{dt} = -k_1([S_0] + \delta)[^1O_2] + S_\Delta k_2[T][^3O_2] - k_6[^1O_2] - k_7[A][^1O_2] \\ &\frac{d[A]}{dt} = -k_7[A][^1O_2] \end{split}$$

Bees on K et al. Overview of computational simulations for PDT treatments based on optimal choice of singlet oxygen Optical Methods for Tumor Treatment and Detection: Mechanisms and Techniques in Photodynamic Therapy XXVI, 100470R 2017 doi: 10.1117/12.2252552

Simulation Parameters for Photofrin®						
$k_0 = (1.895 \text{ x}10^{-2}) * (\text{fluence rate}) \text{ s}^{-1}$	$\delta(\mu M) = 0$					
$k_1 = 1.2 \times 10^5 \mu \text{M}^{-1} \text{s}^{-1}$	$S_{\Delta} = 1$					
$k_2 = 100 \ \mu \text{M}^{-1} \ \text{s}^{-1}$	$[S_0 (t=0)] = 8.5 \mu M$					
$k_3 = 2 \times 10^7 \text{ s}^{-1}$	$[^{3}O_{2} (t=0)] = 83 \mu M$					
$k_4 = 1250 \text{ s}^{-1}$	$[A (t=0)] = 830 \mu M$					
$k_5 = 8 \times 10^7 \text{ s}^{-1}$	$g = 1.0 \mu \text{M/s}$					
$k_6 = 1 \times 10^6 \text{ s}^{-1}$	$\sigma(S_0 - S_1) = 5.978 \times 10^{-18} \text{ cm}^2$					
$k_7 = 2.6 \text{x} 10^6 \mu\text{M}^{-1} \text{s}^{-1}$						



Dosie™device

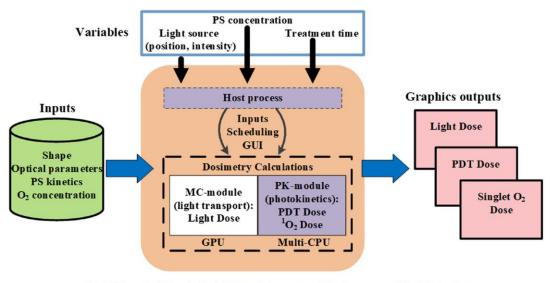
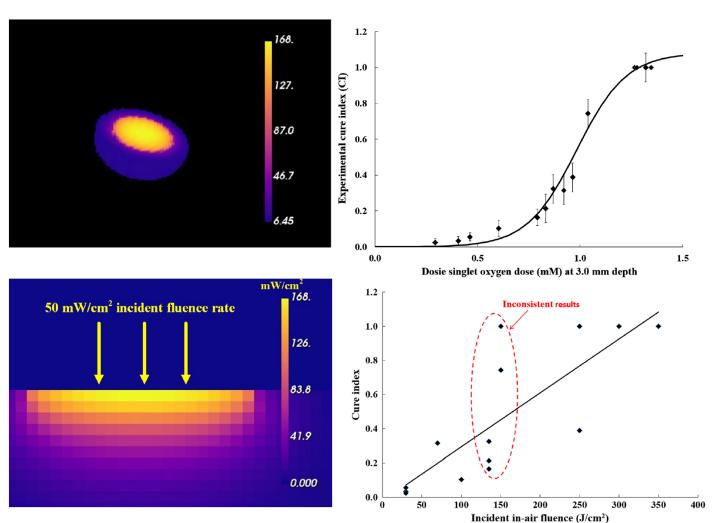


Fig. 1 Schematic diagram for Dosie[™] integrated computer and software computational device. The computer includes both a GPU and a multicore CPU. The software first performs a MC simulation of light transport that is followed by a PK simulation.

$$\frac{d[S_0]}{dt} + \left(\xi \sigma \frac{\phi([S_0] + \delta)[^3 O_2]}{[^3 O_2] + \beta}\right)[S_0] = 0$$

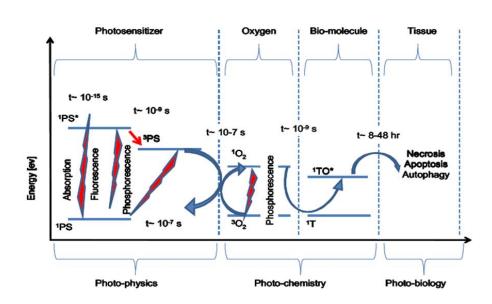
$$\frac{d[^1 O_2]_{rx}}{dt} - \left(\xi \frac{\phi[S_0][^3 O_2]}{[^3 O_2] + \beta}\right) = 0$$



Bees on KW et al. Validation of combined Monte Carlo and photokinetics imulations for the outcome correlation analysis of benzoporphyrin derivative-mediated photodynamic therapy on mice. J. Bi omed. Opt. 24(3), 035006 (2019), doi: 10.1117/1.JBO.24.3.035006.

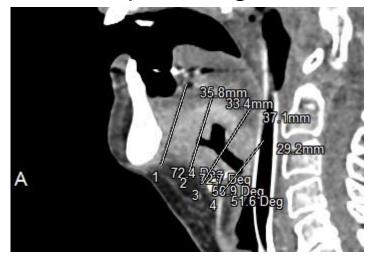
What do we know about the inputs for iPDT optimization/personalization

- Shape
 - Medical imaging helps
- Optical Properties
 - Spatially resolved online monitoring
- PS Kinetics
 - fMRI BOLD
- Oxygen Concentration
 - DRS, fMRI BOLD
- Biological response of different tissues
 - Very little.



Free Space Light source placement

Offline planning

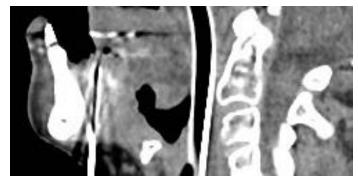


Planned sagittal section

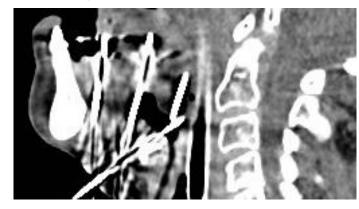
Rigid registration and non deformation allows for robotic needle insertion

Images Courtesy of Colin Hopper. UCL UK

Online Monitoring CT



First in-plane needle insertion



Final sagittal section appearance

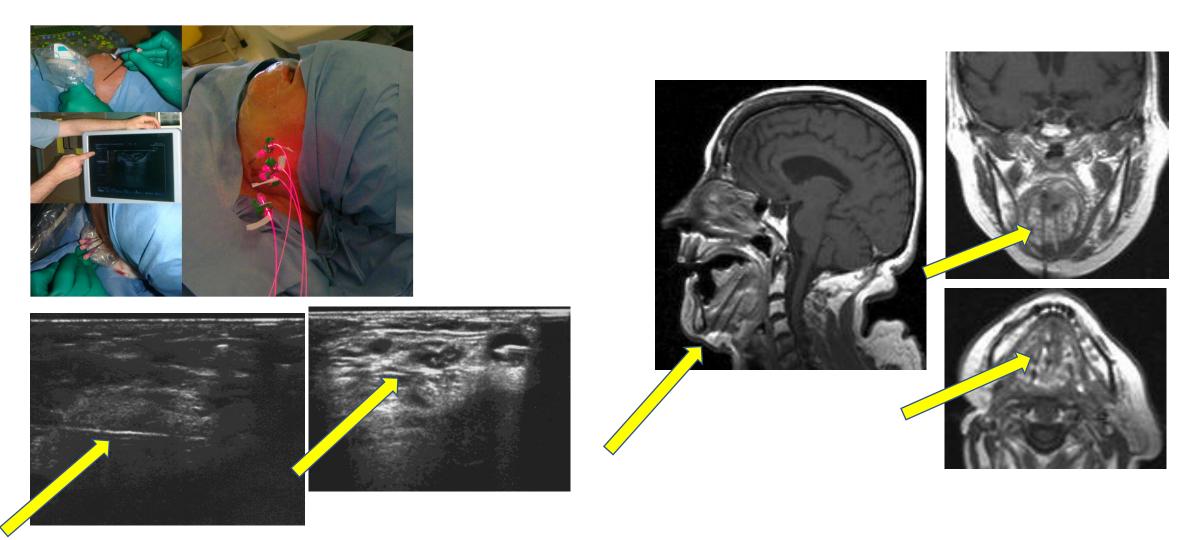
Online Monitoring CT



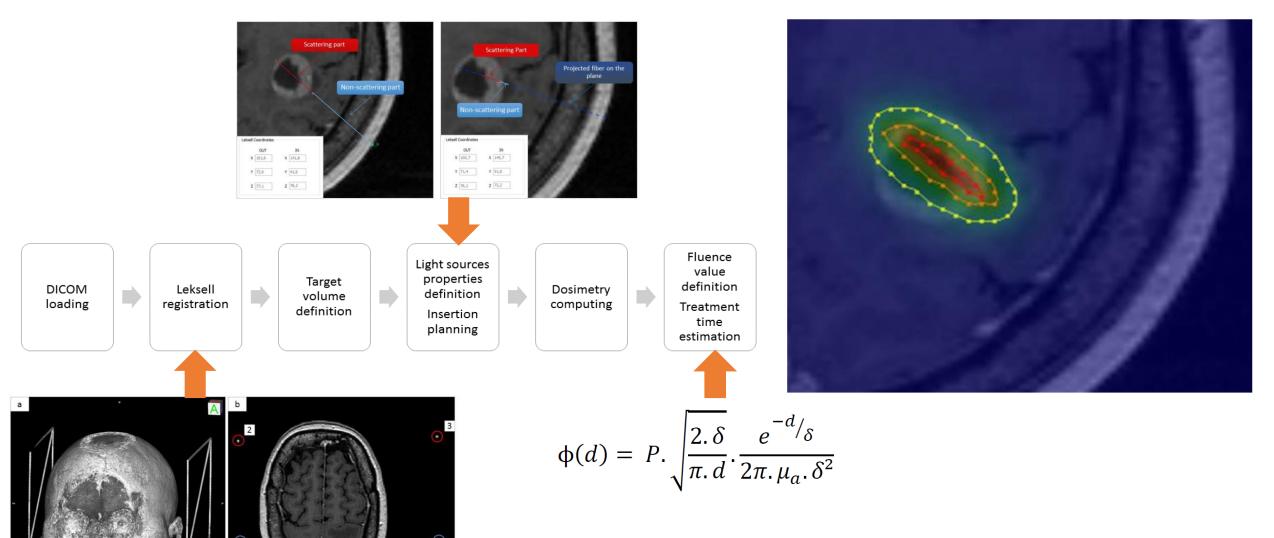
Online Monitoring US



Free Space Light source placement



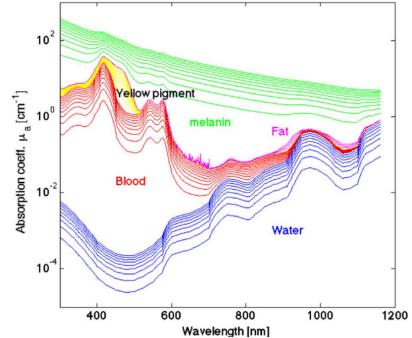
Robotic controlled light source placement

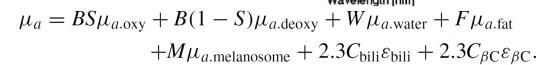


Planning in the absence true μ_a , μ_s and g

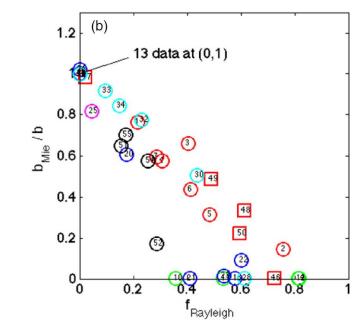
Table 1. Parameters specifying thereduced scattering coefficient of tissues: $a = \mu'_{s500 \text{ nm}}$, such that $\mu'_{s}(\lambda) = a(\lambda/500 \text{ nm})^{-b}$, equation (1); $aa = \mu'_{s500 \text{ nm}}$, such that $\mu'_{s}(\lambda) = aa(f_{\text{Ray}}(\lambda/500 \text{ nm})^{-b} + f_{\text{Mie}}(\lambda/500 \text{ nm})^{-b})$, equation (2); and $f_{\text{Mie}} = 1 - f_{\text{Ray}}$. (na = not available.)

#	$a (\mathrm{cm}^{-1})$	b	a' (cm ⁻¹)	f_{Ray}	$b_{ m Mie}$	Ref.	Tissue
Skin							
1	48.9	1.548	45.6	0.22	1.184	Skin	Anderson et al 1982
2	47.8	2.453	42.9	0.76	0.351	Skin	Jacques 1996
3	37.2	1.390	42.6	0.40	0.919	Skin	Simpson et al 1998
4	60.1	1.722	58.3	0.31	0.991	Skin	Saidi et al 1995
5	29.7	0.705	36.4	0.48	0.220	Skin	Bashkatov et al 2011
6	45.3	1.292	43.6	0.41	0.562	Dermis	Salomatina et al 2006
7	68.7	1.161	66.7	0.29	0.689	Epidermis	Salomatina et al 2006
8	30.6	1.100	na	na	na	Skin	Alexandrakis et al 2005
Brain							
9	40.8	3.089	40.8	0.00	3.088	Brain	Sandell and Zhu 2011
10	10.9	0.334	13.3	0.36	0.000	Cortex (frontal lobe)	Bevilacqua et al 2000
11	11.6	0.601	15.7	0.53	0.000	Cortex (temporal lobe)	Bevilacqua et al 2000
12	20.0	1.629	29.1	0.81	0.000	Astrocytoma of optic nerve	Bevilacqua et al 2000

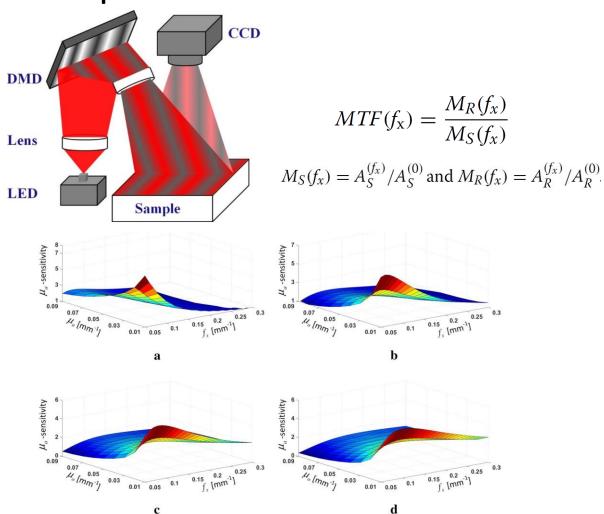




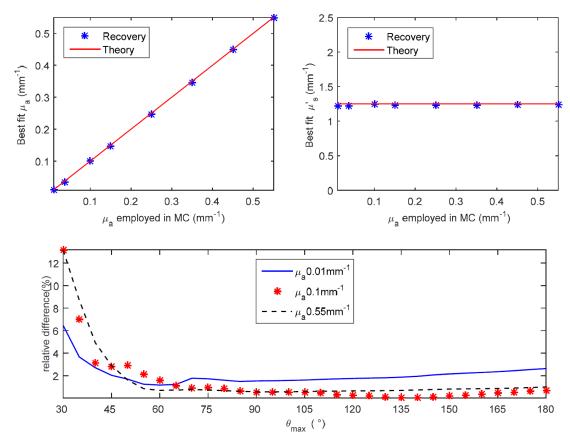
- S HGb oxygen saturation of mixed arterio-venous vasculature
- B average blood volume fraction $(f_{v,blood})$
- W water content $(f_{v.\text{water}})$
- Bili bilirubin concentration (C(M))
- β C β -carotene concentration (C(M))
- F fat content $(f_{v.fat})$
- M melanosome volume fraction ($f_{v.\text{melanosome}}$), or alternatively the molar concentration of melanin monomers (C(M)).



Optical Parameter Quantification

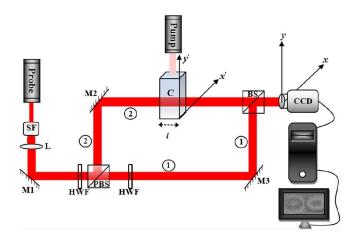


 $\phi_{n}\left(\mu_{a},\mu_{s},r\right) = \frac{1}{4\pi} \sum_{i=1}^{\frac{N+1}{2}} \upsilon_{i}\left(\mu_{a},\mu_{s}\right) G_{i}\left(\omega_{i}\left(\mu_{a},\mu_{s}\right)\right) l_{n}\left(\upsilon_{i}\left(\mu_{a},\mu_{s}\right)\right) \kappa_{n}\left(\upsilon_{i}\left(\mu_{a},\mu_{s}\right)r\right)$



Liu L et al. Interstitial optical parameter quantification of turbid medium based on CW radiance measurements. SPIE 10024, Optics in Health Care and Biomedical Optics VII, 1002436 2016; doi: 10.1117/12.2246053

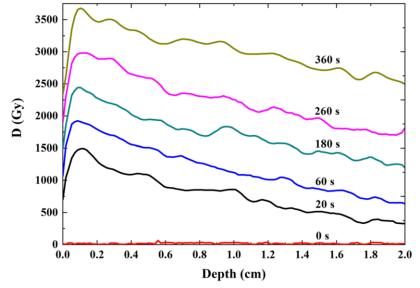
Experimental Methods for Light Dose Profile

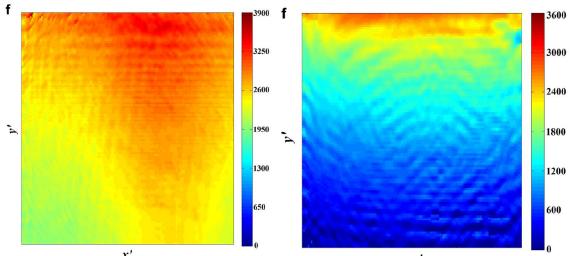


$$D = c\Delta T$$

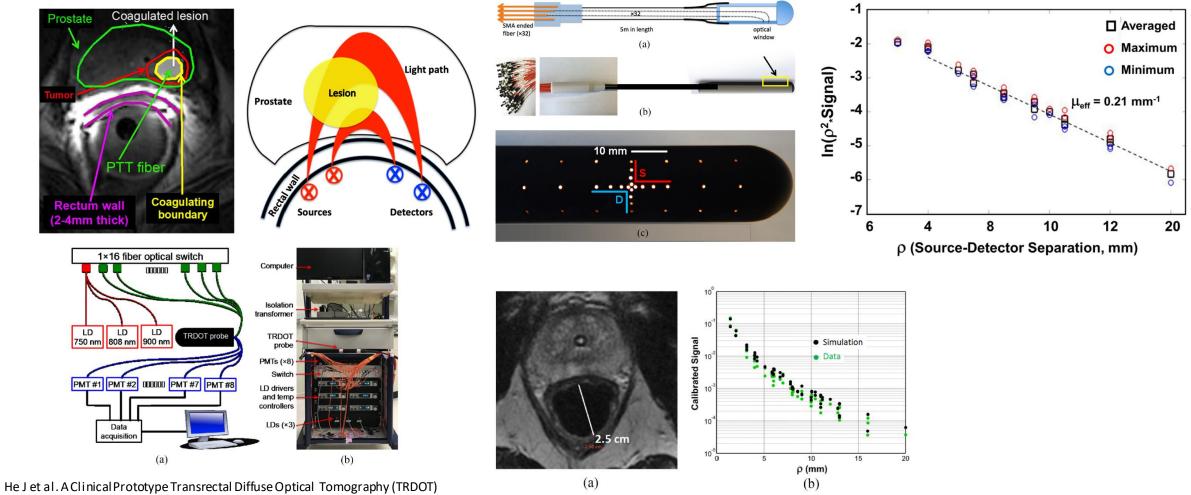
$$D = c \left[\frac{\sqrt{5.998 - 0.644 \left(b + \frac{\Delta \phi}{kl} \times 10^5\right)} - a}{0.322} \right]$$

$$\phi = \frac{D\kappa\alpha}{2c}$$



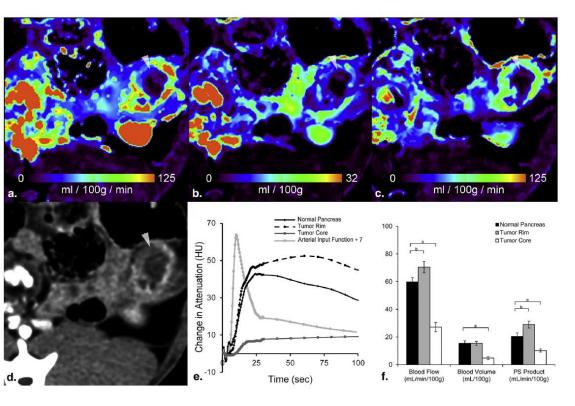


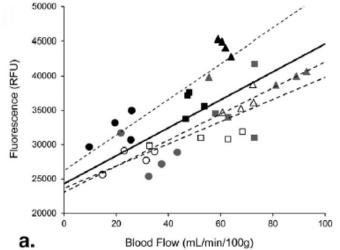
Spatially Resolved Optical Properties Recovery

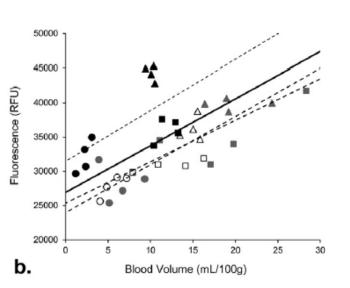


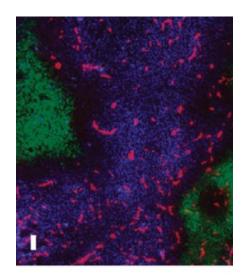
He J et al. A Clinical Prototype Transrectal Diffuse Optical Tomography (TRDOT) System for *In vivo* Monitoring of Photothermal Therapy (PTT) of Focal Prostate Cancer IEEE TRANSACTIONS ON BIOMEDICAL ENGINEERING, VOL. 67, NO. 7, 2020

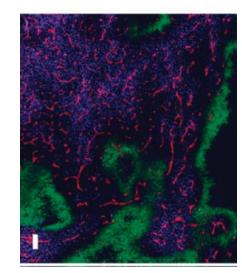
Blood Volume And Flow Was Shown To Correlate With PS Uptake.











Tissue response to PDT

Threshold model(s)

$$PDT_{Dose} = 2.3 \varepsilon [PS]_{\lambda} \phi_{\lambda} (d, \mu_a, \mu_s)$$

$$PDT_{Dose} > T_{target}$$
 $PDT_{Dose} < T_{host}$

$$\frac{T_{target}}{[PS]_{target}} \frac{\phi(0)}{\phi(d)} < \frac{T_{host}}{[PS]_{host}}$$

$$\frac{T_{t \operatorname{arg}et}}{[PS]_{t \operatorname{arg}et}} \frac{\phi(r_{host})}{\phi(r_{t \operatorname{arg}et})} < \frac{T_{host}}{[PS]_{host}}$$

Oxygen Consumption Model

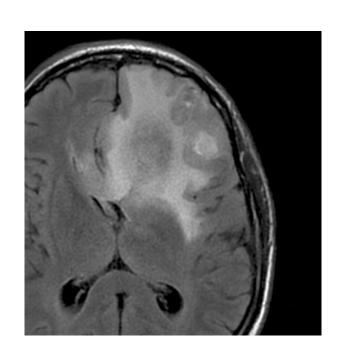
$$\frac{d[T_1]}{dt} = BF_{ex}[PS] - (BF_{ex} + k_p + k_{ot}[O_2])[T_1]$$

$$\frac{\mathrm{d}[^{3}\mathrm{O}_{2}]}{\mathrm{d}t} = D_{\mathrm{cell}}\left(\frac{1}{r^{2}}\frac{\mathrm{d}}{\mathrm{d}r}r^{2}\frac{\mathrm{d}[^{3}\mathrm{O}_{2}]}{\mathrm{d}r}\right) - \Gamma(r,t) \qquad \qquad \Gamma_{\mathrm{met}} = \Gamma_{\mathrm{met}}^{0}\left(\frac{[^{3}\mathrm{O}_{2}]}{\mathrm{k}_{\mathrm{met}} + [^{3}\mathrm{O}_{2}]}\right)$$

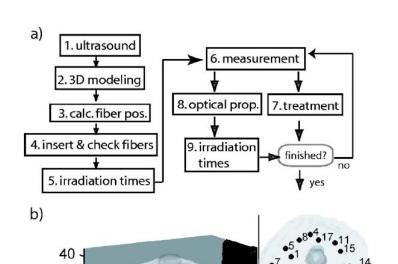
$$B = \frac{k_{isc}}{k_{isc} + k_f} \frac{\sigma}{hv} = \frac{\phi_t \sigma}{hv}$$

$$\Gamma_{\text{met}} = \Gamma_{\text{met}}^0 \left(\frac{[^3 O_2]}{k_{\text{met}} + [^3 O_2]} \right)$$

$$\Gamma_{ ext{PDT}} = \Gamma_{ ext{PDT}}^0 arepsilon_{ ext{PDT}} rac{[ext{PS}]}{[ext{PS}]_0}$$



Clinical implementation Prostate



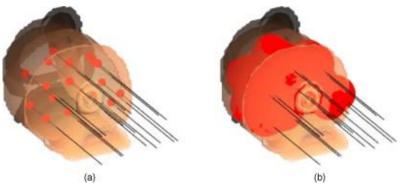
depth (mm)

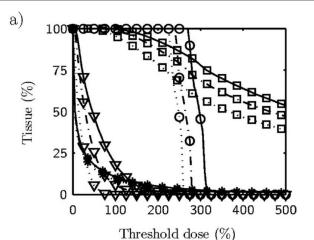
50 40 30 20 10

y(mm)









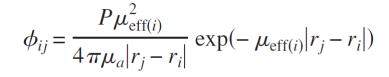
Rectum

Urethra

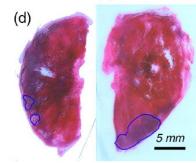
0.1

Normal

1e - 8



$$\ln(\phi_{ij}|r_j - r_i|) = \ln\left(\frac{P\mu_{\text{eff}(i)}^2}{4\pi\mu_a}\right) - \mu_{\text{eff}(i)}|r_j - r_i|.$$

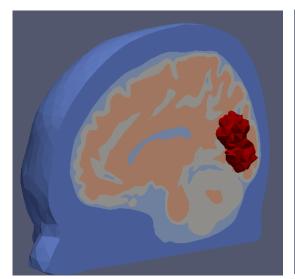


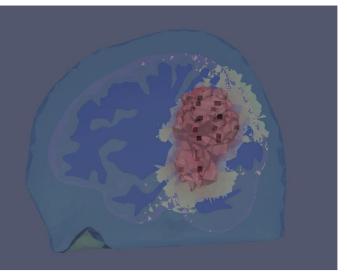
Parameter

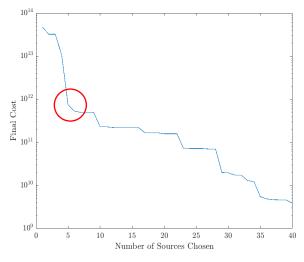
Prostate

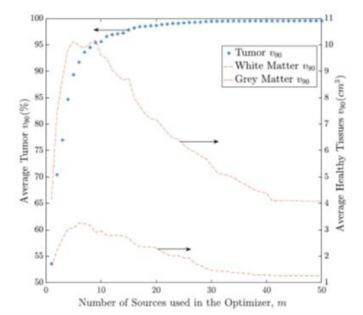
0 30 x (mm) 40

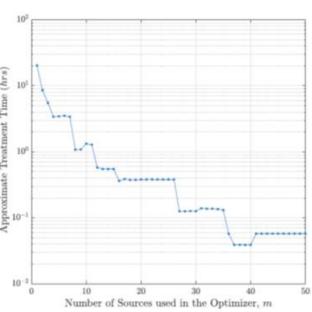
Treatment Planning using PDT-SPACE







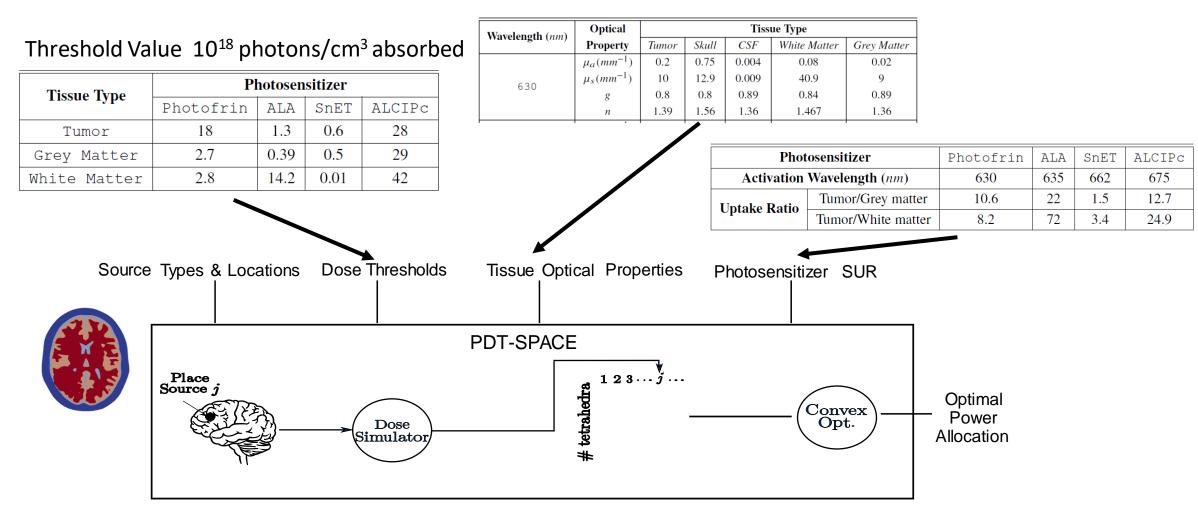




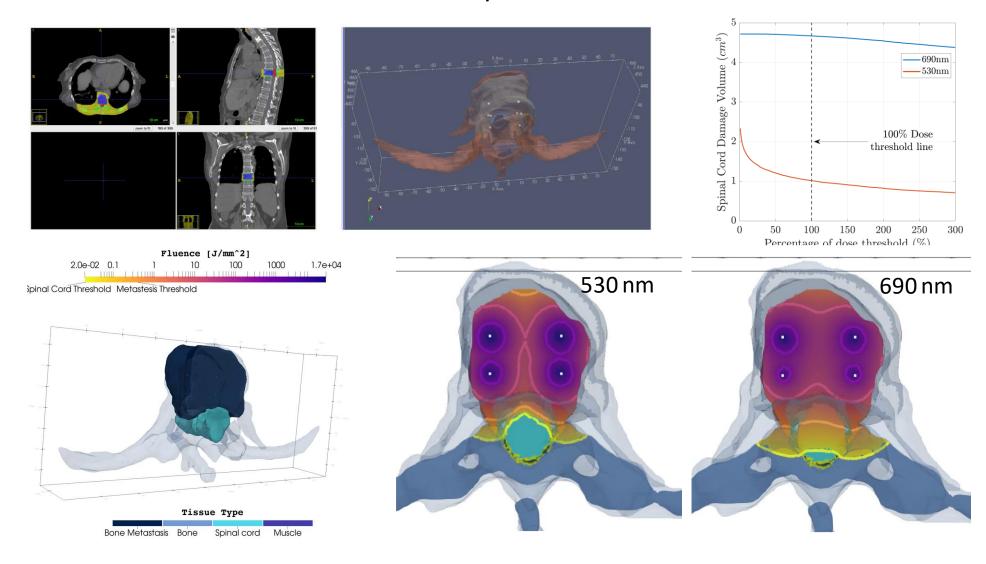
A.-A. Yassine, W. Kingsford. Y. Xu, et al. Automatic interstitial photodynamic therapy via convex optimization *Biomed. Opt. Express*, 2018

PDT-SPACE FOR TREATMENT PLANNING

 $Dose_{PDT} = 2.3 \varepsilon (\lambda) [Photosensitizer] \varphi(d)$



Outcome Predictions for Spinal Metastasis



Approach

Heuristic:

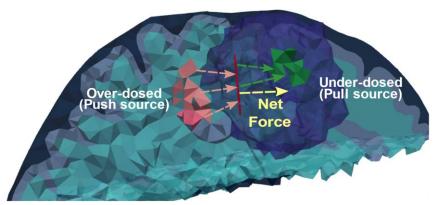
Parallel sources.

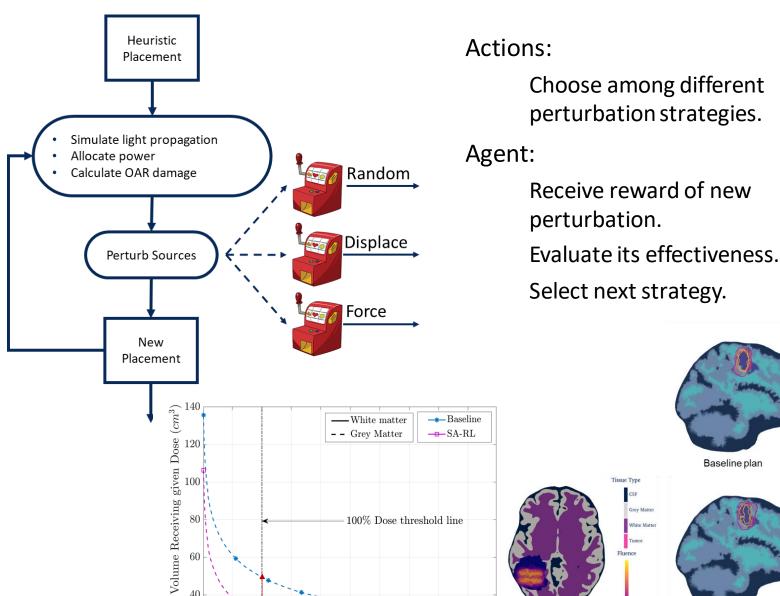
 $0.8 \ cm - 1 \ cm$ between every two sources.

1 *cm* from tumor boundary.

Simulated annealing based.

Accurate but Slow!





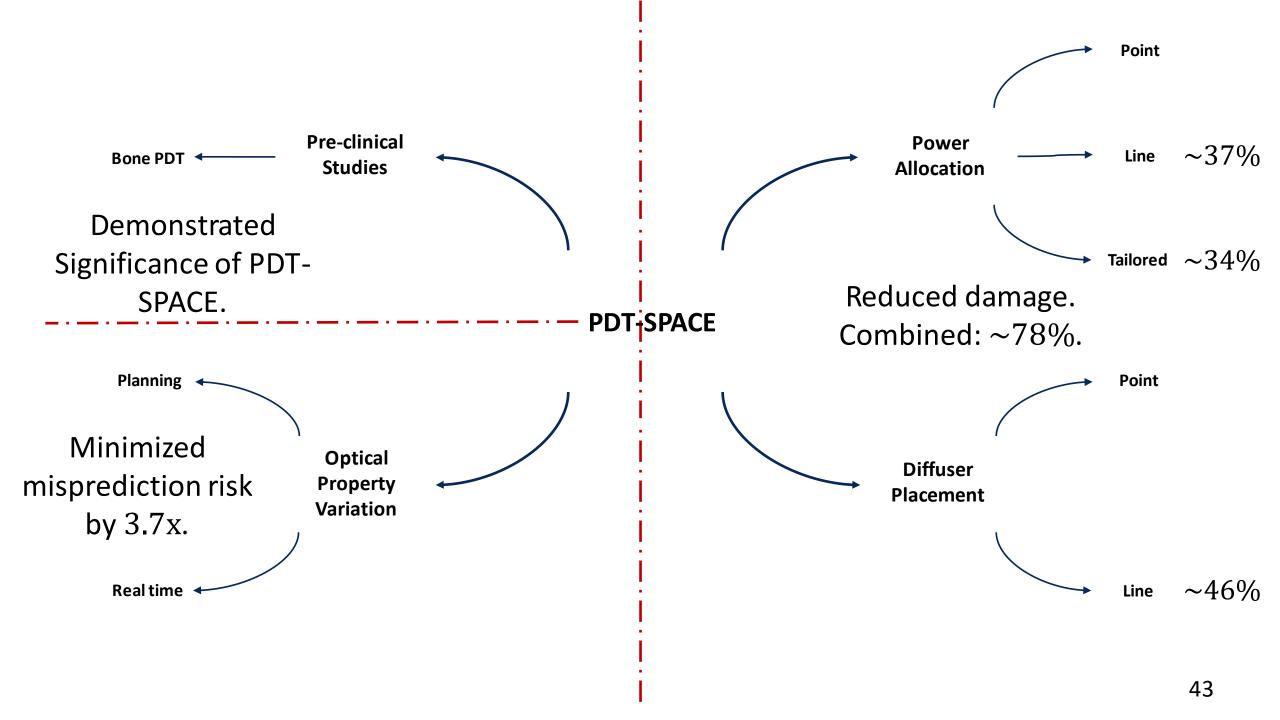
-68%

Perncentage of Dose Threshold (%)

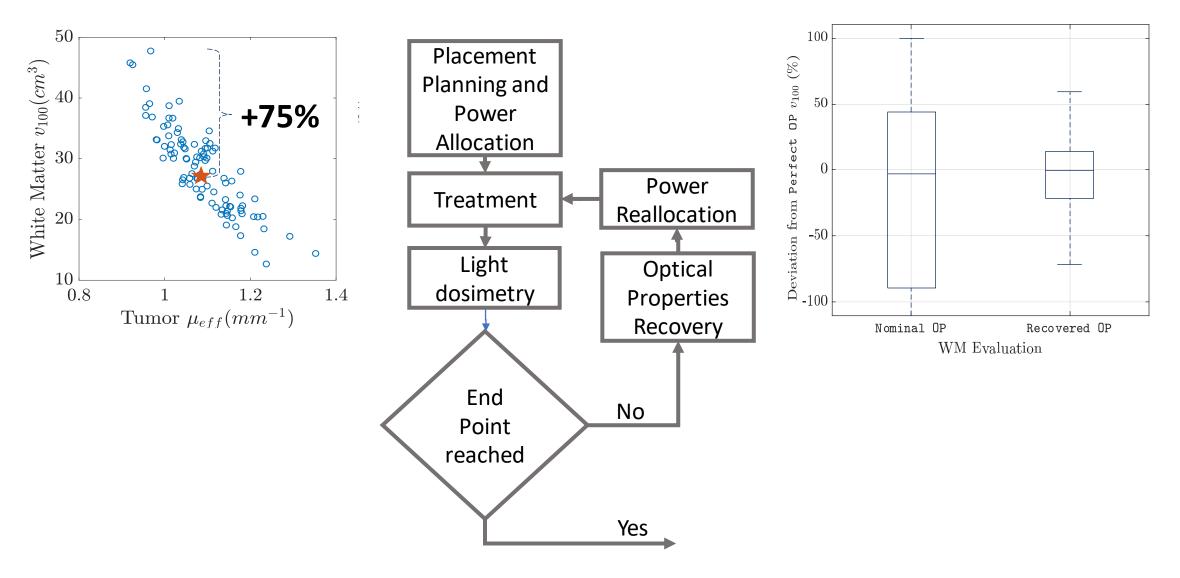
OAR

(b) SA-RL optimized plan

Baseline plan

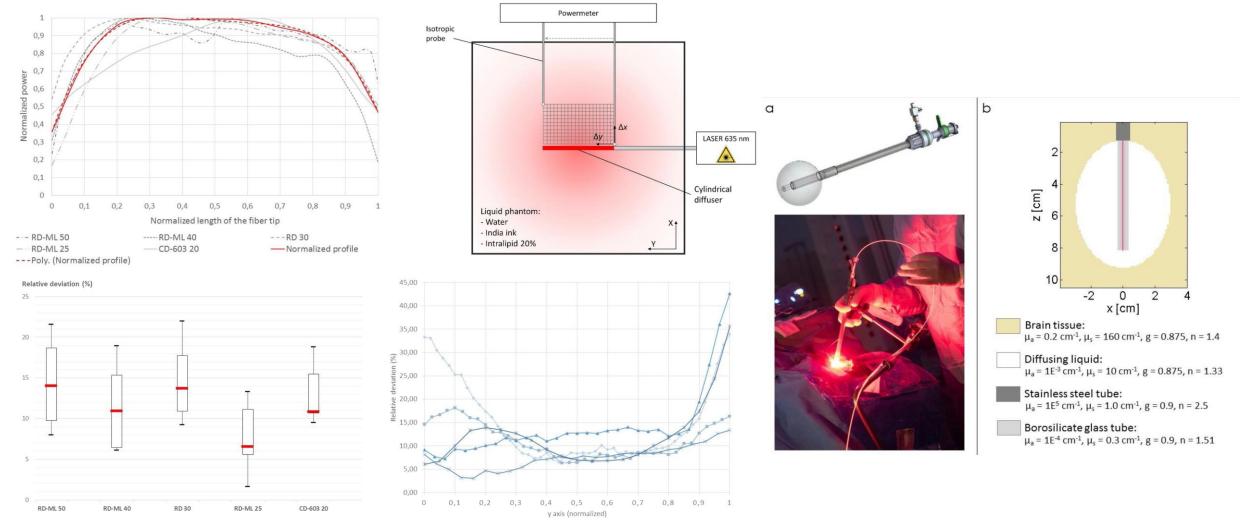


IMPACT OF OPTICAL PROPERITES VARIATION



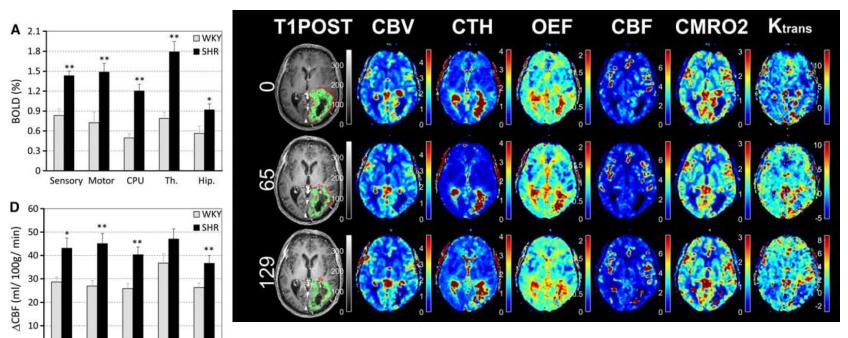
A.-A. Yassine, L. Lilge and V. Betz, "Machine learning for real-time optical property recovery in interstitial photodynamic therapy – A simulation based study," Biomed. Opt. Express (under review)

Considering ideal versus real dectors.



Dupont C et al. Parallelized Monte-Carlo dosimetry using graphics processing units to model cylindrical diffusers used in photodynamic therapy: From implementation to validation Photodiagnosis and Photodynamic Therapy 26 (2019) 351–360

fMRI to adjust PDT planning input parameters



Hansen et al., J Mag Res Img. 2016.

CBV = cerebral blood volume

CTH = cerebral transit heterogeneity

OEF = oxygen extraction fraction

CBF = cerebral blood flow

 $CMRO_2$ = metabolic rate of oxygen

k = diffusion coefficient

$$\alpha = \frac{-\log(2)}{\log(r) - r(\log(e)) + \log(e)}$$

$$r = \frac{t_{1/2peak} - AT}{t_{peak} - AT}$$

$$\beta = \frac{t_{peak} - AT}{\alpha}$$

$$Q = \text{capillary extraction fraction}$$

Diffusion Coefficient from OEF:
$$\int_0^\infty \frac{1}{({\pmb k}\beta)^\alpha \Gamma(\alpha)} \tau^{\alpha-1} e^{\frac{-\tau}{{\pmb k}\beta}} Q(\tau) d\tau$$

Hansen, et alor (2016). Reliable Estimation of Microvascular Flow Patterns in Patients with Disrupted Blood-Brain Barrier Using Dynamic Susceptibility Contrast MRI. Journal of Magnetic Resonance.

Sensory Motor CPU

Ε

ACBV_a (ml/ 100g)

Th.

□ WKY

Summary

- Precision iPDT using Anatomical detail is realized.
- Monitoring for personalize tissue optical properties required
 - Possibly under sampled by sensors
 - Need to employ clinical functional Imaging for heterogenous
- Predicting local pO₂ is possible using fMRI resolution
 - k for oxygen has been determined
- Predicting local PS distribution might be possible based on Blood Flow and Blood Volume
 - Needs to determine the bi-directional extravasation coefficient, k, of each photosensitizer
- Most importantly one needs to know the tissue response

Standardized reporting of PDT response

ALA induced PpIX mediated PDT in vitro

Cell Line	λ	LD ₅₀	T_{LD50}
	[nm]	[μM] ALA	[hv cm ⁻³]
HeLa	540	2.1	7.8 10 ¹⁶
RPE-1	540	3.3	1.210^{17}
HeLa	600-700	3.1	1.210^{17}
SiHa	600-700	6.25	2.410^{17}
MDA-MB-231	600-700	3.5	1.3 10 ¹⁷
2780AD	600-700	4.8	1.810^{17}
A2780-9S	600-700	3.5	1.310^{17}

L) or 16

TLD1433 mediated PDT in vitro

Cell Line	λ	LD ₅₀	T_{LD50}
	[nm]	[µM]	[hv cm ⁻³]
Skmel28	625	2.29	8.710^{16}
HL-60	625	7.7	2.9 10 ¹⁷
CT26	525	0.021	9.810^{15}
CT26.CL25	525	0.011	5.210^{15}
U87	525	0.051	2.410^{16}
F98	525 ± 25	2.81	1.310^{17}
T24	525 ± 25	0.0077	1.5 10 ¹⁵
AY27	525 ± 25	0.0039	7.7 10 ¹⁴
A549	532	0.099	1.610^{16}
RG-2	530	0.0265	2.4 10 ¹⁶

FullMonte

Fast Tetrahedral based photon propagation simulator

Open-source, rich set of input, output, and analysis tools

Robust framework for statistical verification

Data formats shared with class-leading visualization tools (.vtk)

Reads and write file formats from other similar simulators, (TIM-OS, MMC, COMSOL (finite-element

simulations), and MCML)

Platform Linux, Windows and Mac IO

Your Infrastructure

Docker images for SW

Https://gitlab.com/FullMonte/FullMonteSW

CUDAAccel

https://gitlab.com/FullMonte/cudaaccel

MeshTool

https://gitlab.com/FullMonte/MeshTool

PDT-SPACE

https://gitlab.com/FullMonte/pdt-space

Cloud (AWS)

FullMonteWEB

Http://fullmontesuite.herokuapp.com/application/

Access to GPU version

Access to PDT-SPACE

T Young-Schultz, S Brown, et al. FullMonteCUDA: GPU-accelerated Monte Carlo Simulator for Light Propagation BOE 10 (9) 4711-4726 (2019)

A.-A. Yassine, W. Kingsford, et al. Automatic interstitial photodynamic therapy via convex optimization Biomed. Opt. Express, 2018.

J Cassidy, S Lia et al. FullMonte high-performance, customizable Monte Carlo $_{f 49}$ biophotonic simulator" JBO (2018)