

Deep learning-based 3D in vivo dose reconstruction with EPID for magnetic resonance-linear accelerators

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Background

- MR-Linac
 - High soft-tissue contrast
 - No imaging dose
 - Imaging at different temporal scales
 - Functional imaging
- Adaptive radiotherapy
 - Online ART
 - Real-time ART
 - Biology guided ART



Online ART workflow



Patient specific quality assurance (QA) for MR-Linac

- Measurement based QA for reference plan (Used in clinic)
 - Reference plan is not delivered
 - Patient modeling and dose calculation is not included
- Independent dose check for adaptive plan (Used in clinic)
 - No delivery information is included
- EPID-based in vivo dosimetry (Desired in clinic)
 - End-to-end dose verification, including verification of patient geometry and setup, synthetic CT generation or density assignment, dose calculation from TPS, plan transfer and delivery
 - Real-time dose monitoring







Unity EPID



SAD = 143.5 cmSDD = 263.5 cmScale factor = 1.84EPID size: 41 cm \times 41 cm Dimension: 1024×1024 Resolution: $0.4 \text{ mm} \times 0.4 \text{ mm}$ Unattenuated region: [-11 cm, 11 cm] X: Y: [-4.8 cm, 4.8 cm]



Torres-Xirau I, Olaciregui-Ruiz I, van der Heide UA, Mans A. Two-dimensional EPID dosimetry for an MR-LINAC: Proof of concept. Med Phys. 2019;46:4193-4203.

Conventional in vivo dose reconstruction workflow



- Back projection algorithm is not accurate for inhomogeneous region, magnetic field effect can not be considered
- Forward calculation with Monte Carlo has low efficiency, is challenged for real-time dose monitoring

Accuracy of BP algorithm in low density region





Fig. 4. (a) TPS dose, (b) *in vivo* EPID dose and (c) γ distributions corresponding to the field delivered at gantry angle 0° of the liver plan with the worst agreement. The agreement worsens for the parts of the beam traversing low density lung tissue.

Olaciregui-Ruiz I, et al. Automatic dosimetric verification of online adapted plans on the Unity MR-Linac using 3D EPID dosimetry. Radiotherapy and Oncology, 2021, 157: 241-246.

Deep learning-based 3D in vivo dose reconstruction



- No complex patient scatter modeling and commissioning step is needed
- All scatter effects, beam hardening effects, heterogeneity effects, and magnetic field-induced EREs were assumed to be captured by the CNN model

Deep learning-based 3D in vivo dose reconstruction

Use Monte Carlo simulation to acquire 3D patient

Train & Validation dose and 2D portal dose simultaneously Density volume Density volume Portal dose Dataset generated with MC simulation Back-projection Back-projection Input Input Coarse dose Input Input Model 3DResUet Output Label Predicted dose MC dose Back projection to get coarse dose $d^{ini}(u, v, r) = (e^{-\mu \hat{r}} - e^{-\beta \hat{r}}) / (e^{-\mu \hat{r}_{\text{EPID}}} - e^{-\beta \hat{r}_{\text{EPID}}}) \frac{r_{\text{EPID}}^2}{r^2} \varphi(u, v)$

i Test

Portal dose

Coarse dose

MC dose

Comparison

Dataset generated

with

MC

simulation

Monte Carlo simulation with magnetic field

- gDPM —— developed by Xun Jia (JHU)
 - Based on fast MC code-DPM
 - GPU acceleration
 - Separate photon and electron transportation
 - ~60-80 times efficiency improvement with CPU version
- Extend charged particle transportation in magnetic field
 - First order approximation

$$\Delta \boldsymbol{u} = \frac{\boldsymbol{q} \cdot \boldsymbol{s}}{m_0 \gamma v_0^2} [\boldsymbol{v}_0 \times \boldsymbol{B}_0]$$

$$\boldsymbol{u}(\boldsymbol{s}) = \boldsymbol{u}_0 + \Delta \boldsymbol{u}$$

$$\boldsymbol{R} = \frac{m_0 \gamma v_0^2}{\boldsymbol{q} | \boldsymbol{v}_0 \times \boldsymbol{B}_0 |}$$

$$\delta = \frac{\boldsymbol{s}}{R} \ll 1$$

$$\Delta \boldsymbol{u} = \frac{\boldsymbol{s}}{R} \frac{\boldsymbol{u}_0 \times \widehat{\boldsymbol{B}}_0}{| \boldsymbol{u}_0 \times \widehat{\boldsymbol{B}}_0 |}$$

$$\boldsymbol{s} = \min(\boldsymbol{R} \cdot \delta, s_{vox}, s_{hard}, s_{ele})$$



Xun Jia, et al. Phys. Med. Biol. 2011.



FIG. 1. Simplified EGSnrc PRESTA-II step in the presence of a magnetic field. The particle is initially at \vec{x}_o with velocity $\vec{v}(0)$, it is then transported a step length, s, to \vec{x}_{CH} by the CH algorithm which samples direction of motion at an intermediate, $\vec{v}(s/2)$, and final, $\vec{v}(s)$. $\Delta \vec{x}_B$ and $\Delta \vec{u}_B$ are calculated using Eqs. (2) and (6) to obtain the final position, \vec{x}_f , and velocity, \vec{v}_f .

EGSnrc Manual.

Li et al. Med. Phys. 2021(48).

Monte Carlo beam modeling for Unity with 1.5T





Li et al. Med. Phys. 2021(48).

Dataset and augmentation



- Dataset
 - 21 brain cases, 46 NPC cases, 15 lung cases, 14 rectum cases
 - training and validation set (78 cases), test set (18 cases)
 - 576 original treatment beams
- Augmentation
 - rotate the original beam angles by $10^{\circ}-15^{\circ}$ for 2–3 times
 - 1841 and 121 beams for training and validation
 - all volume dose and portal dose were recalculated for each beam

Network structure





Dose reconstruction result





Table 1. Averaged γ -pass rates and MAE parameters for 18 tested patients (mean \pm SD).

Site		Brian (4)	nasopharynx (8)	Lung(3)	Rectum (3)
γ-pass (%)	Dose > 0%, 3%/2 mm	97.42 ± 2.66	98.53 ± 0.95	99.41 ± 0.46	98.63 ± 1.01
	Dose>20%, 3%/2 mm	95.48 ± 3.31	97.20 ± 1.42	95.35 ± 0.57	95.29 ± 2.89
	Dose>50%, 3%/2 mm	94.32 ± 3.77	95.10 ± 2.01	90.40 ± 2.71	95.83 ± 1.56
	Dose > 0%, 2%/2 mm	94.02 ± 6.21	96.46 ± 1.93	98.72 ± 0.88	96.72 ± 1.73
MAE (%)	0.82 ± 0.36	0.88 ± 0.21	0.41 ± 0.19	0.67 ± 0.09	

Dose reconstruction result





Dose reconstruction result





Conclusion



- Proposed a CNN-based 3D in vivo dose reconstruction method
 - Physical processes were learned through accurate MC data-driven model training
 - Simplicity in dose reconstruction and model commission
 - Training data can be generated with clinic TPS by adding a virtual EPID structure

- Limitations
 - No validation was conducted for real measured EPID images
 - Field truncation by real EPID was not considered



Thanks !