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# Simplex noise as training data for learned 3D dose calculation

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#### **Dose Calculation**





## **Machine Learning for Dose Calculation**



#### THE PROBLEM

Machine learning requires data

#### **TRAINING DATA**



## A POTENTIAL SOLUTION

Simplex noise as training data

## **Simplex Noise**





## **Simplex Noise**



#### ALGORITHM

- Generate one noise volume per material
- Rescale each noise volumes to material density
- Generate another noise volume per material
- Sample n<sup>th</sup> scaled noise volume where n<sup>th</sup> unscaled noise volume is minimal

#### EVALUATION

Simplex noise

## **Encoding Fields**

#### **DENSITY CHANNEL**



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#### **Encoding Fields**

$$\Psi_0(\boldsymbol{r}, E) = \frac{e^{-\tau(\boldsymbol{r}_0 \to \boldsymbol{r}, E)} Q_A(\boldsymbol{r}_0, E)}{|\boldsymbol{r} - \boldsymbol{r}_0|^2}$$

#### **DENSITY CHANNEL**

#### **FLUENCE CHANNEL**



## **Encoding Fields**



**HD U-NET** 

Source: Nguyen et. al. 2019





## RESULTS

Simplex noise

#### **MEGAVOLTAGE X-RAYS**



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#### **MEGAVOLTAGE X-RAYS**



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#### CONCLUSIONS

Simplex noise is a viable source of high quantity, high quality training data for dose calculation using machine learning.

A model trained using the proposed framework may be fast enough to use in inverse optimisation algorithms/as a secondary MU calculation.

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