

Primal-Dual, a novel approach to optimisation in treatment planning

An alternate automated planning process for heavy ion beam therapy

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NPL



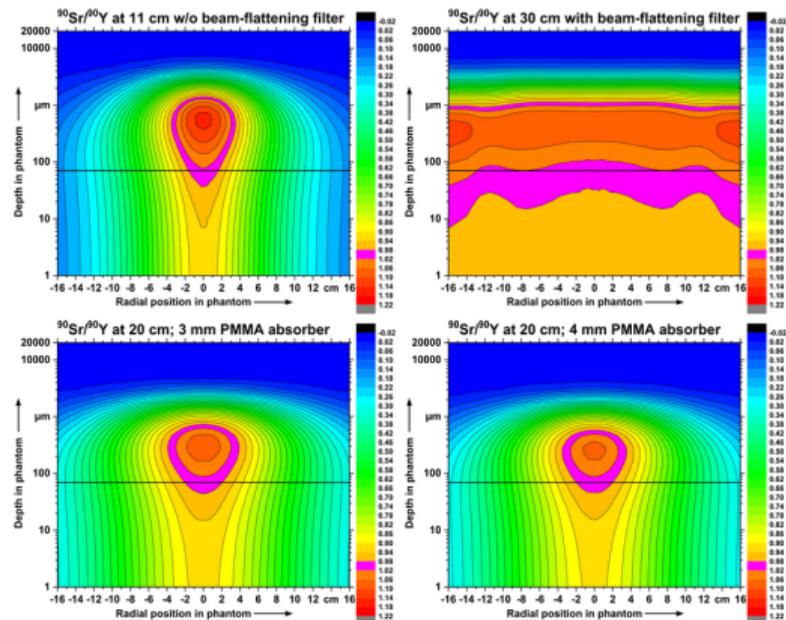
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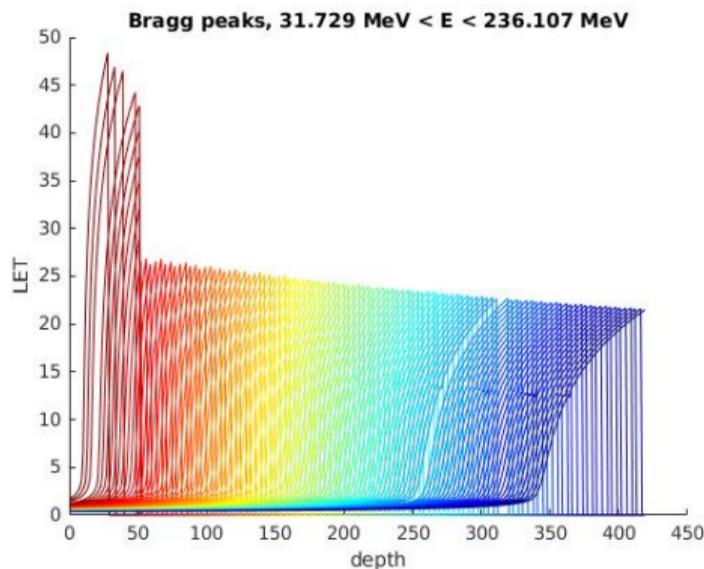
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 - Advantages of the Primal-Dual system: Rate of Convergence
 - Advantages of the Primal-Dual system: Uncertainty Quantification

Pseudo-Algorithm for the Forward Model

- 1 Fix the beam angles; initialise:
 - Beam Shape
 - Beam Energy
- 2 Compute all dose profiles.
- 3 Calculate optimal profile.





Computational Complexity

No. of forward solves =
 No. possible beam shapes \times
 No. possible beam energies \times
 No. variations for Robustness

Current Methodology

For $i = 1, \dots, N$ let \mathcal{D}_i denote the precomputed dose profiles

$$\min_{a_i \geq 0} \left\| \sum_{i=1}^N a_i \mathcal{D}_i - \mathcal{D}_T \right\|_{\text{PTV}}^2 + \sum_{n=1}^N w_n \|a_i \mathcal{D}_i\|_{\Omega_n}^2 \quad (1)$$

Alternate Methodology

Let u denote the particle density

$$\min_{u, f} \| \mathcal{D}u - \mathcal{D}_T \|_{\text{PTV}} + \sum_{n=1}^N w_n \| \mathcal{D}u \|_{\Omega_n}^2 + \alpha \| f \|^2 \quad (2)$$

subject to u satisfying Boltzmann transport equation with admissible inflow.

Pseudo-Algorithm for the Primal-Dual system

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- 5 Use Dual Dose profile to correct beam configuration in forward solver.

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- 4 Compute the Dual Dose profile with a single dual solve.
- 5 Use Dual Dose profile to correct beam configuration in forward solver.
- 6 Repeat steps 2-5.

Advantages

- 1 Mathematical certainty of convergence
- 2 Significantly fewer forward solves
- 3 Accounting for uncertainties in the model

Properties of Primal-Dual

- 1 Existence of solutions to the stationary problem.
- 2 Uniqueness of solutions to the stationary problem.
- 3 Stability of solutions of the time-dependent problem.

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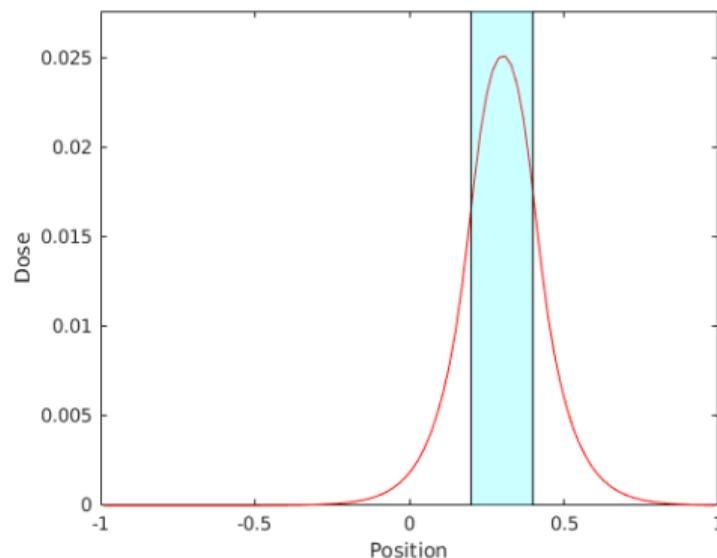


Figure: Optimised solution to the Kolmogorov equation.

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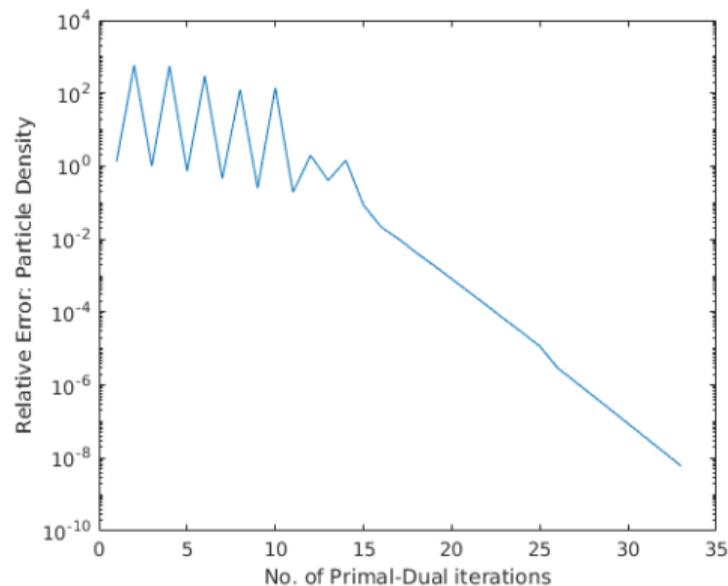


Figure: Relative error between each Primal-Dual iteration.

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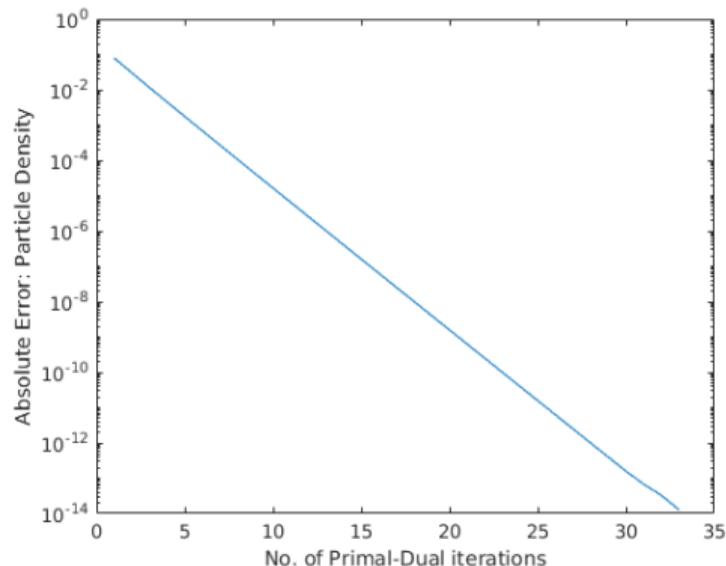


Figure: Absolute error between each Primal-Dual iteration.

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Perturbed system

Let $\epsilon(\mathbf{x})$ be a random function then consider

$$\max_{\mathbb{P}(\epsilon) < 0.95} \min_{u, f} J(u, f)$$

subject to

$$\frac{\partial u}{\partial t} + \Omega \cdot \nabla_{\mathbf{x}} u = \sigma_a u - \iint_{I \times \mathbb{S}^2} \sigma_s u d\Omega' dE$$

$$u(t, \mathbf{x}, E, \Omega) = f(t, \mathbf{x} + \epsilon(\mathbf{x}), E, \Omega)$$

Thank you for listening

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