

Impact / Innovation

- High-dose-rate brachytherapy (**HDR-BT**) using hybrid interstitial (**IC/IS**) applicators is crucial for cervical cancer patients with large/irregular tumours
- Needle arrangement impacts organ-at-risk and tumour dose, correlated with toxicity and local recurrence^{1,2}
- Needle selection is complex, relying on physician expertise,³ highlighting the need for **predictive models**

Goal – Quantify the **dosimetric impact of machine learning predicted needle arrangements** to assess clinical benefits of model in a prospective study

Materials & Methods

Patient Cohort

- 10 cervical cancer patients receiving HDR-BT (December 2020 – October 2022)
- Treated using intracavitary (**IC**) or hybrid interstitial (**IC/IS**) ring or semi-lunar ovoid applicators – Vienna or Venezia (**Figure 1**)

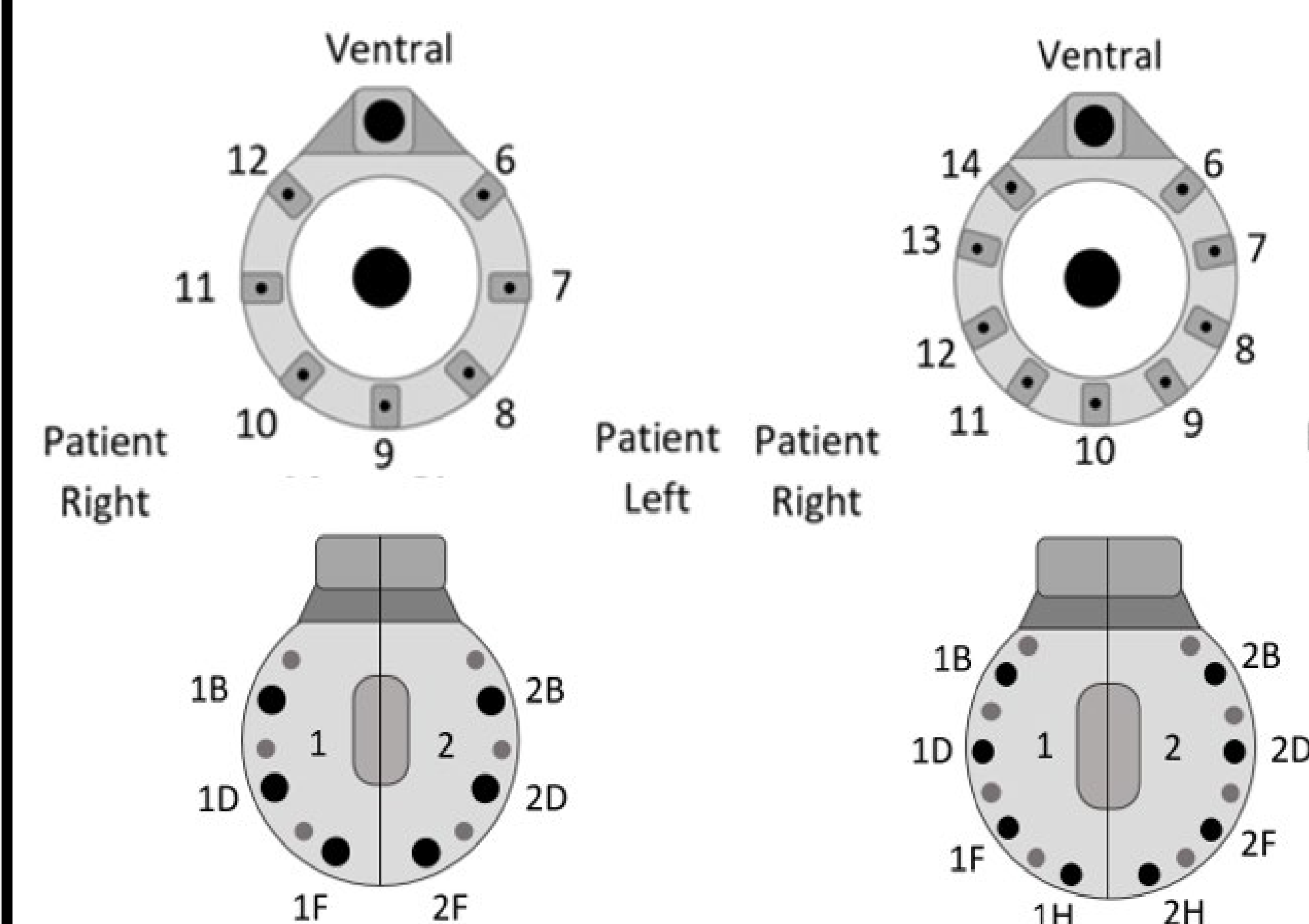


Figure 1. Needle positions for 26 mm (top left) and 30/34 mm (top right) Vienna applicator and the 22 mm (bottom left) and 26/30 mm (bottom right) Venezia applicator

Machine Learning (ML) Prediction

- Model predicts IC or IC/IS implant based on target volume geometry (**HR-CTV**)
- Arrangement of needles predicted for IC/IS cases
- Performance metrics compare ML predictions to clinical treatment plan needles

Dosimetric Assessment

- Dosimetric differences between ML-predicted and clinical needle arrangements assessed for three treatment fractions
- Target coverage matched to clinical plan, differences in organs-at-risk doses assessed

Results

Patient Cohort

- 5 of the 10 patients had ML-predicted IC/IS applicator, in agreement with treatment applicator
- These were eligible for replanning to compare needle arrangement dosimetry

Machine Learning Performance (Table 1)

- Average agreement between ML-predicted and clinical needles ~80% for all metrics
- 40% of patients had **full agreement** between ML-predicted and clinical needles

Performance Metric	Patient 1	Patient 2	Patient 3	Patient 4	Patient 5	Average
Accuracy	66.7%	100%	57.1%	87.5%	100%	82.3%
Balanced Accuracy	66.7%	100%	58.3%	87.5%	100%	82.5%
Precision	80.0%	100%	66.7%	80.0%	100%	85.3%
Recall (Sensitivity)	66.7%	100%	50.0%	100%	100%	83.3%
Specificity	66.7%	100%	66.7%	75.0%	100%	81.7%

Table 1. Performance metrics for machine learning model needle arrangement prediction compared to clinical needle arrangement

Patient #	Bladder D2cc (cGy)		Rectum D2cc (cGy)		Sigmoid D2cc (cGy)		Bowel D2cc (cGy)	
	Fx 1	Fx 2 & 3	Fx 1	Fx 2 & 3	Fx 1	Fx 2 & 3	Fx 1	Fx 2 & 3
1	-48.0	-17.4	-51.4	-8.0	-94.1	-9.8	-88.1	-48.0
2	4.7	7.0	-4.0	3.8	-43.3	4.1	-6.4	4.7
3	-95.4	-12.5	-62.3	8.4	-55.3	9.5	-9.7	-95.4
4	-10.6	-4.0	19.5	-2.6	-0.6	2.9	-0.1	-10.6
5	-53.3	-28.5	-31.3	4.1	-6.1	1.9	-12.6	-53.3
Average	-40.5	-11.1	-25.9	1.1	-39.9	1.7	-23.4	-40.5

Table 2. Differences between ML-predicted needle replan and the clinical plan dosimetry. A negative value (shown in green) indicates a **reduction** and a positive value (shown in red) indicates an **increase** in organ-at-risk doses when using the ML-predicted needles.

Dosimetric Comparison (Table 2)

- Dosimetric differences between ML and clinical plans were most significant for **first fraction**
- Largest changes occurred when:
 - Needles not used clinically until the second and third fractions
 - Fewer needles were used for the first fraction
- Using the ML-predicted needles resulted in average organ-at-risk doses **improving**
- Factors impacting dosimetry, from most to least important appear to be:
 - Use of hybrid interstitial needles
 - Number of needles used
 - Arrangement of needles

Conclusions

- ML-predicted needle arrangements have **greatest dosimetric improvement for first fraction**
- For the following fractions, differences between the ML-predicted and clinical needle arrangements had minimal impact on plan quality, leading to **comparable plans**
- Further work assessing model performance across a large cohort of patients in a multi-centre setting is required

The presented machine learning-based decision support tool shows strong predictive capabilities in a prospective setting, supporting the utility of such a model in a clinical setting.

References

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