Evaluation of deformable image registration algorithm for determination of accumulated dose for brachytherapy of cervical cancer patients

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Abstract

Purpose: This study was designed to assess the dose accumulation (DA) of bladder and rectum between brachytherapy fractions using hybrid-based deformable image registration (DIR) and compare it with the simple summation (SS) approach of GEC-ESTRO in cervical cancer patients.

Material and methods: Patients (n = 137) with cervical cancer treated with 3D conformal radiotherapy and three fractions of high-dose-rate brachytherapy were selected. CT images were acquired to delineate organs at risk and targets according to GEC-ESTRO recommendations. In order to determine the DA for the bladder and rectum, hybrid-based DIR was done for three different fractions of brachytherapy and the results were compared with the standard GEC-ESTRO method. Also, we performed a phantom study to calculate the uncertainty of the hybrid-based DIR algorithm for contour matching and dose mapping.

Results: The mean ± standard deviation (SD) of the Dice similarity coefficient (DICE), Jaccard, Hausdorff distance (HD) and mean distance to agreement (MDA) in the DIR process were 0.94 ±0.02, 0.89 ±0.03, 8.44 ±3.56 and 0.72 ±0.22 for bladder and 0.89 ±0.05, 0.80 ±0.07, 15.46 ±10.14 and 1.19 ±0.59 for rectum, respectively. The median (Q1, Q3; maximum) GyEQD2 differences of total D2cc between DIR-based and SS methods for the bladder and rectum were reduced by –1.53 (–0.86, –2.98; –9.17) and –1.38 (–0.80, –2.14; –7.11), respectively. The mean ± SD of DICE, Jaccard, HD, and MDA for contour matching were 0.98 ±0.008, 0.97 ±0.01, 2.00 ±0.70 and 0.20 ±0.04, respectively for large deformation. Maximum uncertainty of dose mapping was about 3.58%.

Conclusions: The hybrid-based DIR algorithm demonstrated low registration uncertainty for both contour matching and dose mapping. The DA difference between DIR-based and SS approaches was statistically significant for both bladder and rectum and hybrid-based DIR showed potential to assess DA between brachytherapy fractions.

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Key words: deformable image registration, brachytherapy, cervix, HDR.

Purpose

Combining external beam radiotherapy (EBRT) and chemotherapy with high-dose-rate brachytherapy (HDR-BT) as boost is the standard approach to treat locally advanced cervical cancer. For HDR-BT, several types of intracavitary applicators are commonly used with or without interstitial needles. After completing the applicator insertion in each fraction, image-guided treatment planning is used to calculate three-dimensional (3D) dose distribution. Some recommendations have been published by the Groupe Européen de Curiotérapie-European Society for Therapeutic Radiology and Oncology (GEC-ESTRO) working group, and the American Brachytherapy Society (ABS) based on 3D image-guided treatment planning for contouring, applicator reconstruction, prescribing, and reporting [1,2,3,4,5,6]. The recommended prescription dose for high-risk clinical target volume (HR-CTV) with $\alpha/\beta = 10$ is 80-90 GyEQD2 (45-50 Gy in 1.8-2.0 Gy/fraction as EBRT), expressed as the equivalent dose in 2 Gy per fraction (EQD2). The maximum dose to 2 cm$^3$ (D2cc) of the organs at risk (OARs) such as bladder, rectum and sigmoid with $\alpha/\beta = 3$ should not exceed 90 GyEQD2, 75 GyEQD2 and 75 GyEQD2, respectively. The GEC-ESTRO recommended simple summation (SS) of EBRT dose prescription and D2cc to predict toxicity

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in OARs. This is based on the assumption (the worst-case assumption) that each volume of interest has the same location in EBRT and during each HDR-BT fraction. Some studies have indicated that organ motions, differences in filling of bladder/rectum, presence of gas in the rectum, and differences in location of applicator cause the SS approach to overestimate the total dose in OARs, thereby limiting elevation of the HR-CTV dose or target coverage [7,8,9,10].

Recently, some investigations have proposed a new strategy based on image registration to improve calculation of maximum dose to OARs. In several studies, rigid image registration (RIR) has been applied to assess dose accumulation (DA); however, the complexity of organs’ deformation due to the anatomic changes and applicator insertion limits the accuracy of calculation [11,12,13]. To overcome these limitations, several deformable image registration (DIR) algorithms have been investigated [14,15,16,17,18]. Validation of the accuracy of DIR algorithms is a challenging task due to complexity, clarity issues, and uncertainties in the image. There are some methods to evaluate the accuracy of DIR including propagation of the same points or contours between registered images, registration of images of physical or virtual phantoms, and patient image registration [19].

On the other hand, high accuracy of the DIR algorithm does not necessarily address the accuracy of DA. Hayashi et al. [20] estimated the DIR accuracy by creating a virtual phantom and using the Dice similarity coefficient (DICE). Although they reported DICE equal to 0.96 in the virtual phantom, the average dose difference was 8%, which was associated with uncertainty of the DIR algorithm. Reniers et al. [21] also reported 5-10% differences between DA using DIR and GEC-ESTRO recommendations depending on the DIR scheme. However, their study did not examine a sufficient number of patients.

To the best of our knowledge, there is no study investigating the effect of the DIR algorithm in terms of dosimetry by virtual phantom and a subsequent statistical population of patient data to obtain reasonable results for comparing DIR and SS approaches for DA. The aim of this study is to assess DA of bladder and rectum using a hybrid-based DIR algorithm and compare it statistically with the GEC-ESTRO method in multiple fractions of cervical cancer brachytherapy (BT).

Material and methods

This study was conducted based on two main parts including virtual phantom and patient data.

Virtual phantom data

To investigate the uncertainty of the hybrid-based DIR algorithm, we created a virtual phantom as ground truth data by commercial software ImSimQA v4.0 (Oncology Systems Limited, UK) as displayed in Figure 1A-C. “ImSimQA tool can serve as a virtual deformable quality assurance tool by simulating clinically observed organ deformations” [22].

In this study, we simulated a simple human pelvis. In order to produce warped image sets based on applying geometric deformation, there are two approaches: global and local deformation. Both global and local deformation are based on radial basis function (RBF) and different kernel functions.

Global deformations

The ImSimQA uses RBF with thin-plate spline (TPS) as an interpolation to create global deformation. In this regard, two sets of points must be chosen manually on the images including source points (SPs) and target points (TPs). This tool creates a grid in each slice of image and then assigns x, y to all points (SPs and TPs) which are manually selected by clicking with the mouse on the image. Vectors with origin from SPs to TPs determine deformation direction of the grid. Then mapping function \( f(x, y) \) which defined RBF with TPS kernels in the software will map SPs to TPs to create global warped image sets. More details about the \( f(x, y) \) function can be found in the Bookstein et al. study [23].

Local deformations

ImSimQA uses RBF with compact support radial basis function (CSRBF) as an interpolation to create local deformation. This will be applied on the image same as the global deformation process with the only difference in kernels. More details about CSRBF can be found in the study by Wendland et al. [24].

We created two groups of global and local warped image sets. We considered average 5 mm and 15 mm distances between SPs and TPs to create groups of warped image sets named small and large warped image sets, respectively (Figure 1E-H).

Threshold-based segmentation was performed on image sets prior to registration by the commercial software Artiview v3.20.1 (Aquilab, France) as illustrated in Figure 1D-H. All images were imported to the Sagiplan (Eckert & Ziegler BEBIG Co., Germany) BT treatment planning system. A tandem and ovoid applicator (Fletcher type) with the same position, size, length, dwell position, and dwell time was added to create the same isodose curves across all virtual phantoms.

We divided the designed phantoms to primary and secondary categories for the registration process. The images of the phantom without deformation were considered as primary (fixed or reference images), whilst images of the warped phantom were regarded as secondary images (moving or target images). RIR followed by DIR was performed between the primary and each secondary image to calculate the contour matching and dose mapping. When performing DIR between two image sets, a DIR algorithm without uncertainty should generate a new image set with the same structures (contour matching) and dose distribution (dose mapping) equal to the primary image.

Patient data

Data for \( n = 137 \) patients with locally advanced cervical cancer treated with 3D conformal radiotherapy (3DCRT) and HDR-BT from 2016 to 2018 were included in this study. All patients were treated with 3DCRT with
50 Gy in 25 fractions and 3 fractions of HDR-BT (8.6 Gy/Fr) [25] using tandem and ovoids (Fletcher type without interstitial needles). The 3DCRT was planned to cover the clinical target volume (CTV) comprising the cervix, entire uterus, bilateral parametria, upper half of vagina, and lymph nodes (common, external and internal iliac) by at least 95% of the prescription dose. Before each fraction of BT, a Foley catheter filled with contrast agent (7cc) was inserted into the bladder. 120 cm$^3$ of normal saline solution was injected into the bladder and clamped immediately before the CT scanning. Thereafter, CT images (GE HiSpeed Dual-Slice) with 3 mm slice thickness were acquired to delineate OARs and the target to calculate the dose distribution according to GEC-ESTRO recommendations. Also, MRI images (T2 weighted) were acquired to determine HR-CTV precisely just before the first BT fraction. HDR-BT was planned using SagiPlan as the treatment planning system and a $^{60}$Co source (Eckert & Ziegler BEBIG Co., Germany) to deliver 90% of the dose to the HR-CTV. We considered dose constraints for OARs including bladder, rectum, and sigmoid according to GEC-ESTRO recommendations.

**Deformable image registration and dose accumulation**

In this study, we applied a hybrid-based DIR algorithm with sum of the squared differences (SSD) as the similarity measure by a commercial system (MIM v6.7; MIM software, Inc.). The hybrid-based DIR algorithm of this software considered both intensity and contours (should be selected manually) of images. The cost function term contains two parts: $C_{\text{similarity}}$ and $C_{\text{smooth}}$.

$$\text{cost function} = C_{\text{similarity}} + \lambda C_{\text{smooth}}$$

$C_{\text{similarity}}$ minimizes the SSD between primary and secondary images. $C_{\text{smooth}}$ is a regularization term that maintains smoothness of the control point grid and the resulting DVF. $\lambda$ is the balance of these two terms, which by default has been set to a value of 0.5.
Initially, we performed rigid registration to align images (using the whole image), after which DIR was performed. This approach was done for both virtual phantom and patient images. Contours of bladder, rectum, and sigmoid from patient data and all structures in the virtual phantom were selected for the DIR process. Also, RIR and DIR process durations were calculated.

In order to remove the rectal gas and bladder Foley catheter filling effects in the DIR process, we masked contours to the values 1000, –1000, and 0 for bladder, rectum, and sigmoid, respectively (Figure 2).

Patient’s CT image data of the first fraction (Fr1) were considered as primary, whilst second (Fr2) and third fraction (Fr3) images were considered as secondary images in the registration process. Displacement vector fields obtained from the DIR process (from warped Fr2 to Fr1 and warped Fr3 to Fr1) were applied to the dose distribution for DA. All doses were converted to EQD2 using a linear-quadratic model with α/β = 3 for normal tissues (OARs). DVH parameters including D0.1cc, D1cc, D2cc, and D5cc of bladder and rectum were calculated by accumulated DVH and SS approaches. Also, in the virtual phantom, D0.1cc, D1cc, D2cc, and D5cc between structures of two registered images were used to calculate the dose mapping.

The overall framework of this study is summarized in Figure 3.

**Deformable image registration accuracy**

Different metrics have been developed to evaluate of DIR accuracy [26,27,28]. We used DICE, Jaccard, Hausdorff distance (HD), and mean distance to agreement (MDA) metrics between structures to quantitatively calculate the uncertainty of contour matching in the virtual phantom. We also calculated these metrics for patient data. DICE and Jaccard are given as:

\[
\text{DICE} = 2 \frac{|A \cap B|}{|A| + |B|} \\
\text{Jaccard} = \frac{|A \cap B|}{|A| + |B| - |A \cap B|}
\]

DICE and Jaccard measure the overlap of two contours where A and B denote the contours of primary and secondary images, respectively. DICE and Jaccard range from 0 to 1. If A and B completely overlapped, DICE and Jaccard would be 1; otherwise it would be 0 for complete non-overlapping conditions.

HD describes the maximum closest distance for each apex of the two volumes (mm). Also, MDA is defined as the average distance of each point on one contour to the closest points on the other (mm). These two parameters can also be measured as below:

\[
\text{HD} (X, Y) = \sup \inf d (X, Y) \\
\text{MDA} (X, Y) = \text{avg} \inf d (X, Y)
\]

Where sup, inf, and avg stand for supremum, infimum, and average [29]. These metrics were calculated for both RIR and DIR processes.

**Statistical analysis**

DVH parameters from DIR-based and SS approaches were compared using the Wilcoxon signed rank sum test. IBM SPSS v23 (SPSS, Inc.) was used for all statistical analyses, while mean ± standard deviation (SD) or the median (Q1, Q3; maximum) was used for summarizing the data. Interquartile range (IQR) was also employed to find outliers in the data, which equaled differences between the third and first quartiles. Three criteria were considered including 1.5 IQR as a mild outlier, 2 IQR and 2.5 IQR as moderate outliers, and 3 IQR as an extreme outlier. A p-value less than 0.05 was considered statistically significant.

**Results**

**Virtual phantom**

Table 1 presents the uncertainty of contour matching and dose mapping in the DIR process. The results are shown for all metrics (DICE, Jaccard, HD, MDA). DIR has the highest accuracy for contour matching, where the differences between DIR and RIR were significant (p < 0.001 for all metrics).

For, D0.1cc, D1cc, D2cc, and D5cc, between two registered images (primary and each secondary image), the overall mean ± SD difference of the dose distribution among all structures in the DIR process was 0.012 ±0.008 Gy and
Fr – fraction, RIR – rigid image registration, DIR – deformable image registration

Fig. 3. Flowchart of entire procedure used in this study. Virtual phantom was created as ground truth data to calculate uncertainty of DIR algorithm. RIR followed by DIR was performed between fixed image and each warped image set. When performing DIR between two image sets, a DIR algorithm without uncertainty should generate a new image set with the same structures (contour matching) and dose distribution (dose mapping) equal to the primary image. Patient’s CT image data sets for first fraction (Fr1) were considered as primary whilst second (Fr2) and third fraction (Fr3) images were considered as secondary images in the registration process, respectively. RIR was applied to align images, then DIR was performed. Displacement vector fields obtained from DIR process were applied to dose distribution for dose accumulation. All doses were converted to EQD2 using linear-quadratic model with “α/β = 3” for normal tissues. DVH parameters such as D0.1cc, D1cc, D2cc, and D5cc of bladder and rectum from accumulated DVH and simple summation approaches were calculated and results were compared

0.043 ±0.013 Gy, leading to 1.05 ±0.26% and 3.58 ±0.83% for small and large deformations, respectively. These results indicated that the maximum uncertainty of D2cc for the DIR method was about 3.58%.

**Patients**

The mean duration of RIR and DIR processes was 20 and 50 seconds, respectively. The mean ± SD values of DICE, Jaccard, HD, and MDA in RIR and DIR processes are reported in Table 2. There were statistically significant differences between the metrics calculated by DIR compared to RIR (p < 0.001 for all metrics), where DIR had the highest accuracy to match contours for bladder and rectum. The median (Q1, Q3; maximum) GyEQD2 differences of total D2cc between DIR-based and SS approaches for the bladder and rectum were reduced to −1.53 (−0.86, −2.98; −9.17) and −1.38 (−0.80, −2.14; −7.11), respectively (Figure 4).

The differences of D0.1cc, D1cc, D2cc, and D5cc between the DIR and SS approaches for bladder and rectum are outlined in Table 3. The DIR-based DA was lower than SS’s DA for all patients (except for D0.1cc in some patients). The median (Q1, Q3; maximum) GyEQD2 differences of total D2cc between DIR-based and SS approaches for the bladder and rectum were reduced to −1.53 (−0.86, −2.98; −9.17) and −1.38 (−0.80, −2.14; −7.11), respectively (Figure 4).

The median volume differences for bladder and rectum were 90 and 53 cc, respectively (Table 4). The median volume differences for bladder...
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Discussion

This study is the first to use hybrid-based DIR with a virtual phantom and subsequent statistical population of patient data to evaluate uncertainty and DA, respectively for bladder and rectum in cervical cancer patients. We observed that hybrid-based DIR had 0.16 mm and 0.2 mm uncertainty for small and large deformations, respectively. This resulted in 1.05% and 3.58% uncertainty in dose mapping when the contours overlapped. In comparison with Hayashi et al. [20], our hybrid-based DIR algorithm was found to have a lower uncertainty (3.58% vs. 8% reported by Hayashi et al.). Concerning patient data, we found that the differences of DIR-based and SS approaches were statistically significant for D_{0.1cc}, D_{1cc}, D_{2cc}, and D_{5cc} for both bladder (p < 0.001) and rectum (p < 0.001), indicating that the hybrid-based DIR algorithm has a potential for DA between brachytherapy fractions.

There are two general categories to evaluate uncertainty of the DIR algorithm: physical and virtual phantoms. The virtual phantom is the preferred method, since the physical phantom suffers from some limitations including inability to know the transformation of each voxel as well as difficult and time-consuming creation of each clinical situation [19]. In addition, different noise patterns can be added to virtual phantom images to simulate noise variation between the image sets [19]. It is possible to apply a known geometric function to the virtual phantom for creating ground truth data. Nevertheless, it should be noted that applying a geometric deformation function and DIR algorithm should not be the same model as it causes bias in the results. In our study, we applied RBF with different kernels as a geometric deformation function (RBF is landmark-based transformation function) to the virtual phantom to create local and global warped image sets. We used hybrid-based DIR to resolve the deformation, which is a different model from RBF with different kernels.

It is important to choose suitable DIR algorithms. Some investigations have applied RIR and intensity-based DIR for DA [11, 13]. Due to a lack of degree of freedom for the rigid algorithm, it is impossible to perform local mapping. Sabater et al. [30] investigated DA by RIR for

| Table 1. Uncertainty of contour matching and dose mapping in DIR process |
|---------------------------------------------------------------|
| | Small deformation | Overall | Large deformation | Overall |
| | Global | Local | Mean ±SD | Global | Local | Mean ±SD |
| DICE_{DIR} | 0.71 ±0.25 | 0.97 ±0.02 | 0.84 ±0.22 | 0.71 ±0.08 | 0.76 ±0.21 | 0.74 ±0.16 |
| Jaccard_{DIR} | 0.60 ±0.24 | 0.95 ±0.04 | 0.77 ±0.24 | 0.56 ±0.10 | 0.66 ±0.21 | 0.61 ±0.17 |
| HD_{RIR} (mm) | 9.53 ±3.59 | 2.66 ±1.53 | 6.09 ±4.40 | 14.59 ±4.57 | 13.18 ±5.00 | 13.89 ±5.05 |
| MDA_{RIR} (mm) | 3.58 ±1.69 | 0.30 ±0.26 | 1.94 ±2.03 | 4.34 ±1.08 | 2.87 ±1.06 | 3.60 ±1.30 |
| DICE_{DIR} | 0.99 ±0.007 | 0.99 ±0.008 | 0.99 ±0.008 | 0.98 ±0.006 | 0.98 ±0.010 | 0.98 ±0.008 |
| Jaccard_{DIR} | 0.97 ±0.01 | 0.98 ±0.01 | 0.97 ±0.15 | 0.97 ±0.013 | 0.97 ±0.02 | 0.97 ±0.01 |
| HD_{DIR} (mm) | 1.61 ±0.71 | 1.67 ±0.88 | 1.64 ±0.80 | 1.95 ±1.00 | 2.00 ±0.79 | 2.00 ±0.70 |
| MDA_{DIR} (mm) | 0.18 ±0.04 | 0.15 ±0.04 | 0.16 ±0.04 | 0.20 ±0.04 | 0.20 ±0.03 | 0.20 ±0.04 |
| D_{DSC} (Gy) | 0.012 ±0.004 | 0.010 ±0.001 | 0.012 ±0.008 | 0.058 ±0.006 | 0.041 ±0.006 |
| D_{DSC} (Gy) | 0.020 ±0.013 | 0.009 ±0.006 | 0.012 ±0.008 | 0.052 ±0.012 | 0.037 ±0.010 | 0.043 ±0.013 |
| D_{DSC} (Gy) | 0.014 ±0.007 | 0.010 ±0.005 | 0.012 ±0.008 | 0.047 ±0.012 | 0.031 ±0.014 |
| D_{DSC} (Gy) | 0.015 ±0.007 | 0.009 ±0.006 | 0.052 ±0.006 | 0.036 ±0.006 |

RIR – rigid image registration, DIR – deformable image registration, DICE – dice similarity coefficient, HD – Hausdorff distance, MDA – mean distance to agreement. Differences between RIR and DIR metrics were significant (p < 0.001)

| Table 2. DIR accuracy metrics for bladder and rectum |
|---------------------------------------------------------------|
| | Bladder | Rectum |
| | Mean ±SD | Mean ±SD |
| DICE_{RIR} | 0.72 ±0.11 | 0.60 ±0.11 |
| Jaccard_{RIR} | 0.58 ±0.12 | 0.43 ±0.11 |
| HD_{RIR} (mm) | 18.96 ±7.58 | 24.45 ±10.66 |
| MDA_{RIR} (mm) | 4.28 ±2.20 | 5.11 ±2.39 |
| DICE_{DIR} | 0.94 ±0.02 | 0.89 ±0.05 |
| Jaccard_{DIR} | 0.89 ±0.03 | 0.80 ±0.07 |
| HD_{DIR} (mm) | 8.44 ±3.56 | 15.46 ±10.14 |
| MDA_{DIR} (mm) | 0.72 ±0.22 | 1.19 ±0.59 |

RIR – rigid image registration, DIR – deformable image registration, DICE – Dice similarity coefficient, HD – Hausdorff distance, MDA – mean distance to agreement for extreme outlier data were 103 and 79 cc, respectively. The median volume differences for bladder and rectum for all patients were 50 and 20 cc, respectively.
patients treated with 3 or 5 BT fractions. They found that there were no significant differences in terms of \( D_{0.1cc} \), \( D_{1cc} \), \( D_{2cc} \), and \( D_{5cc} \) between SS and RIR-based summation for bladder and rectum in both BT fractionation regimes. Greater differences were found with 5 versus 3 fractions for RIR, suggesting that with an increased number of treatment fractions, the uncertainty due to interfractional changes grows. Jamema et al. [31] compared the intensity and contour-based DIR for DA. They found that the contour-based approach had greater accuracy compared to the intensity-based DIR. The intensity-based algorithm does not include a regularization term to consider the mechanical properties of OAR walls, causing systematic underestimation of the dose. In addition, the presence of a Foley catheter in the bladder and gas in the rectum affect the DIR process, which can be eliminated by masking contours to the special Hounsfield unit number.

Validation of image registration is a challenging task. There are several methods to validate the image registration including propagation of the same points or contours between registered images, registration of images of physical or virtual phantoms, and patient image registration. Each of them involves certain issues including construction of a physical phantom and design of a virtual phantom to simulate complexity of the real situation, uncertainty in identifying boundary of contours, finding fiducial markers accurately, effect of sample points,

![Graph A](image1.png)

![Graph B](image2.png)

**Fig. 4.** Differences of DA for \( D_{2cc} \) between DIR-based and SS methods of A) bladder and B) rectum for individual patients

| \( D_{0.1cc} \) (Gy) | -4.08 (-1.59, -9.69) | -2.26 (-0.97, -4.64) |
|---------------------|----------------------|----------------------|
| \( D_{1cc} \) (Gy)  | -2.01 (-1.07, -4.15) | -1.42 (-0.74, -2.56) |
| \( D_{2cc} \) (Gy)  | -1.53 (-0.86, -2.98) | -1.38 (-0.80, -2.14) |
| \( D_{5cc} \) (Gy)  | -1.18 (-0.65, -2.63) | -1.11 (-0.71, -1.74) |

\( Q_1 \) – first quartile, \( Q_3 \) – third quartile

**Table 3.** Differences of DA between DIR-based and SS methods for bladder and rectum

\( p < 0.001 \)
and third fractions. Thus, the DIR process shrinks hotspot had a greater volume difference (154 cc) in the bladder 

\[ n = 114 \]

which deliver high doses \[15,35\]. One patient (\(n = 1\)) based DA was –9.18 Gy less than SS, which may have 

occurred due to large inter-fraction volume differences. DIR­ 

to a reasonable sample size and hybrid-based DIR 

algorithm with high accuracy.

There are many uncertainties in BT including organ motions (inter- or intra-fractions), applicator displacement after CT scan, differences in filling of bladder/rectum, presence of gas in the rectum, and uncertainty in dose calculation using TG-43 formalism, affecting evaluation of the delivered dose \[33,34\]. Changes in the bladder filling and gas in rectum between fractions cause displacements of 0.1 cc, 1 cc, 2 cc, and 5 cc volumes of OARs which deliver high doses \[15,35\]. One patient (\(n = 114\)) had a greater volume difference (154 cc) in the bladder compared to the first fraction of bladder volume. DIR-based DA was –9.18 Gy less than SS, which may have occurred due to large inter-fraction volume differences. In other words, this case had a small bladder volume in the first fraction and a large bladder volume in the second and third fractions. Thus, the DIR process shrinks hotspot points from large contour to small contour, thereby reducing the \(D_{2cc}\) value. Andersen et al. \[32\] observed that volume differences and \(D_{2cc}\) were significant for bladder. Bladder volume causes changes in the maximum dose to the rectum and sigmoid, and the interfraction volume difference is important in the DIR process. Siavashpour et al. \[35\] suggested an optimum volume for the bladder (70 cc) which decreases doses to OARs. We also suggest 70 cc for the bladder as it reduces the dose to OARs followed by a diminished interfraction volume difference, improving DIR accuracy as reported by a previous study \[15\].

The DIR-based DA of the rectum for the patient 16 was –7.17 Gy less than SS. For this case, there was huge amount of gas inside the rectum in the second fraction compared to the first fraction (the volume difference was 127 cc), Table 4 and Figure 4 show that the number of outlier data of the bladder for each criterion (mild outlier, moderate outliers, extreme outlier) is greater than that for the rectum. One reason might be that the interfraction volume differences of the bladder were larger than those for the rectum.

In some patients, the accumulated dose of \(D_{0.1cc}\) for the bladder was observed to be greater than by the SS approach (supplements Figure 1). It is possibly due to the uncertainty of the hybrid-based DIR algorithm. \(D_{0.1cc}\) of the bladder in these cases was close to the source where the dose gradient is high. On the other hand, the hybrid-based DIR algorithm expands the small bladder contour (second or third Fr) to the large bladder contour (first Fr), causing increased \(D_{0.1cc}\) due to the steep gradient dose region \[31\].

One disadvantage of DICE and Jaccard is that it does not guarantee voxel-to-voxel overlapping inside the contour, though it does not affect the results \[18\]. Wachter-Gerstner et al. \[36\] reported that \(D_{2cc}\) calculated from DVH of the external contour is sufficient to estimate bladder and rectum wall doses as both are hollow organs.

There were some limitations in our study including uncertainty in delineation of organs in CT images, inter-observer contouring uncertainty on OAR, uncertainty of hybrid-based DIR, and assuming that the entire prescription EBRT dose was delivered to OARs. MRI-based BT decreases uncertainty of delineation organs due to better soft tissue contrast. Hellebust et al. \[37\] found 5.8% inter-observer variability using MRI images compared to 10-11% obtained by Saarnak et al. \[38\] using CT images for

| Table 4. Outlier data for differences of \(D_{2cc}\) between DIR-based and SS methods |

| Bladder | Number of included (outlier) patients | Percentage (%) | Rectum | Number of included (outlier) patients | Percentage (%) |
|---------|--------------------------------------|----------------|---------|--------------------------------------|----------------|
| IQR     | –2.12 (Gy)                           |               | –1.33 (Gy)                  |           |                           |
| \(Q_1 + 1.5\) IQR | –6.16 (Gy)                           | 123 (14)      | 89.8 (10.2)                  | –4.125 (Gy) | 129 (8)                   | 94.2 (5.8) |
| \(Q_1 + 2\) IQR | –7.22 (Gy)                           | 128 (9)       | 93.4 (6.6)                   | –4.79 (Gy)  | 134 (3)                   | 97.8 (2.2) |
| \(Q_1 + 2.5\) IQR | –8.28 (Gy)                           | 132 (5)       | 96.4 (3.6)                   | –5.455 (Gy) | 136 (1)                   | 99.3 (0.7) |
| \(Q_1 + 3\) IQR | –9.34 (Gy)                           | 0             | 0                               | –6.12 (Gy)  | 136 (1)                   | 99.3 (0.7) |

IQR – interquartile range, \(Q_1\) – third quartile, total patients – 137
DIR for dose accumulation in brachytherapy

bladder and rectum. It is hard to distinguish to what extent dose differences between two approaches are related to algorithm uncertainty. Therefore all of the results calculated with the DIR algorithm contain DIR uncertainty.

In this study, we investigated DIR only among BT fractions but it is more challenging to combine EBRT and HDR-BT to assess DA due to the presence of the applicator in BT. Also, we did not consider interfraction motion in EBRT, intrafraction motion in BT, or the effect of DIR in radiobiological modeling. Further research is required to confirm our results and especially evaluate the DIR algorithm with ground truth data the same as for the physical phantom. We plan to investigate the feasibility of predicting rectal toxicity with deformable accumulated and dose map features for locally advanced cervical cancer.

Conclusions

We evaluated the hybrid-based DIR algorithm using a virtual phantom and demonstrated low registration uncertainty for both contour matching and dose mapping. Our results demonstrated that differences in DA between DIR-based and SS approaches were statistically significant for both bladder and rectum. Also, the results indicated that the hybrid-based DIR algorithm has potential for DA between brachytherapy fractions.

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Disclosure

The authors report no conflict of interest.

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