Concordance of FDG PET/CT metabolic tumour volume versus DW-MRI functional tumour volume with T2-weighted anatomical tumour volume in cervical cancer

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Abstract

Background: ¹⁸F-fluoro-deoxyglucose positron emission tomography with computed tomography (FDG PET/CT) has been employed to define radiotherapy targets using a threshold based on the standardised uptake value (SUV), and has been described for use in cervical cancer. The aim of this study was to evaluate the concordance between the metabolic tumour volume (MTV) measured on FDG PET/CT and the anatomical tumour volume (ATV) measured on T2-weighted magnetic resonance imaging (T2W-MRI); and compared with the functional tumour volume (FTV) measured on diffusion-weighted MRI (DW-MRI) in cervical cancer, taking the T2W-ATV as gold standard.

Methods: Consecutive newly diagnosed cervical cancer patients who underwent FDG PET/CT and DW-MRI were retrospectively reviewed from June 2013 to July 2017. Volumes of interest was inserted to the focal hypermetabolic activity corresponding to the cervical tumour on FDG PET/CT with automated tumour contouring and manual adjustment, based on SUV 20%–80% thresholds of the maximum SUV (SUVmax) to define the MTV₂₀–₈₀, with intervals of 5%. Tumour areas were manually delineated on T2W-MRI and multiplied by slice thickness to calculate the ATV. FTV were derived by manually delineating tumour area on ADC map, multiplied by the slice thickness to determine the FTV(manual). Diffusion restricted areas was extracted from b₀ and ADC map using K-means clustering to determine the FTV(semi-automated). The ATVs, FTVs and the MTVs at different thresholds were compared using the mean and correlated using Pearson’s product-moment correlation.

Results: Twenty-nine patients were evaluated (median age 52 years). Paired difference of mean between ATV and MTV was the closest and not statistically significant at MTV₃₀ (−2.9cm³, −5.2%, p = 0.301). This was less than the differences between ATV and FTV(semi-automated) (25.0cm³, 45.1%, p < 0.001) and FTV(manual) (11.2cm³, 20.1%, p = 0.001). The correlation of MTV₃₀ with ATV was excellent (r = 0.968, p < 0.001) and better than that of the FTVs.

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Background
Precise determination of cervical tumour boundary is important in radiotherapy to deliver the highest possible radiation dose to cancerous tissues while minimizing that to surrounding healthy tissues.

Given its superior soft tissue contrast, MRI is the modality of choice for the anatomical delineation of tumour outline and local tumour extent, especially in determining whether parametrial invasion is present to differentiate early from advanced stage disease. Despite excellent spatial resolution, delineation of tumour extent can be limited using conventional T2-weighted (T2W) sequences in certain scenarios, e.g., isointense tumours and diffusely infiltrative lesions, in assessing response of tumours to therapy and, in particular, in differentiating residual or recurrent disease from post-treatment fibrosis due to the overlap of morphological appearances [1].

The clinical utilisation of functional imaging in gynaecological malignancy is evolving [2–4]. In the era of more sophisticated treatment options such as image-guided adaptive radiotherapy, functional imaging techniques such as diffusion-weighted MRI (DW-MRI) and ¹⁸F-fluorodeoxyglucose positron emission tomography integrated with computed tomography (FDG PET/CT) have been demonstrated to provide information for more precise definition of radiation target [5–7]. DW-MRI allows characterization of biological tissues based on their water diffusion property that changes with the integrity of cellular membranes and tissue cellularity [8]. Quantitative assessment can be derived from the apparent coefficient diffusion (ADC) maps obtained from DW-MRI [9]. For instance, ADC has been used to differentiate between normal and cancerous cervical tissue, and the latter was found to correlate negatively with tumour cellular density and grading [10, 11]. Additional DW-MRI has been demonstrated in the literature to outperform T2W imaging alone in depicting local recurrence and differentiating it from post-treatment changes such as fibrosis [12–14]. The use of ADC for measuring target volumes with different tissue characteristics for dose prescription in image-guided adaptive brachytherapy [15] and various segmentation methods with DW-MRI [16] have also been studied. However, further investigation in its clinical application to radiotherapy treatment planning is warranted.

FDG PET/CT utilises FDG, a glucose analogue, to provide valuable metabolic information based on the increased glucose uptake and glycolysis of cancer cells, and can depict metabolic abnormalities before morphological alterations occur [17]. FDG PET/CT has been employed to define radiotherapy targets using a threshold based on the standardised uptake value (SUV) for over a decade [18], and that for cervical cancer has been recently demonstrated [19]. Modification of radiation treatment volumes to FDG-avid lymph nodes and primary tumour can facilitate the accurate definition of tissues with metabolically active disease and the avoidance of normal tissue; hence allowing dose boosts to FDG-avid tumour volumes and lower doses to the bone marrow, urinary bladder and rectum [20, 21].

Despite promising results of using functional imaging to delineate radiation target in cervical cancer, the segmentation methods and thresholds used are highly variable in the literature. The aim of this study is to evaluate the concordance between the metabolic tumour volume (MTV) measured on FDG PET/CT and the anatomical tumour volume (ATV) measured on T2W imaging; and compared with the functional tumour volume (FTV) measured on DW-MRI in cervical cancer, using the T2W ATV as gold standard [22].

Methods
Patient selection
The retrospective study was reviewed by local institutional review board and informed consent was waived. We reviewed the local database and all consecutive patients with newly diagnosed cervical cancers who underwent both FDG PET/CT and MRI as pre-treatment imaging from June 2013 to July 2017 were included. Cases with incomplete inclusion of the tumour on MRI were excluded. The median time difference between the two examinations was 4 days (range 0 to 32).

FDG PET/CT
Patient preparation and image acquisition
Whole-body FDG PET/CT (coverage from the skull base to the upper one third of the thighs) was performed on a combined PET/CT scanner (Discovery VCT, 64 multislice spiral CT; GE Healthcare Bio-Sciences Corp.), using a standardised protocol. After 6 h of fasting, 222–370 MBq...
(4.8 MBq/kg) of weight-adjusted FDG was administered intravenously. Following a 60-min uptake time, whole-body emission PET was obtained with 6 bed positions of 2 min and 30 s acquisition time in each bed position. PET was attenuated with CT data and reconstructed with an ordered-subset expectation maximization iterative reconstruction algorithm (14 subsets and 2 iterations) and subsequently fused with CT images for further analysis. The CT imaging parameters were as follows: 120 kVp; 200–400 mA; 0.5 s per CT rotation; pitch, 0.984:1; and 2.5-mm intervals, with or without 60–100 mL (1.5 mL/kg) intravenous contrast medium.

**Metabolic tumour volume (MTV)**
Both SUV and volumetric analysis were performed using Advantage Volume Share on ADW 4.7 workstation (GE Healthcare, Chicago, Illinois, United States). Focal hyper-metabolic activity in the uterine cervix corresponding to the cervical tumour was visually identified, where a 3D volume of interest (VOI) was inserted (Fig. 1). Automated tumour contouring with manual adjustment was performed to include the boundaries of the lesion in the axial, coronal, and sagittal planes, and to avoid the urinary bladder. SUV measurement was performed by normalization of the injected dose to lean body mass. Lean body mass was used for normalization instead of total body mass because it is less dependent on body habitus across populations [23]. Maximum SUV (SUVmax) was automatically generated. MTV was measured using an SUV-based automated contouring program. The voxels presenting SUV ≥ 20% to 80% thresholds of the SUVmax within the contouring margin were incorporated to define the metabolic tumour volumes (MTV20 to MTV80), with intervals of 5%.

**MRI**

**Patient preparation and image acquisition**
Patients were prepared for MRI after 6 h of fasting and 20 mg hyoscine butylbromide (Buscopan, Boehringer Ingelheim, Germany) was given intramuscularly at the start of each examination to reduce bowel peristalsis. All examinations were performed on a 3.0-T MRI system (Achieva 3.0 T TX, Philips Healthcare, Best, the Netherlands) using a dedicated 16-channel phased array torso coil.

The standard sequences included sagittal T2 W turbo spin-echo (TSE) and an oblique axial T2W TSE (perpendicular to the long axis of the cervix). Additional axial T2W TSE was acquired to ensure the same anatomical coverage and slice profile as the DW-MRI. Post-contrast 3D T1 W TSE was acquired after DW-MRI (Table 1).

DW-MRI was performed using single-shot spin-echo echo-planar imaging, immediately after the axial T2W TSE imaging. It was acquired in free breathing with background body signal suppression (presaturation inversion recovery fat suppression) and parallel imaging with sensitivity encoding [SENSE] factor of 2 (Table 1). Image acquisition with 13 b-values (0–1000 s/mm²) were performed in the axial plane covering 20 slices to include the entire cervical cancer, using motion-probing gradients in three orthogonal axes to generate the geometric averaged DW signal. The full inclusion of the entirety of the tumour on the DW-MRI images was confirmed visually for every case.

**Anatomical tumour volume (ATV)**
Tumour areas were manually delineated on T2W images in sagittal and oblique axial planes and multiplied by the slice thickness to calculate the sagittal and oblique axial tumour volumes. Two reviewers, EL (8-year experience in MRI with special interest in gynaecological oncology imaging) and AL (5-year experience in MRI), separately placed the ROIs on the T2W images in the sagittal and oblique axial planes, respectively. The volumes were averaged between the two reviewers to determine the ATV.
Functional tumour volume (FTV)

Averaged DW signal was used to generate the ADC maps using the Levenberg-Marquardt fitting algorithm under the mono-exponential model described by the function:

$$\frac{S_b}{S_0} = e^{-b \cdot \text{ADC}}$$

where $S_b$ represents the mean signal intensity with the diffusion gradient, $b$, $S_0$ is the mean signal intensity when $b = 0$ s/mm$^2$. VOIs were manually drawn by two reviewers, EL and AL, for each lesion. The first set of VOIs were strict manual delineations of the tumour by both reviewers and excluded the surrounding normal tissue based on the hypointense signal of the tumour on the ADC map with cross reference to the axial T2W images. FTV by the two reviewers was then calculated using these VOIs multiplied by slice thickness. The volumes were averaged between the two reviewers to determine the FTV (manual). The second set of VOIs was drawn by the same two reviewers, EL and AL, to include all of the tumour and did not require exclusion of surrounding normal tissue. Volumetric k-means clustering was then calculated using these VOIs multiplied by slice thickness. The volumes were averaged between the two reviewers to determine the FTV (manual). The second set of VOIs was drawn by the same two reviewers, EL and AL, to include all of the tumour and did not require exclusion of surrounding normal tissue. Volumetric k-means clustering was then calculated using these VOIs multiplied by slice thickness. The volumes were averaged between the two reviewers to determine the FTV (semi-automated).

Statistical analysis

The ATVs measured on T2W images, FTVs on DW-MRI and the MTVs at different thresholds on FDG PET/CT were compared. The ATVs, FTVs and MTVs were correlated using Pearson’s product-moment correlation. R version 3.4.1

Table 1 Summary of MRI scan parameters

| Sequences     | T2-Weighted TSE | T2-Weighted SPAIR | T2-Weighted TSE | T2-Weighted TSE | DWI | CE 3D T1-weighted TSE |
|---------------|-----------------|-------------------|-----------------|-----------------|-----|-----------------------|
| Plane         | Sagittal        | Coronal           | Axial           | Oblique Axial   | Axial 3D |                        |
| TR/TE (ms)    | 4000/80         | 3500/80           | 2800/100        | 2800/100        | 2000/54 | 3/1.4                 |
| Turbo factor  | 30              | 21                | 12              | 14              | NA       | NA                    |
| Field of view (mm) | 240 × 240        | 230 × 230          | 402 × 300       | 220 × 220       | 406 × 300 | 370 × 203              |
| Matrix size   | 480 × 298        | 352 × 300          | 787 × 600       | 316 × 311       | 168 × 124 | 248 × 134              |
| Slice thickness (mm) | 4               | 4                 | 4               | 4               | 4       | 1.5                   |
| Intersection gap (mm) | 0               | 0                 | 0               | 0               | 0       | 0                     |
| Bandwidth (Hz/pixel) | 230             | 186               | 169             | 162             | 15.3    | 724                   |
| Number of excitations | 2               | 1                 | 1               | 1               | 2       | 1                     |

CE: contrast-enhanced, DWI: diffusion-weighted imaging; TR/TE: repetition time/echo; TSE: turbo spin echo

Fig. 2 A semi-automated method was used to extract diffusion restricted areas from corresponding $b_0$ and ADC map. VOIs were manually inserted to include the entire tumour. Voxels were automatically separated into 3 groups based on $S_0$ values using a k-means clustering method: solid tumour mass with high cellularity having intermediate ADC and intermediate $S_0$ intensities; normal tissue with low cellularity or cystic tissues having high ADC [5, 24] and high $S_0$ intensities; fat and fibrotic tissues having low ADC low $S_0$ intensities. A study by Gong et al. [25] has shown that slice-by-slice K-means clustering, using both $S_0$ images and ADC, is a promising method for reliable delineation of heterogeneous tumours in patients with metastatic gastrointestinal stromal tumours. FTV (semi-automated) was calculated by discarding the fat and fibrotic cluster and the normal tissue cluster, leaving the solid tumour mass cluster. Parametric map generation and semi-automatic functional volume segmentation were performed using in-house programs using MATLAB (The Mathworks Inc., Natick, MA, USA) (Fig. 2). The volumes were averaged between the two reviewers to determine the FTV (semi-automated).
was used for statistical analysis. A two-tailed $p$-value < 0.05 was considered statistically significant.

**Results**

**Demographics**
A total of 29 patients were evaluated with median age of 52 years (range 27–76 years). Further clinicopathological characteristics were tabulated in Table 2.

**Quantitative measurements**
Mean SUVmax of the cervical tumours was 9.2, range 3.3–16.7. Mean ADC of the cervical tumours was 0.934 ± 0.120 mm²/s. The mean ATVs measured in sagittal and oblique axial planes are 51.7 and 59.3 cm³ respectively.

**Paired differences of mean between (ATV and FTV) vs (ATV and MTV at different SUV thresholds)**
The paired difference of mean between ATV and MTV₃₀, mathematically represented by mean ATV – mean MTV₃₀, was −2.9 cm³, −5.2%, $p = 0.301$. This difference was not statistically significant and was the closest to ATV compared with all other FTVs and MTVs measured at other SUVmax thresholds, including the differences between ATV and FTV(semi-automated) (mean ATV – mean FTV(semi-automated) = 25.1 cm³, 45.1%, $p < 0.001$) and between ATV and FTV(manual) (mean ATV – mean FTV(manual) = 11.2 cm³, 20.1%, $p = 0.001$). The means of the ATV, FTV(semi-automated), FTV(manual) and MTV₂₀ to MTV₈₀ (with intervals of 5%) and the differences of their means with ATV are shown in Table 3.

**Correlation**
The correlations of MTV₂₀–₅₀ with ATV were excellent ($r > 0.9$, $p < 0.001$), and those of MTV₃₀–₄₀ were better than that with FTV(semi-automated) and FTV(manual). There was a gradual decline in correlation of MTVs with ATV as the percentage threshold of SUVmax increased (Table 4).

**Discussion**
Our study demonstrated that among all metabolic threshold levels and FTV(semi-automated) and FTV(manual), the MTV₃₀ had the least absolute difference from ATV and was the only parameter investigated which did not show a statistically significant difference from ATV. In addition, MTV₃₀ showed excellent positive correlation with ATV.

MRI is indispensable in the local disease assessment of cervical cancer. The ability of combined functional volume assessment and local disease extent using MRI alone could present as a promising imaging algorithm for patients with cervical cancer. However, the evidence to support this is limited in the current literature and most studies have used the manual segmentation method based on DW-MRI images [26, 27].

The choice of imaging modality used in tumour contouring or segmentation technique can result in varying derived tumour volume [16, 28, 29]. There is no consensus of the methodology of tumour segmentation using DW-MRI or ADC values. Clustering is a method, which groups similar data, and the k-means algorithm is often chosen for image segmentation and grouping voxels of same signal intensities and has been used for classification of functional imaging data [30]. This algorithm is

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**(Table 2)**

| Median age in years (range) | 52 (27–76) |
|----------------------------|------------|
| No of patients, n (%)      |            |

| FIGO stage | IB | 8 (27.6%) |
|------------|----|-----------|
|            | II | 2 (6.9%)  |
|            | II | 7 (24.1%) |
| III | 10 (34.5%) |
| IV  | 1 (3.4%)  |
| Unstaged | 1 (3.4%) |

| Histology | Squamous cell carcinoma | 16 (55.2%) |
|-----------|-------------------------|------------|
|           | Adenocarcinoma          | 8 (27.6%)  |
|           | Others                  | 5 (17.2%)  |

**(Table 3)**

|            | Mean (cm³) | Paired difference with mean ATV (cm³) | (%)  | $p$ value |
|------------|------------|--------------------------------------|------|-----------|
| ATV        | 55.5       |                                      |      |           |
| FTV(manual)| 44.3       | 11.2                                 | 20.1%| 0.001     |
| FTV(semi-automated) | 30.5 | 25.1                                 | 45.1%| <0.001    |
| PET MTV₂₀ | 83.3       | −27.8                                | −50.0%| <0.001    |
| PET MTV₂₅ | 69.2       | −13.7                                | −24.7%| 0.001     |
| PET MTV₃₀ | 58.4       | −2.9                                 | −5.2%| 0.301     |
| PET MTV₃₅ | 49.0       | 6.5                                  | 11.8%| 0.026     |
| PET MTV₄₀ | 41.4       | 14.1                                 | 25.3%| 0.001     |
| PET MTV₄₅ | 34.9       | 20.6                                 | 37.1%| <0.001    |
| PET MTV₅₀ | 29.3       | 26.2                                 | 47.2%| <0.001    |
| PET MTV₅₅ | 24.3       | 31.2                                 | 56.2%| <0.001    |
| PET MTV₆₀ | 19.4       | 36.1                                 | 65.0%| <0.001    |
| PET MTV₆₅ | 15.0       | 40.5                                 | 73.0%| <0.001    |
| PET MTV₇₀ | 11.0       | 44.5                                 | 80.2%| <0.001    |
| PET MTV₇₅ | 7.3        | 48.2                                 | 86.9%| <0.001    |
| PET MTV₈₀ | 4.4        | 51.1                                 | 92.1%| <0.001    |

The differences of their means with ATV. The paired difference of mean between ATV and MTV₂₀ was not statistically significant and was the closest to ATV compared with all other FTVs and MTVs (BOLD)
simple and efficient and has been shown to be able to differentiate ADC values of benign and malignant pathologies [24]. A recent study has shown that K-means clustering is a promising method for reliable tumour delineation of heterogeneous tumours in patients with metastatic gastrointestinal stromal tumours [25]. The relative signal intensity [31] and region growing [32] methods are alternative segmentation techniques which were described to have limitations related to their dependence on $b$-value and acquisition method for DW-MRI images, and sensitivity to signal-to-noise ratio [16].

FDG PET/CT has the advantage of identifying the metabolic activity and providing information on tumour biology. It is increasingly recognized as a useful tool for directing RT planning during intensity-modulated radiation therapy, volumetric-modulated arc treatment [20] and image-guided brachytherapy [33], thereby allowing targeted dose escalation to target tissues with high metabolic activity, and reducing dose to surrounding tissues [34]. The utility of FDG PET/CT has been shown to lead to less gastrointestinal toxicity in patients with gynaecological malignancies [35].

Various tumour segmentation techniques using FDG PET/CT exist: manual contouring, which is based on visual assessment for determining tumour outline; thresholding, which uses a minimum SUV value to identify target; and gradient edge detection, in which tumour delineation is based on the changes in signal across a given area [36]. As SUV thresholding has been the focus in initial investigative approaches and is the most commonly employed method of FDG/PET-based tumour volume segmentation [18], it was the segmentation technique of choice in this study.

Table 4 Correlation of MTVs at all SUVmax thresholds were significantly correlated with T2W anatomical volume and with DW-MRI functional volume ($p < 0.05$)

| ATV       | Pearson’s product-moment correlation (r) | $p$ value |
|-----------|-----------------------------------------|-----------|
| FTV(semi-automated) | 0.963                                  | <0.001    |
| PET MTV_{30} | 0.955                                  | <0.001    |
| PET MTV_{25} | 0.962                                  | <0.001    |
| PET MTV_{30} | 0.968                                  | <0.001    |
| PET MTV_{35} | 0.975                                  | <0.001    |
| PET MTV_{40} | 0.968                                  | <0.001    |
| PET MTV_{45} | 0.948                                  | <0.001    |
| PET MTV_{50} | 0.921                                  | <0.001    |
| PET MTV_{55} | 0.877                                  | <0.001    |
| PET MTV_{60} | 0.830                                  | <0.001    |
| PET MTV_{65} | 0.803                                  | <0.001    |
| PET MTV_{70} | 0.777                                  | <0.001    |
| PET MTV_{75} | 0.777                                  | <0.001    |
| PET MTV_{80} | 0.781                                  | <0.001    |

Pearson correlation coefficient (r) larger than 0.9 are highlighted in bold. ATV: anatomical tumour volume; FTV: functional tumour volume; MTV: metabolic tumour volume

In conclusion, MTV delineation on FDG PET/CT appears promising and superior as a functional imaging modality when compared with DW-MRI in tumour contouring with MTV_{30} being the best correlate to ATV.
Abbreviations
ADC: apparent diffusion coefficient; ATV: anatomical tumour volume; CE: contrast-enhanced; DW-MRI: diffusion weighted magnetic resonance imaging; FDG PET/CT: 18F-fluoro-deoxyglucose positron emission tomography with computed tomography; FTV: functional tumour volume; MTV: metabolic tumour volume; PET: positron emission tomography; SUV: standardised uptake value; SUVmax: maximum standardised uptake value; TR/TE: repetition time/echo; TSE: turbo spin echo; VOI: 3D volume of interest

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Availability of data and materials
The datasets used and/or analysed during the current study are available from the corresponding author on reasonable request. These data are primary data obtained by one of the authors.

Authors’ contributions
AL, the first author, segmented and calculated the tumour volumes on DW-MRI; segmented and calculated the tumour volumes on PET/CT and T2W-MRI; analysed the data; and was a major contributor in writing the manuscript. JP provided skills in automated tumour volume segmentation and calculation on MRI; segmented and calculated the tumour volumes on DW-MRI; analysed the data; and was a major contributor in writing the manuscript. XX and EH provided skills in automated tumour volume segmentation and calculation on MRI; and contributed in editing the manuscript. EL, the senior author, obtained ethics approval for the study; retrieved the patient list; segmented and calculated the tumour volumes on DW-MRI; segmented and calculated tumour volumes on PET/CT and T2W-MRI; and was a major contributor in writing the manuscript. All authors read and approved the final manuscript.

Ethics approval and consent to participate
Ethics approval was obtained from Institutional Review Board of the University of Hong Kong/Hospital Authority Hong Kong West Cluster (HKU/HA HKUC IRRB, IRRB Reference Number UW 16–521). The study is a retrospective study involving human data that had already been collected and did not require additional recruitment of human subjects, waiving the need for informed consent.

Consent for publication
Not applicable.

Competing interests
The authors declare that they have no competing interests.

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