A Noninvasive Body Setup Method for Radiotherapy by Using a Multimodal Image Fusion Technique

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Abstract

Purpose: To minimize the mismatch error between patient surface and immobilization system for tumor location by a noninvasive patient setup method. Materials and Methods: The method, based on a point set registration, proposes a shift for patient positioning by integrating information of the computed tomography scans and that of optical surface landmarks. An evaluation of the method included 3 areas: (1) a validation on a phantom by estimating 100 known mismatch errors between patient surface and immobilization system. (2) Five patients with pelvic tumors were considered. The tumor location errors of the method were measured using the difference between the proposal shift of cone-beam computed tomography and that of our method. (3) The collected setup data from the evaluation of patients were compared with the published performance data of other 2 similar systems. Results: The phantom verification results showed that the method was capable of estimating mismatch error between patient surface and immobilization system in a precision of <0.22 mm. For the pelvic tumor, the method had an average tumor location error of 1.303, 2.602, and 1.684 mm in left–right, anterior–posterior, and superior–inferior directions, respectively. The performance comparison with other 2 similar systems suggested that the method had a better positioning accuracy for pelvic tumor location. Conclusion: By effectively decreasing an interfraction uncertainty source (mismatch error between patient surface and immobilization system) in radiotherapy, the method can improve patient positioning precision for pelvic tumor.

Keywords

patient setup, radiotherapy, mismatch error between patient surface and immobilization system, multimodal image fusion, tumor location

Abbreviations

CBCT, cone-beam computed tomography; CT, computed tomography; 3D, 3-dimensional; IR, infrared; MBSI, mismatch error between patient surface and immobilization system; MSA, modified simplex algorithm; MVCT, megavoltage computed tomography

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Introduction

One important process of radiotherapy is to accurately retain tumor inside a treatment field, which depends on precise and reproducible immobilization setup of patient. The mismatch error between patient surface and immobilization system (MBSI) is an obvious interfraction uncertainty in fractionated irradiation.\(^1\)\(^-\)\(^5\) It may cause geometrical error of 2 to 5 mm\(^3\)\(^4\),\(^6\) in the tumor position during treatment and may degrade the effectiveness of radiation treatment. With the tumor location error, a high dose will be delivered to the healthy cells and tissues near the treatment area. As a result, side effect may occur, such as rectal bleeding, incontinence for the pelvis, and heart complications for the chest. Therefore, we propose a feasible solution to minimize this error.

Clinically, gantry-mounted X-ray imaging systems\(^7\)\(^-\)\(^15\) can be used to correct the MBSI when guiding patient setup. With X-ray images, radiation therapists can see the tumor by naked eyes and can directly relate it to the treatment machine isocenter. Usually, the cone-beam computed tomography (CBCT) is used to capture the internal anatomy images and then to be compared with the planning computed tomography (CT). Or the kV X-ray imaging can be used to figure several fiducial markers that are implanted around the tumor. Thereby, the therapists/clinicians can objectively align the tumors using image registration. For the simplicity of implementation and the reduction of X-ray imaging times, a hybrid system (combining infrared [IR] camera and X-ray imaging technique), such as ExacTrac/Novalis Body system,\(^16\)\(^-\)\(^20\) is developed. The IR camera technique is used for an initial patient setup and the target is accurately located by an X-ray imaging system. Furthermore, some novel techniques\(^21\)\(^-\)\(^23\) only use IR cameras to provide real-time feedback on patient setup errors.

In this article, we also present a noninvasive body setup method based on IR cameras, but our method is capable of commonality and generalization. It requires minimum equipment support and is not sensitive to the minor changes in body shape. The method uses the planning CT data to reconstruct a 3-dimensional (3D) surface. By comparing the surface with the optical surface landmarks, we can re-setup patient until the tumor is properly aligned in the treatment room. For a performance evaluation, we validated the method not only on a phantom but also on the pelvis of patients. We also compared our method with other similar systems.

Methods

Mismatch Error Between Patient Surface and Immobilization System

For optical guidance techniques in tumor location, MBSI would cause a deviation from the estimated tumor to the actual one. As shown in Figure 1, it is because that MBSI can change the IR marker’s position from its original one that is seen in the planning CT series. Accordingly, the bimodal image fusion based on the IR markers would be different from the truth.

In our work, we quantify MBSI as the tumor location error caused by the mismatch. Explicitly, MBSI is measured as the geometric shift between the true tumor position and the theoretical one. The true tumor position is acquired from the CT scans. The theoretical position is inferred from a reconstructed optical surface, based on the IR marker–based fusion. This would be introduced with more details in the section “Determination of an Optimal Transformation Matrix”.

The Noninvasive Body Setup Method

The noninvasive body setup method is comprised of 2 steps. Firstly, we use an iterative optimization to find an optimal transformation matrix \(P\). Based on \(P\), we merge the CT images with the optical data into a single image, and the MBSI is estimated. Secondly, patient setup is guided by the setup errors, which are calculated in the fused image.

Determination of an optimal transformation matrix. Figure 2 exhibits the first step of our method. To optimize \(P\), its initial value \(P\) is determined by resorting to the 6 IR markers that are placed on the immobilization system (eg, thermoplastic mask and vacuum cushion). For conciseness, only 2 IR markers are plotted in Figure 2. The calculation has been previously introduced in the work of Zhang et al.\(^24\) In brief, patients undergo CT scans with IR markers and immobilization system. By using an optical positioning system (Northern Digital Inc, Waterloo, Ontario, Canada) to detect these markers, we can relate the CT data space to the optical space. The initial \(4 \times 4\) transformation matrix \(P\) from the optical space to the CT space can be acquired as
\[ P = \begin{pmatrix} R & t \\ 0 & 1 \end{pmatrix}. \]  

(1)

where \( R \) is a 3 \times 3 rotation matrix with 3 parameters of \((\varphi, \theta, \gamma)\). \((\varphi, \theta, \gamma)\) are defined as the angles around the \(x\), \(y\) and \(z\) axes, respectively. \( t \) is a 3 \times 1 translation vector encompassing 3 elements of \([t_x, t_y, t_z]^T\) which are in the \(x\), \(y\) and \(z\) directions, respectively.

As exhibited in Figure 2D, because of MBSI, the optical surface landmarks cannot be transformed by \(P\) to match with the CT reconstructed surface. To correct it, we use 6 geometrical parameters of \(\Delta \varphi, \Delta \theta, \Delta \gamma, \Delta t_x, \Delta t_y, \Delta t_z\) to describe MBSI and to fine-tune \(P\). Here, \(\Delta \varphi, \Delta \theta\) and \(\Delta \gamma\) are 3 rotation errors around \(x\), \(y\) and \(z\) axes, respectively, and \(\Delta t_x, \Delta t_y\) and \(\Delta t_z\) are 3 translation errors along \(x\), \(y\) and \(z\) axes, respectively. The optimal transformation matrix \(\hat{P}\) is expressed as

\[ \hat{P} = \begin{pmatrix} \hat{R} & \hat{t} \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} \Delta R \cdot R & \Delta R \cdot t + \Delta t \\ 0 & 1 \end{pmatrix}, \]  

(2)

in which the 3 rotation angels of \(\Delta R\) are \(\Delta \varphi, \Delta \theta\) and \(\Delta \gamma\), and the 3 parameters of \(\Delta t\) are \(\Delta t_x, \Delta t_y\) and \(\Delta t_z\). \(\hat{P}\) is estimated through
co-registration of 2 surfaces. One reference surface is extracted from the planning CT scans, denoted as a point set \( \Psi \). The other surface, defined as a point set \( \Phi \), is constituted by \( M \) optical surface landmarks. As exhibited in Figure 2C, the optical surface landmarks are detected using a 3D position detection probe (an accessory of the optical positioning system; Northern Digital Inc). When we apply the probe to touch a surface point, the position of the probe tip is transferred to a computer and is recorded as an optical surface landmark. The 2 point sets are defined as

\[
\Psi = \{ \psi_i | i = 1, 2, \ldots, N \},
\]

\[
\Phi = \{ \phi_j | j = 1, 2, \ldots, M \},
\]

where the vector \( \psi_i = (\psi_{x_i}, \psi_{y_i}, \psi_{z_i})^T \) is the \( i \)th 3D spatial surface point in the CT space, the vector \( \phi_j = (\phi_{x_j}, \phi_{y_j}, \phi_{z_j})^T \) is the \( j \)th 3D spatial surface landmarks in the optical space, and \( N > M \). \( \bar{P} \) is optimized by resorting to a one-norm distance transformation technique. The cost function \( C \) is

\[
C = \frac{1}{M} \sum_{j = 1}^{M} D(\phi_j),
\]

\[
D(\phi_j) = \| \bar{R} \cdot \phi_j + \bar{i} - \psi_k \|_1, \quad \text{for} \quad \| \bar{R} \cdot \phi_j + \bar{i} - \psi_k \|_1 \\
\leq \| \bar{R} \cdot \phi_j + \bar{i} - \psi_j \|_1, \quad \forall \psi_j \in \Psi \text{ and } \psi_k \in \Psi.
\]

In equation (6), \( \| \cdot \|_1 \) is defined as one norm. \( \bar{P} \) would reach optimum by iteratively updating \( \Delta \bar{R} \) and \( \Delta \bar{i} \) until \( C \) achieves the minimum value. In our implementation, we used the modified simplex algorithm (MSA) to search for \( \bar{P} \). The initial simplex was constructed by randomly generated vertices. At each iteration of MSA, the trial steps were generated by the operations of reflection (factor \( x \in [0, 1.5] \)), expansion (factor \( r \in [0.2, 0] \)), contraction (factor \( \beta \in [0.5] \)), and shrinkage (factor \( \sigma \in [0.5] \)). The values of \( x, r, \beta, \) and \( \sigma \) can be determined by the users and their default values are \( x = 1, r = 2, \beta = 0.5, \) and \( \sigma = 0.5 \). The search will be terminated if (a) the number of iterations reaches 10 000 or (b) \( C \) is <1.0 mm.

**Calculation of tumor location error.** Before the treatment, the position of the machine isocenter (ie, the tumor’s intended position) is detected using a 3D position detection probe. It can be transformed into the CT space, based on \( \bar{P} \), to be compared with the tumor directly for tumor location error. According to the error, the radiation therapists can adjust the patient setup.

### Mismatch Error Evaluation

#### Phantom Verification

To verify our method on its MBSI estimation ability, we used a phantom (ie, a head and upper thorax region surface reconstructed from the CT data of a real patient) to simulate the mismatch by a computer and applied our method to gauge it.

After applying a known geometric shift to the phantom, the exact tumor position was acquired from it. Likewise, we sampled the phantom for \( M \) optical surface landmarks and transferred them to our method for an estimated tumor position. The Euclidean distance between the exact tumor position and the estimated one was the MBSI estimation error of our method. The phantom experiment was conducted iteratively 100 times and the result was exhibited in the section “Results.”

#### Tumor Location Evaluation

To evaluate our method on its capability of directing patient setup and of decreasing MBSI, 5 patients with pelvic tumor were enrolled into the investigation, including 2 males and 3 females. Tumors were located in the prostate and the rectum. Twelve setup data were collected.

Considering that, in the hospital where the experiment was conducted, the treatment couch did not allow rotational correction, we set \( \Delta \bar{R} \) as \( I_{3 \times 3} \). Our method only updated the translational component \( \Delta \bar{i} \) to search for \( \bar{P} \). Then, by using \( \bar{P} \) to calculate the tumor location error, we repositioned patient by translating the treatment couch.

In detail, patients achieved immobilization using a customized vacuum cushion. Then they were setup using the in-room laser and skin tattoos. The tumor location errors were acquired by a CBCT. This process was performed by a radiation therapist with more than 5-year experience. Subsequently, we applied our method to measure the tumor location error too. The difference in the tumor location errors that were reckoned using CBCT and using our method is statistically summarized in Table 1. Additionally, to assess whether the MBSI correction took effect, the 3D tumor location error (denoted as the Euclidean distance between the tumor and its intended position) was compared with the one acquired by the IR marker–based data fusion.

| Statistics | \( \Delta x \) | \( \Delta y \) | \( \Delta z \) |
|------------|-------------|-------------|-------------|
| Average    | 1.303       | 2.602       | 1.684       |
| SD         | 1.331       | 1.867       | 1.761       |
| Median     | 0.900       | 2.600       | 1.150       |
| P75        | 1.600       | 3.400       | 1.950       |
| P90        | 3.327       | 4.710       | 4.164       |
| P95        | 4.581       | 6.930       | 6.252       |

Abbreviations: P75, 75th percentiles; P90, 90th percentiles; P95, 95th percentiles; SD, standard deviation.

*\( \Delta x, \Delta y \) and \( \Delta z \) denote the translation errors along 3 axes of left-right, anterior-posterior and superior-inferior directions, respectively.

#### Comparison With Other Similar Systems

Two similar systems (ie, surface imaging systems) have been introduced and validated in the study by Wiencierz et al. Five patients with tumors in the pelvic region were available in this
validation. After an initial patient setup with ink marks, the former 2 systems gave their proposal patient shifts based on a captured 3D patient surface. Simultaneously, an experienced radiation therapist resorted a megavoltage CT (MVCT) to suggest an exact patient shift. The tumor location difference between the shift proposed by the similar system and the one derived from the MVCT scans was recorded as an evaluation parameter. Figure 3 exhibited the comparison between the statistical results reported in the study of Wiencierz et al.25 and our results that were collected in the section “Tumor Location Evaluation.”

Results
The phantom verification suggested that the MBSI estimation error ranged from 0.0176 to 0.2248 mm. The statistical results of tumor location evaluation are listed in Table 1. The error distribution has a 75th percentile of 1.6 mm for \(x\)-axis, 3.4 mm for \(y\)-axis, and 1.95 mm for \(z\)-axis and has a 95th percentile of 4.581 mm for \(x\)-axis, 6.930 mm for \(y\)-axis, and 6.252 mm for \(z\)-axis. Ninety-five percent of cases reached to the level of clinical acceptance for the pelvic body region.26-32 Figure 4 exhibited the 3D tumor location errors acquired from our method and from the infrared marker–based data fusion. Except for the data set No. 5 and No. 12, the 3D tumor location errors of other data sets were significantly reduced by our method. The 2 exceptional cases, only showing a small diminution, might be caused by the lower MBSI in their actual instances.

Figure 3 showed the pelvic tumor location performance comparison of our method and other 2 systems. The translation errors of the mentioned 2 systems were reported by Wiencierz et al.25 In Figure 3, our method exhibited a preferable accuracy in all 3 orthogonal directions.

Discussion
The reduction in MBSI can improve the tumor location accuracy. In the phantom verification, our method has been proven to be capable of estimating MBSI in an excellent precision. This point is more evident in Figure 5, that is, a cumulative histogram of the results from the phantom verification. It shows that the MBSI estimation error lies within 0.22 mm and 90% of cases have an estimation error of \(<0.163\) mm. The phantom verification simulates an ideal scenario where the tumor stays immobile relative to the skin and the MBSI only causes a rigid rotation and/or translation from the CT reconstructed surface to the optical one. However, the most practical occasions suggest that a tumor may move slightly under skin and different optical surface landmarks may have different geometrical deformations from their corresponding CT reconstructed surface points. Hence, in the tumor location evaluation, our method was validated on real patients. The region of interest is the pelvic region where surface usually changes for the sake of respiration and soft tissue deformation, such that MBSI generates frequently. The results exhibited in Figure 4 and Table 1 both suggested that MBSI correction provided by our method was effective to
reduce the tumor location error and was promising to offer a clinically acceptable positioning accuracy for the pelvic tumor.

The accuracy comparison among 3 methods, shown in Figure 3, further illustrated that the proposed method was competent for directing patient setup. Compared with the 2 similar systems, the preferable accuracy of our method may result from 3 features: (1) our method can capture a coverage of the whole-body surface area which is significantly larger than other 2 similar systems; (2) the optical surface landmark set used for co-registration is selected artificially to contain bony landmarks as much as possible, for example, the hipbone in the pelvic region; and (3) a rigid registration technique is used to align optical landmarks with the reference surface, by resorting to the bony landmarks. Based on the above 3 features, our method is insusceptible to the minor changes in body shape. The mentioned changes happen instantaneously and are slowly time varying afterward, such as the body shape change caused by different supine postures. It is because that the optical surface is a large-scale point set; furthermore, the set is mainly composed of bony landmarks. For the structural rigidity of bones, the distance between any 2 points has less change among multiple treatment fractions; and a rigid transformation can be used to approximate the relationship between the optical points set and the reference one. Hence, by using our method to merge 2 input images, the resulting image will not be distorted by a small-region surface deformation. Besides, coverage of the whole bony structure, providing more information for the co-registration, further decreases the influence of a local surface deformation. However, when the surface deformation is caused by large intrafractional changes (eg, respiratory motion, daily variations, etc) or significant interfractional variations (eg, notable weight loss), the proposed method fails to find the optimal transformation and a deformable transformation model is required.

To keep up with the increasingly higher requirement of the tumor location accuracy, our future work is to improve our method for a better precision, especially the tumor in soft body region. The solutions can be categorized into 2 areas: the first one is weight loss and the second one is respiration.

**Conclusion**

In the study, we introduced a noninvasive body setup method to guide tumor location with minimizing MBSI. The method was evaluated on a phantom and the pelvis of patients and was also compared with other 2 similar systems. The phantom verification results suggested that our method was considered excellent to estimate MBSI. The tumor location evaluation results indicated that our method could effectively decrease MBSI and 95% of cases achieved clinically acceptable accuracy for the pelvic tumor location. Through the performance comparison with other 2 similar systems, our method was proved to be a promising way to direct patient setup.

**Authors’ Note**

All procedures performed in studies involving human participants were in accordance with the ethical standards of the hospital institutional committee and with the 1964 Helsinki Declaration and its later amendments or comparable ethical standards.

**Declaration of Conflicting Interests**

The author(s) declared no potential conflicts of interest with respect to the research, authorship, and/or publication of this article.

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**References**

1. Coffey M, Holmberg O. Technical issues of safety in a resource-limited environment. In: Dicker AP, Williams TR, Ford EC, eds. Quality and Safety in Radiation Oncology: Implementing Tools and Best Practices for Patients, Providers, and Payers. 1st ed. New York, NY: Springer; 2016:169-178.
2. Surucu M, Shah KK, Roese JC, Choi M, Small W, Emami B. Adaptive radiotherapy for head and neck cancer implications for clinical and dosimetry outcomes. Technol Cancer Res Treat. 2010;16(2):218-223. doi:10.1177/153303461662165.
3. Olch AJ, Lavey RS. Evaluation of a new relocatable head fixation system. In: Schlegel W, Bortfeld T, eds. The Use of Computers in Radiation Therapy. 1st ed. Berlin Heidelberg: Springer; 2000: 594-596. doi:10.1007/978-3-642-59758-9_223.
4. Lang S, Shrestha B, Graydon S, et al. Clinical application of flattening filter free beams for extracranial stereotactic radiotherapy. Radiother Oncol. 2013;106(2):255-259. doi:10.1016/j.radonc.2012.12.012.
5. Chen AM, Daly ME, Cui J, Mathai M, Benedict S, Purdy JA. Clinical outcomes among patients with head and neck cancer treated by intensity-modulated radiotherapy with and without adaptive replanning. Head Neck. 2014;36(11):1541-1546.
6. Gilbeau L, Octave-Prignot M, Loncol T, Renard L, Scalliet P, Grégoire V. Comparison of setup accuracy of three different thermoplastic masks for the treatment of brain and head and neck tumors. Radiother Oncol. 2001;58(2):155-162.
7. Sonke J-J, Rossi M, Wolthus J, van Herk M, Damen E, Belderbos J. Frameless stereotactic body radiotherapy for lung cancer using four-dimensional cone beam CT guidance. Int J Radiat Oncol Biol Phys. 2009;74 (2):567-574. doi:10.1016/j.ijrobp.2008.08.004.
8. Grills IS, Hugo G, Kestin LL, et al. Image-guided radiotherapy via daily online cone-beam CT substantially reduces margin requirements for stereotactic lung radiotherapy. Int J Radiat Oncol Biol Phys. 2008;70(4):1045-1056. doi:10.1016/j.ijrobp.2007.07.2352.
9. Topolnjak R, Sonke J-J, Nijkamp J, et al. Breast patient setup error assessment: comparison of electronic portal image devices and cone-beam computed tomography matching results. Int J Oncol Biol Phys. 2009;74(2):567-574. doi:10.1016/j.ijrobp.2008.08.004.
10. Dawson LA, Jaffray DA. Advances in image-guided radiation therapy. J Clin Oncol. 2007;25(8):938-946. doi:10.1200/JCO.2006.09.9515.

11. Den RB, Doemer A, Kubicek G, et al. Daily image guidance with cone-beam computed tomography for head-and-neck cancer intensity-modulated radiotherapy: a prospective study. Int J Radiat Oncol Biol Phys. 2010;76(5):1353-1359. doi:10.1016/j.ijrobp.2009.03.059.

12. Ottosson W, Baker M, Hedman M, Behrens CF, Sjöström D. Evaluation of setup accuracy for NSCLC patients; studying the impact of different types of cone-beam CT matches based on whole thorax, columna vertebralis, and GTV. Acta Oncol. 2010;49(7):1184-1191. doi:10.3109/0284186X.2010.500303.

13. Simpson DR, Lawson JD, Nath SK, Rose BS, Mundt AJ, Mell LK. A survey of image-guided radiation therapy use in the United States. Cancer. 2010;116(16):3953-3960. doi:10.1002/cncr.25129.

14. Rudat V, Nour A, Hammoud M, Alaradi A, Mohammed A. Image-guided intensity-modulated radiotherapy of prostate cancer. Strahlenther Onkol. 2016;192(2):109-117.

15. Boda-Heggemann J, Lohr F, Wenz F, Flentje M,uckenberger M. kV cone-beam CT-based IGRT. Strahlenther Onkol. 2011;187(5):284-291.

16. Graf R, Boehmer D, Budach V, Wust P. Residual translational and rotational errors after kV X-ray image-guided correction of prostate location using implanted fiducials. Strahlenther Onkol. 2010;186(10):544-550. doi:10.1007/s00066-010-2030-8.

17. Gevaert T, Verellen D, Tournel K, et al. Setup accuracy of the Novalis ExacTrac 6DOF system for frameless radiosurgery. Int J Radiat Oncol Biol Phys. 2012;82(5):1627-1635. doi:10.1016/j.ijrobp.2011.01.052.

18. Graf R, Boehmer D, Budach V, Wust P. Interfraction rotation of the prostate as evaluated by kilovoltage X-ray fiducial marker imaging in intensity-modulated radiotherapy of localized prostate cancer. Med Dosim. 2013;37(4):396-400. doi:10.1016/j.meddos.2012.02.006.

19. Graf R, Boehmer D, Nadobny J, Budach V, Wust P. Appropriate patient instructions can reduce prostate motion. Radiat Oncol. 2012;7(1):125-129. doi:10.1186/1748-717X-7-125.

20. Yartsev S, Bauman G. Target margins in radiotherapy of prostate cancer. Br J Radiol. 2016;89(1067):20160312. doi:10.1259/bjr.20160312.

21. Gopan O, Wu Q. Evaluation of the accuracy of a 3D surface imaging system for patient setup in head and neck cancer radiotherapy. Int J Radiat Oncol Biol Phys. 2012;84(2):547-552. doi:10.1016/j.ijrobp.2011.12.004.

22. Gierga DP, Riboldi M, Turcotte JC, et al. Comparison of target registration errors for multiple image-guided techniques in accelerated partial breast irradiation. Int J Radiat Oncol Biol Phys. 2008;70(4):1239-1246. doi:10.1016/j.ijrobp.2007.11.020.

23. Wasza J, Bauer S, Hornegger J. Real-time motion compensated patient positioning and non-rigid deformation estimation using 4-D shape priors. Med Image Comput Comput Assist Interv. 2012;15(Pt 2):576-583. doi:10.1007/978-3-642-33418-4_71.

24. Zhang J, Ye G, Chen X, Chen Y. Positioning errors of the conventional method in nasopharyngeal carcinoma radiotherapy: a clinical study of an optical patient position guidance system. J Med Imaging Health Inform. 2015;5(3):622-629. doi:10.1166/jmhi.2015.1410.

25. Wiencierz M, Kruppa K, Lüdemann L. Clinical validation of two surface imaging systems for patient positioning in percutaneous radiotherapy. ARXIV. 2016;1602:1-9.

26. Mell LK, Kochanski JD, Roeske JC, et al. Dosimetric predictors of acute hematologic toxicity in cervical cancer patients treated with concurrent cisplatin and intensity-modulated pelvic radiotherapy. Int J Radiat Oncol Biol Phys. 2006;66(5):1356-1365. doi:10.1016/j.ijrobp.2006.03.018.

27. Albuquerque K, Giangreco D, Morrison C, et al. Radiation-related predictors of hematologic toxicity after concurrent chemoradiation for cervical cancer and implications for bone marrow–sparing pelvic IMRT. Int J Radiat Oncol Biol Phys. 2011;79(4):1043-1047. doi:10.1016/j.ijrobp.2009.12.025.

28. Würschmidt F, Petersen C, Wahl A, Dahlé J, Kretschmer M. [18F]fluorocholcholine-PET/CT imaging for radiation treatment planning of recurrent and primary prostate cancer with dose escalation to PET/CT-positive lymph nodes. Radiat Oncol. 2011;6(1):1-8. doi:10.1016/j.radonc.2014.12.019.

29. Dearnaley D, Syndikus I, Sumo G, et al. Conventional versus hypofractionated high-dose intensity-modulated radiotherapy for prostate cancer: preliminary safety results from the CHHiP randomised controlled trial. Lancet Oncol. 2012;13(1):43-54. doi:10.1016/S1470-2045(11)70293-5.

30. Hinton BK, Fiveash JB, Wu X, Dobelbower MC, Kim RY, Jacob R. Optimal planning target volume margins for elective pelvic lymhatic radiotherapy in high-risk prostate cancer patients. ISRN Oncol. 2013;2013:941269. doi:10.1155/2013/941269.

31. Breit A, Heuck A, Lukas P, Kneschaurek P, Mayr M. Tumor Response Monitoring and Treatment Planning: Advanced Radiation Therapy. Munich, Germany: Springer Science & Business Media; 2013.

32. Geinitz H, Roach M III, van As N. Radiotherapy in prostate cancer: innovative techniques and current controversies. Vol 46. Radiation Oncology. 1st ed. Berlin: Springer Berlin Heidelberg; 2015. doi:10.1007/978-3-642-37099-1.