

Clinical association between plan complexity and the local-recurrence-free-survival of non-small-cell lung cancer patients receiving stereotactic body radiation therapy

Chenguang Li^{1,2*}, Shutong Yu^{3,1*}, Junyue Shen^{1*}, Baosheng Liang⁴, Xinhui Fu^{3,1}, Ling Hua^{3,1}, Huimin Hu^{3,1}, Ping Jiang³, Runhong Lei³, Ying Guan⁵, Tian Li⁶, Quanfu Li^{7#}, Anhui Shi^{1#}, Yibao Zhang^{3,1#}

1. Key Laboratory of Carcinogenesis and Translational Research (Ministry of Education/Beijing), Department of Radiation Oncology, Peking University Cancer Hospital & Institute, Beijing 100142, China.
2. Department of Physics and Astronomy, University of British Columbia, 325-6224 Agricultural Road, Vancouver, BC V6T1Z1, Canada
3. Institute of Medical Technology, Peking University Health Science Center, Beijing 100191, China
4. Department of Biostatistics, School of Public Health, Peking University, Beijing 100191, China.
5. Beijing United Family Hospital, Beijing 100015, China.
6. Department of Health Technology and Informatics, The Hong Kong Polytechnic University, Hong Kong 999077, China
7. Department of Medical Oncology, Ordos Central Hospital, Ordos 017000, China

Corresponding authors:

Quanfu Li, Email: 1729259137@qq.com

Anhui Shi, Email: anhuidocor@163.com

Yibao Zhang, Email: zhangyibao@pku.edu.cn

Address: 52 Fucheng Road, Haidian District, Beijing 100142, China.

* Chenguang Li, Shutong Yu and Junyue Shen contributed equally to this work.

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The authors have no relevant conflicts of interest to disclose.

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Abstract

Purpose: To investigate the clinical impact of plan complexity on the local recurrence-free survival (LRFS) of non-small cell lung cancer (NSCLC) patients treated with stereotactic body radiation therapy (SBRT).

Methods: Data from 123 treatment plans for 113 NSCLC patients were analyzed. Plan-averaged beam modulation (PM), plan beam irregularity (PI), monitor unit/Gy (MU/Gy) and spherical disproportion (SD) were calculated. The γ passing rates (GPR) were measured using ArcCHECK 3D phantom with 2%/2mm criteria. High complexity (HC) and low complexity (LC) groups were statistically stratified based on the aforementioned metrics, using cutoffs determined by their significance in correlation with survival time, as calculated using the R-3.6.1 packages. Kaplan-Meier analysis, Cox regression, and Random Survival Forest (RSF) models were employed for the analysis of local recurrence-free survival (LRFS). Propensity-score-matched pairs were generated to minimize bias in the analysis.

Results: The median follow-up time for all patients was 25.5 months (interquartile range 13.4-41.2). The prognostic capacity of PM was suggested using RSF, based on Variable Importance and Minimal Depth methods. The 1-, 2-, and 3-year LRFS rates in the HC group were significantly lower than those in the LC group ($p=0.023$), when plan complexity was defined by PM. However, no significant difference was observed between the HC and LC groups when defined by other metrics ($p>0.05$). All γ passing rates exceeded 90.5%.

Conclusions: This study revealed a significant association between higher PM and worse LRFS in NSCLC patients treated with SBRT. This finding offers additional clinical evidence supporting the potential optimization of pre-treatment quality assurance protocols.

Keywords

NSCLC; SBRT; Local recurrence; Quality assurance; Plan complexity

Introduction

The implementation of Stereotactic Body Radiation Therapy (SBRT) for early-stage non-small-cell lung cancer (NSCLC) patients has improved the local tumor control rates and a reduced incidence of normal tissue complications when compared to conventional fractionated treatment¹. Nevertheless, local recurrence still occurs in a range of 8.1% to 16.1% of NSCLC patients treated with SBRT as their primary therapy². Furthermore, a recent study revealed that 56% of in-field recurrences happened within the central high-dose region³. Therefore, investigating potential biological⁴ and dosimetric factors is of clinical importance to mitigate the risk of recurrence in the future.

The radiobiological effects of radiotherapy are predominantly influenced by the dose distribution in treatment plans. More complex plans can lead to increased disparities between the planned and actual delivered radiation doses⁵. To reduce dosimetric deviation, plan parameters that reflect complexity, such as the γ passing rate (GPR), are routinely monitored as part of pre-treatment quality assurance (QA) procedures, either through dosimetric simulation⁵ or physics measurements using phantoms, detector arrays, or films⁶. GPR serves as an indicator of the consistency between the planned dose and the measured/simulated dose distribution⁵.

In addition to GPR, other less used definitions of plan complexity include plan-averaged beam modulation (PM), plan beam irregularity (PI), and monitor unit/Gy (MU/Gy), etc. PM quantifies the extent to which a large open field is divided into multiple smaller segments. PI characterizes deviations in aperture shapes from a regular circle⁷. MU/Gy reflects the overall modulation level of the plan by the ratio of monitor units to the average dose of the target volume. According to the literature^{8,9}, tumor shape may also influence plan complexity, and spherical disproportion (SD) can be used to reflect the degree to which the tumor shape approaches a sphere. Higher values of PM, PI, MU/Gy and SD, or lower values of GPR, suggest increased beam complexity.

However, it's important to note that physics deviations do not necessarily induce clinical impact. Moreover, the clinical effectiveness and sensitivity of various plan parameters may differ. To bridge the gap between physics-based treatment plans and their clinical consequences, it is crucial to investigate the impact of dosimetric parameters in conjunction with patient outcomes. In the case of NSCLC patients undergoing SBRT, the relationship between plan complexity and local recurrence-free survival (LRFS) remains unclear. This retrospective cohort study aims to explore this clinical relevance, potentially providing complementary evidence to support informed decision-making. For instance, it may help in selecting more effective plan complexity parameters for pre-treatment quality assurance, ultimately improving the clinical outcomes of NSCLC patients treated with SBRT.

Methods and patients

1.1 Physics plan parameters

The ArcCHECK (Sun Nuclear™) 3D phantom was employed to measure the 3D GPR for each plan using 2 mm/2% criteria¹⁰. PM, PI, MU/Gy and SD were calculated using Python 3.7, based on equations (1-2), equations (3-5), equation(6) and equation(7), respectively⁷.

$$BM_i = 1 - \frac{\sum_j (MU_{ij} * AA_{ij})}{MU_i * U(AA_{ij})} \quad (1)$$

$$PM = \frac{\sum_i (BM_i * MU_i)}{MU_p} \quad (2)$$

$$AI_{ij} = \frac{AP_{ij}^2}{4\pi * AA_{ij}} \quad (3)$$

$$BI_i = \frac{\sum_j (MU_{ij} * AI_{ij})}{MU_i} \quad (4)$$

$$PI = \frac{\sum_i (BI_i * MU_i)}{MU_p} \quad (5)$$

$$MU/Gy = \frac{MU_p}{D_{PTV}} \quad (6)$$

$$SD = \frac{A_{GTV}}{\sqrt[3]{36\pi V_{GTV}^2}} \quad (7)$$

where MU_{ij} is the monitor units (MU) of control point j in beam i ; AA_{ij} is the area of all MLC openings at each control point; $U(AA_{ij})$ is the union area of all apertures of

beam i ; AP is the perimeter of MLC aperture. D_{PTV} is PTV mean dose; A_{GTV} and V_{GTV} are the surface area and volume of the gross tumor volume (GTV), respectively.

1.2 Acquisition of clinical data

This single-institution retrospective cohort study was approved by ethics committee (IRB#2020KT155) and was exempted from the requirement for informed consent. A total of 123 treatment plans from 113 NSCLC patients were examined. These patients underwent SBRT treatment according to consistent protocols at our hospital between April 2012 and May 2021. For the 9 patients with multiple targets, separate treatment plans were created for each target due to their relatively large separation. Plan complexity was assessed for each individual plan. The inclusion criteria were as follows: 1) patients with pathologically confirmed NSCLC, or diagnosed as primary NSCLC by multidisciplinary board; 2) patients with stage I-II NSCLC who received no previous local treatments; 3) patients with a lung metastasis smaller than 5 cm and a low burden as well as a good Eastern Cooperative Oncology Group performance status (ECOG-PS)^{11–14}.

1.3 Motion management and treatment planning

All patients were immobilized in supine position using personalized thermoplastic masks (F481B, Guangzhou Renfu Medical Equipment, China). Abdominal compression was applied to reduce respiratory motion for all patients. The GTV was contoured as the macroscopic tumor identified on the planning CT. The internal target volume (ITV) was extended from GTV by incorporating the margins evaluated on dynamic images. Planning target volume (PTV) was created by an isotropic expansion of 5 mm from ITV, compatible with the setup accuracy guided by fractional cone beam CT images and double confirmation by oncologists and physicists. Organs at risk (OARs) were delineated for dose optimization, including the trachea, great vessels, spinal cord, esophagus, heart, lungs, ribs and brachial plexus^{15,16} etc.

As shown in table 1, various PTV prescription dose and fractionation schedules were used in this study, based on a personalized evaluation of the target size, location, and patients' ECOG-PS. Volumetric modulated arc therapy (VMAT) techniques were used

to optimize all treatment plans within the Varian Eclipse treatment planning system, utilizing 6MV or 10MV flattening-filter-free (FFF) photon beams from Varian TrueBeam systems. Density and heterogeneity corrections were applied during the treatment planning stage. All plans aimed to cover at least 95% of PTV with prescription dose. Dose constraints for organs at risk (OARs) were established in accordance with recommendations from RTOG 0915¹⁷, RTOG 0813¹⁶ and EORTC 221133¹⁸ etc.

Table 1 Clinical characteristics of the cohort

Group	Total: 123 plans
Age, Median (Q1-Q3)	69 (63-77)
Gender, n (%)	
Male	71 (57.7%)
Female	52 (42.3%)
Pathology, n (%)	
SCC	21 (17.1%)
AD	85 (69.1%)
NSCLC, NOS	11 (8.9%)
Others ^a	6 (4.9%)
Stage, n (%)	
I	104 (84.6%)
II	4 (3.3%)
IV	15 (12.2%)
Treatment duration (days), Median (Q1-Q3)	10 (7-11)
PTV prescription/fractions (f), n (%)	
48Gy-55Gy/4f	20 (16.3%)
50Gy-60Gy/5f	33 (26.8%)
60Gy-64Gy/8f	68 (55.3%)
Others ^b	2 (1.6%)
PTV mean dose (Gy), Median (Q1-Q3)	60.9 (55.3-64.9)
PTV volume (cm ³), Median (Q1-Q3)	32.3 (20.4-48)
BED10, Median (Q1-Q3)	117.1 (112.4-120.8)
Tumor's location, n (%)	
Central	12 (9.8%)
Peripheral	111 (90.2%)
CCT, n (%)	
Yes	21 (17.1%)
No	102 (82.9%)
PM, Median (Q1-Q3)	0.44 (0.39-0.52)
PI, Median (Q1-Q3)	107.3 (62-167)
LRFS (months), Median (Q1-Q3)	24.2 (13.2-39.8)
Local Recurrence, n (%)	
Yes	15 (12.2%)
No	108 (87.8%)
γ passing rate (%) ^c , Median (Q1-Q3)	96.9 (95.5-97.9)
MU/Gy, Median(Q1-Q3)	31.50(25.70-43.00)
SD, Median(Q1-Q3)	1.279(1.227-1.331)

1. a, including: adenosquamous carcinoma, large cell carcinoma, invasive mucinous adenocarcinoma and carcinoid tumor.
2. b, including: one plan had prescribed dose of 56Gy/7fx, one plan had prescribed dose of 60Gy/9fx.
3. c, γ passing rates (GPR) were measured using ArcCHECK 3D phantom with 2%/2mm criteria.

4. Abbreviations: Q1-Q3=first quartile to third quartile; SCC=Squamous cell carcinoma; AD=Adenocarcinoma; NSCLC, NOS=Non-small cell lung cancer, not otherwise specified; BED10=Biologically effective dose when $\alpha/\beta = 10$; CCT=Whether concurrent chemotherapy or targeted therapy was used; PM=Plan averaged beam modulation; PI=Plan averaged beam irregularity; LRFS=Local recurrence-free survival; SD= Spherical disproportion.

1.4 Propensity score matching

To mitigate bias in this retrospective study, we conducted propensity score matching (PSM) using the MatchIt package¹⁹, ensuring balance in patient baseline covariates, as detailed in Table 2. The nearest neighbor matching method was used to create the synthetic populations with matched baseline covariates. Propensity scores were calculated through a logistic regression model that considered the following variables: age, gender, pathology, stage, BED10 (biologically effective dose calculated using the linear quadratic equation, assuming $\alpha/\beta = 10$)²⁰, tumor's location (central or peripheral), treatment durations between the first and last treatment days, volume of PTV, and the use of concurrent chemotherapy or targeted therapy (CCT). The variables used for PSM included those not directly related to plan complexity but potentially affecting LRFS^{21,22}. These variable selections were aligned with previous relevant researches^{23–28}.

Table 2 clinical characteristics of the propensity-score matched PM cohort

Group	HC (n=47)	LC (n=47)	
Age, Median (Q1-Q3)	68 (63-78)	71 (64-78)	p=0.64
Gender, n (%)			p=1.00
Male	29 (61.7%)	28 (59.6%)	
Female	18 (38.3%)	19 (40.4%)	
Pathology, n (%)			p=0.53
SCC	6 (12.8%)	10 (21.3%)	
AD	38 (80.9%)	34 (72.3%)	
NSCLC, NOS	3 (6.4%)	3 (6.4%)	
Stage, n (%)			P=1.00
I	38 (80.9%)	39 (83.0%)	
II	2 (4.3%)	2 (4.3%)	
IV	7 (14.9%)	6 (12.8%)	
Treatment duration (days), Median (Q1-Q3)	10 (7-11)	10 (7-10)	p=0.78
PTV volume (cm ³), Median (Q1-Q3)	32.3 (19.6-52.7)	34.8 (20.1-47.0)	p=0.82
BED10, Median (Q1-Q3)	117.5 (113.0-120.3)	116.0 (112.7-119.8)	p=0.73
Tumor's location, n (%)			p=1.00
Central	6 (12.8%)	5 (10.6%)	
Peripheral	41 (87.2%)	42 (89.4%)	
CCT, n (%)			p=0.59
Yes	10 (21.3%)	7 (14.9%)	
No	37 (78.7%)	40 (85.1%)	
PM, Median (Q1-Q3)	0.51 (0.46-0.55)	0.37 (0.33-0.41)	P<0.001

Abbreviations: Q1-Q3 = first quartile to third quartile; HC = High complexity plan; LC = Low complexity plan; SCC = Squamous cell carcinoma; AD = Adenocarcinoma; NSCLC, NOS = Non-small cell lung cancer, not otherwise specified; BED10 = Biologically effective dose when $\alpha/\beta = 10$; CCT = Whether concurrent chemotherapy or targeted therapy was used; PM = Plan averaged beam modulation.

1.5 Follow-up protocols

Patients underwent a series of examinations throughout the treatment process to establish a reference baseline. These assessments occurred at the following intervals: before treatment initiation, weekly during treatment, 4-6 weeks after treatment completion, every 3 months for the first 2 years, and every 6 months for the subsequent 3 years. The examinations included chest and abdominal contrast-enhanced CT scans, cranial magnetic resonance imaging (MRI), and superficial

lymph node ultrasound. The evaluation of early and late treatment-related toxicities followed the guidelines outlined in the Common Terminology Criteria for Adverse Events, version 5.0.

Local failure was defined as the occurrence of recurrent or progressed lesions within the PTV region. Determination of failure was made through a meticulous comparison of cross-sectional diagnostic images with the treatment plans. To minimize potential misdiagnoses²⁹, patients exhibiting suspicious changes in CT scans underwent tissue biopsy, thoracic MRI, or positron emission tomography (PET) as necessary. Tumor markers were monitored at each follow-up, and a multidisciplinary board, comprising experienced thoracic radiation oncologists and radiologists, was convened when needed. In this study, patient outcomes were assessed in terms of LRFS, as it is the parameter most clinically relevant to the delivered dose distribution in patient tumors³⁰ and may be influenced by the aperture complexity of SBRT plans⁷.

1.6 Statistical analysis

Fisher's exact test was performed for the categorical data. T-test was used for normally distributed continuous data, otherwise Wilcoxon test was performed. Shapiro-Wilk method was used for the normality test. Pearson correlation test was performed for the continuous data to assess the linear association between two variables, while Spearman rank correlation test was performed when at least one of the two variables was an ordinal categorical variable. Survival curves were determined using Kaplan-Meier method and were compared using Log-rank test. Cox regression and Random Survival Forest (RSF) models were used to estimate hazard ratio and 95% confidence intervals of each factor of interest. A two-sided p-value < 0.05 was considered as statistically significant.

To standardize for the varying absolute magnitudes of each factor, all numeric variables were subjected to Z-score normalization prior to PSM. We conducted one-to-one matching without replacement, utilizing a caliper width of 0.15, and the score-matched pairs were subsequently employed in our analyses.

The high complexity plan (HC) and low complexity plan (LC) groups were

statistically stratified according to complexity metrics respectively. Consistent with previous studies^{31,32}, the cutoff value was determined based on the significance of correlation with survival time. The application used the `coxph` function (`survMisc` package, R-3.6.1) from the survival package to fit a Cox proportional hazard model to the binary (local recurrence) and continuous covariates (local recurrence-free survival time). The cutpoint was then computed using the `cutp` function (`survMisc` package) to achieve a balance between sensitivity and specificity³¹.

Compared to the Cox regression model, the advantage of the RSF model is that it does not require the satisfaction of proportional hazards assumptions and log-linear assumptions³³. Additionally, RSF can be used to assess the importance of various variables in survival predictions. For instance, Variable Importance (VIMP) and Minimal Depth (MD) methods were used to evaluate the significance of each variable.

Results

The median follow-up time was 25.5 months for all patients (interquartile range 13.4-41.2). Local recurrence was observed in 15 targets out of 123 tumor sites (12.2%), which is comparable to the rates of 8.1% - 16.1% reported by other institutions². The 3D GPR of all 123 plans was higher than 90.5% and met clinical requirements. Cohort PM, PI, MU/Gy, GPR and SD varied ranging from 0.27 to 0.80, 14.41 to 470.13, 18.07 to 63.07, 90.5% to 99.8% and 1.101 to 1.993, respectively. More details can be found in Table 1. Fig. 1 displays the correlation coefficients between different variables.

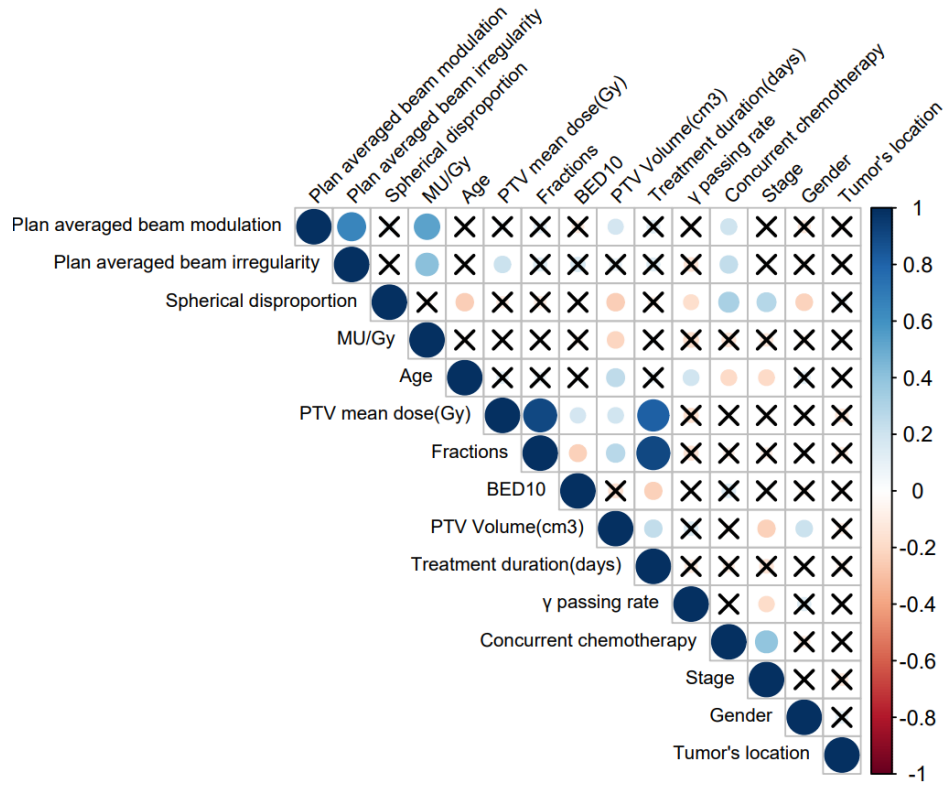


Figure 1. A heatmap showing the Pearson correlation coefficient or Spearman correlation coefficient between different variables. The X marked insignificant values according to the significant level: two-sided p-value < 0.05.

As shown in Table 2, PSM generated 47 data pairs using a cutoff value of 0.43 for PM. The Standard Mean Difference for PSM pairs based on logistic regression was 0.087 (<0.1), indicating a good match between the two groups. No significant differences in the variables between the two groups were observed.

As shown in Fig. 2, PM is an important variable in RSF model-based LRFS prediction, as confirmed by both VIMP (a) and MD (b) methods. When using a multivariable Cox regression model, a violation of the proportional hazards assumption was observed. As depicted in Fig. 3, the 1-, 2-, and 3-year LRFS rates of patients in the HC group were significantly lower than the corresponding rates of the LC group (Log-rank test: $p = 0.023$) when plan complexity was defined by PM using a cutoff value of 0.43. Considering other metrics were not significant in RFS analyses, the corresponding survival analysis results were not provided hereafter.

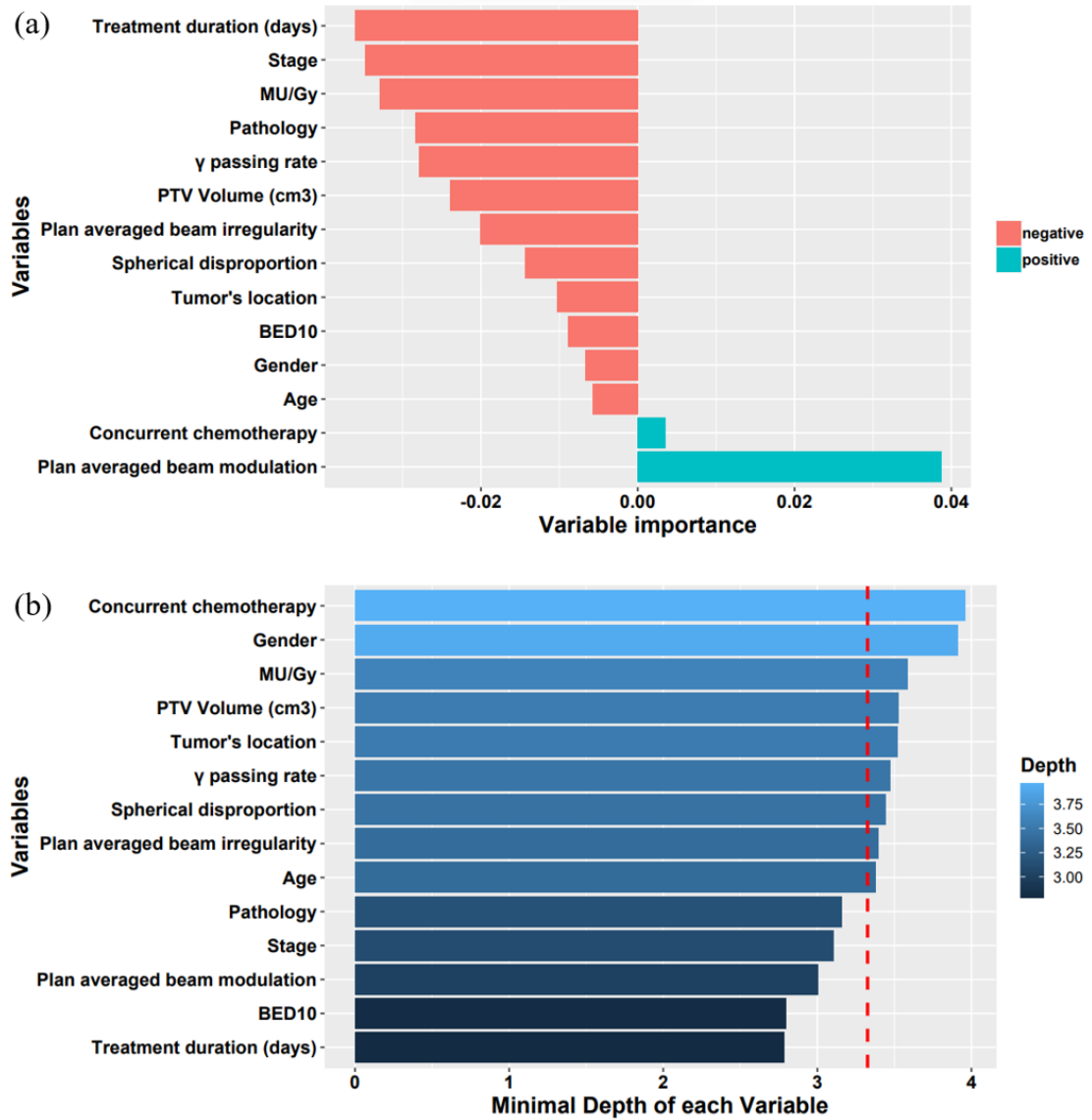


Figure 2. (a) Random Forest Variable Importance (VIMP). The blue and red bars indicate positive and negative VIMP respectively. Positive VIMP means that the variable improves the accuracy of prediction, proportional to the length of bars. (b) Minimal Depth (MD) variable selection. Low MD indicates important variables. The vertical dashed line is the threshold of maximum value for variable selection, which uses the mean of the MD distribution. Variables with MD lower than this threshold are important in forest prediction.

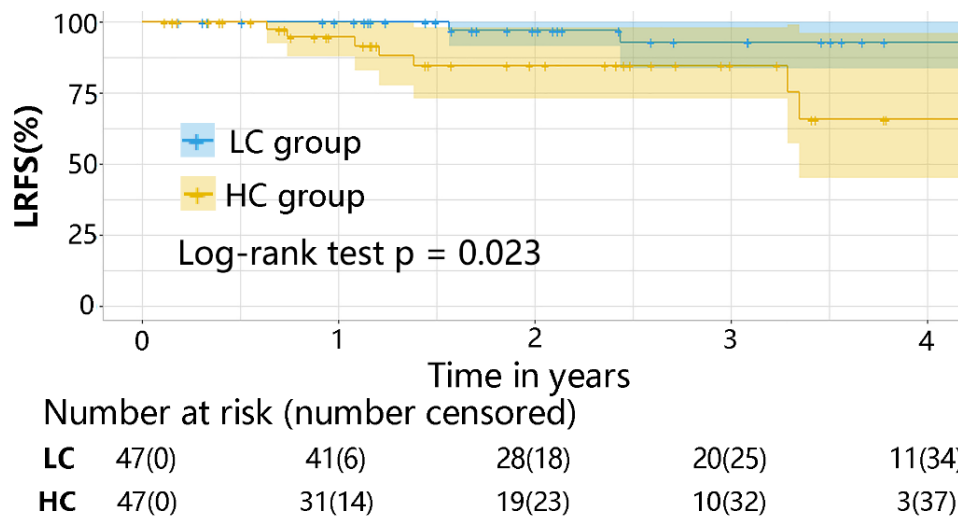


Figure 3. Kaplan-Meier plots of local recurrence-free survival for all patients. The plan complexity was determined by PM: plan averaged beam modulation. The shading areas indicate the 95% confidence intervals for the survival curves. The vertical tick marks on the curves indicate censored patients. Abbreviations: HC=High complexity; LC=Low complexity; LRFS=Local recurrence-free survival.

Discussion

Previous studies reported high local control rates after SBRT, ranging from 88.0% to 97.5% at 2- to 5-years^{1,2,34}, involving various RT regimen and BED10. Several hypotheses have been proposed to explain local recurrences after SBRT, including radiobiological and physical considerations. Radioresistant clonogens within the primary tumor site was a possible explanation to the inconsistent results of improving local control through dose-escalating and increasing BED10⁴. The underlying mechanisms of such resistance may include the proliferation of tumor stem cells during irradiation, the inherent insensitivity result from microenvironment alterations, the hypoxia status in target volume and etc³⁵. However, unlike the radiobiological reasons, physical or technical explanations have been less investigated especially on the basis of clinical data. Recently, Nantavithya analyzed the pattern of local recurrence after SBRT by incorporating dosimetric parameters and geometric information, suggesting that some in-field recurrences might be ascribable to inadequate clearance of tumor clonogens due to underdose³. The LRFS of 90.5% at 2-year and 87.2% at 3-year observed in our study was comparable to previous reports of other centers, where 3-year LRFS ranged from 83.9% to 91.7%^{36,37}. Based on these

clinical data, the impact of physics plan parameters on the local failure after SBRT was investigated retrospectively.

Compared with conventional radiotherapy, SBRT delivers much higher fractional conformal dose to the target³⁸, hence more severe consequences such as recurrence or OAR injury could happen if the high gradient dose missed the target. Respiratory motion, small field, and high dose rate using FFF beams have made it more challenging to perform SBRT for patients with lung cancer. Based on conventional fractionated plans, theoretical simulation³⁹ and dosimetric measurement⁴⁰ have been performed to evaluate the accuracy of dose delivery associated with plan complexity. This work provided additive and complementary clinical evidence that was missing in literature especially for NSCLC patients treated with SBRT, demonstrating significant association between high local recurrence and plan averaged beam modulation (PM) that can be used to guide clinical decision making in the future practice.

The observed significant correlation between higher PM and lower LRFS may be ascribable to the following physical explanations: 1) The combination of small target volume of SBRT cases and smaller beam segmentations in high PM plans increased their misalignment possibility⁴¹, inducing larger deviations from the treatment plans than that of low PM cases; 2) It is more challenging for the smaller beam apertures in the high PM plans to follow the mobile target due to respiration, compromises the concordance between planned and delivered doses. Although fractionation may partially wash out the interplay effect, the less fraction number in SBRT reduced this opportunity⁴¹⁻⁴³; 3) Over modulation may reveal the limitations in the beam models used in treatment planning systems; 4) SBRT delivers much higher fractional dose than conventional treatment, hence Flattening-Filter-Free (FFF) beams of higher dose rates (1400 MU/min or 2400 MU/min) are often used to improve the clinical efficiency, which also amplify the error of dose delivery associated with geometric discrepancy per time unit⁴⁴. These factors may collectively undermine the robustness of delivered dose against motion for lung SBRT patients in a comprehensive way, and consequently reduced the LRFS.

As a broadly used metric for patient specific quality assurance, all measured GPR

values were larger than 90.5% and met clinical requirement. However, GPR was not associated significantly with patient clinical outcomes. As a possible explanation, rigid phantom-based GPR measurements was less sensitive to the plan complexity compared with less repeatable patient anatomies. Our observation of GPR was consistent with many previous studies⁴⁵⁻⁴⁷, suggesting pre-treatment GPR cannot adequately reflect the complex clinical situations. In addition, the results of GPR are dependent on the different measurement approaches and algorithms^{48,49}, which reduces its quantitative consistency. To the contrary, the computation of PM is unique for a specific plan, making it more generalizable cross institutions.

Statistically, this study used PSM to ensure that exposed and unexposed subjects have similar distributions in the confounders, so as to better study the causal effect of exposure⁵⁰. This method is different from multivariable models, which relied more heavily on correctly specified model. The multivariable analysis was not carried out because a violation of the proportional hazards assumption was identified when applying the multivariable Cox regression model. The Cox model presupposes that the hazard ratio between two individuals remains constant over time. If this assumption is violated, the model may yield inaccurate estimates of the hazard ratios, potentially leading to misleading interpretations of the covariates' effects on survival times. Therefore, consistent with previous relevant researches^{51,52}, PSM and RSF models were used in this study.

As for limitations, this retrospective study was based on a relatively small dataset. A prospective and randomized trial based on larger patient volume is desirable in the future to reconfirm the clinical value of PM for pretreatment QA.

Conclusions

Based on the long-term outcomes of NSCLC patients treated with SBRT, this study has contributed clinical evidence indicating that a higher degree of local recurrence is significantly correlated with higher plan-averaged beam modulation (PM). As a result, it suggests that plan complexity should be managed on a patient-specific basis in the future practice of lung SBRT.

References

1. Timmerman RD, Paulus R, Pass HI, et al. Stereotactic Body Radiation Therapy for Operable Early-Stage Lung Cancer: Findings From the NRG Oncology RTOG 0618 Trial. *JAMA Oncol.* 2018;4(9):1263-1266. doi:10.1001/jamaoncol.2018.1251
2. Sun B, Brooks ED, Komaki RU, et al. 7-year follow-up after stereotactic ablative radiotherapy for patients with stage I non-small cell lung cancer: Results of a phase 2 clinical trial. *Cancer.* 2017;123(16):3031-3039. doi:10.1002/cncr.30693
3. Nantavithya C, Gomez DR, Chang JY, et al. An improved method for analyzing and reporting patterns of in-field recurrence after stereotactic ablative radiotherapy in early-stage non-small cell lung cancer. *Radiother Oncol.* 2020;145:209-214. doi:10.1016/j.radonc.2020.01.002
4. Werner-Wasik M, Swann RS, Bradley J, et al. Increasing tumor volume is predictive of poor overall and progression-free survival: secondary analysis of the Radiation Therapy Oncology Group 93-11 phase I-II radiation dose-escalation study in patients with inoperable non-small-cell lung cancer. *Int J Radiat Oncol Biol Phys.* 2008;70(2):385-390. doi:10.1016/j.ijrobp.2007.06.034
5. Younge KC, Matuszak MM, Moran JM, McShan DL, Fraass BA, Roberts DA. Penalization of aperture complexity in inversely planned volumetric modulated arc therapy. *Med Phys.* 2012;39(11):7160-7170. doi:10.1118/1.4762566
6. Miura H, Tanooka M, Inoue H, et al. DICOM-RT Plan Complexity Verification for Volumetric Modulated Arc Therapy. *Int J Med Physics, Clin Eng Radiat Oncol.* 2014;3:117-124. doi:10.4236/ijmpcero.2014.33017
7. Du W, Cho SH, Zhang X, Hoffman KE, Kudchadker RJ. Quantification of beam complexity in intensity-modulated radiation therapy treatment plans. *Med Phys.* 2014;41(2):21716. doi:10.1118/1.4861821
8. Sümer E, Tek E, Türe OA, Şengöz M, Dinçer A, Özcan A, et al. The effect of tumor shape irregularity on Gamma Knife treatment plan quality and treatment outcome: an analysis of 234 vestibular schwannomas. *Sci Rep* 2022;12:21809. <https://doi.org/10.1038/s41598-022-25422-9>.
9. Duan L, Qi W, Chen Y, Cao L, Chen J, Zhang Y, et al. Evaluation of complexity and deliverability of IMRT treatment plans for breast cancer. *Sci Rep* 2023;13:21474. <https://doi.org/10.1038/s41598-023-48331-x>.
10. Miften M, Olch AJ, Mihailidis D, et al. Tolerance limits and methodologies for IMRT measurement-based verification QA: Recommendations of AAPM Task Group No. 218. *Med Phys.* 2018;45(4).
11. Oken MM, Creech RH, Tormey DC, et al. Toxicity and response criteria of the Eastern Cooperative Oncology Group. *Am J Clin Oncol.* 1982;5(6):649-655.
12. Gomez DR, Tang C, Zhang J, et al. Local Consolidative Therapy Vs.

Maintenance Therapy or Observation for Patients With Oligometastatic Non-Small-Cell Lung Cancer: Long-Term Results of a Multi-Institutional, Phase II, Randomized Study. *J Clin Oncol*. 2019;37(18):1558-1565. doi:10.1200/JCO.19.00201

13. Iyengar P, Wardak Z, Gerber DE, et al. Consolidative Radiotherapy for Limited Metastatic Non-Small-Cell Lung Cancer: A Phase 2 Randomized Clinical Trial. *JAMA Oncol*. 2018;4(1):e173501. doi:10.1001/jamaoncol.2017.3501

14. Iyengar P, Kavanagh BD, Wardak Z, et al. Phase II trial of stereotactic body radiation therapy combined with erlotinib for patients with limited but progressive metastatic non-small-cell lung cancer. *J Clin Oncol Off J Am Soc Clin Oncol*. 2014;32(34):3824-3830. doi:10.1200/JCO.2014.56.7412

15. Steber CR, Hughes RT, Soike MH, et al. Five- Versus Ten-Fraction Regimens of Stereotactic Body Radiation Therapy for Primary and Metastatic NSCLC. *Clin Lung Cancer*. 2021;22(1):e122-e131. doi:10.1016/j.clcc.2020.09.008

16. Bezjak A, Paulus R, Gaspar LE, et al. Safety and Efficacy of a Five-Fraction Stereotactic Body Radiotherapy Schedule for Centrally Located Non-Small-Cell Lung Cancer: NRG Oncology/RTOG 0813 Trial. *J Clin Oncol*. 2019;37(15):1316-1325. doi:10.1200/JCO.18.00622

17. Videtic GMM, Hu C, Singh AK, et al. A Randomized Phase 2 Study Comparing 2 Stereotactic Body Radiation Therapy Schedules for Medically Inoperable Patients With Stage I Peripheral Non-Small Cell Lung Cancer: NRG Oncology RTOG 0915 (NCCTG N0927). *Int J Radiat Oncol Biol Phys*. 2015;93(4):757-764. doi:10.1016/j.ijrobp.2015.07.2260

18. Adebahr S, Collette S, Shash E, et al. LungTech, an EORTC Phase II trial of stereotactic body radiotherapy for centrally located lung tumours: a clinical perspective. *Br J Radiol*. 2015;88(1051):20150036. doi:10.1259/bjr.20150036

19. Ho D, Imai K, King G, Stuart EA. MatchIt: Nonparametric Preprocessing for Parametric Causal Inference. *J Stat Softw*. 2011;42(8):1-28. <http://www.jstatsoft.org/v42/i08/>.

20. Fowler JF. The linear-quadratic formula and progress in fractionated radiotherapy. *Br J Radiol*. 1989;62(740):679-694. doi:10.1259/0007-1285-62-740-679.

21. Lin Q, Sun X, Zhou N, Wang Z, Xu Y, Wang Y. Outcomes of stereotactic body radiotherapy versus lobectomy for stage I non-small cell lung cancer: a propensity score matching analysis. *BMC Pulm Med* 2019;19:98. <https://doi.org/10.1186/s12890-019-0858-y>.

22. Liu X, Gao M, Cheng Z, Cai Z, Yu L, Niu G, et al. Stereotactic body radiotherapy compared with video-assisted thoracic surgery after propensity-score matching in elderly patients with pathologically-proven early-stage non-small cell lung cancer. *Precis Radiat Oncol* 2022;6:279–88. <https://doi.org/10.1002/pro6.1175>.

23. Kawaguchi T, Takada M, Kubo A, et al. Performance status and smoking status are independent favorable prognostic factors for survival in non-small cell lung cancer: a comprehensive analysis of 26,957 patients with NSCLC. *J Thorac Oncol.* 2010;5(5):620-630. doi:10.1097/JTO.0b013e3181d2dcd9
24. Kestin L, Grills I, Guckenberger M, et al. Dose-response relationship with clinical outcome for lung stereotactic body radiotherapy (SBRT) delivered via online image guidance. *Radiother Oncol.* 2014;110(3):499-504. doi:10.1016/j.radonc.2014.02.002
25. Senthil S, Haasbeek CJA, Slotman BJ, Senan S. Outcomes of stereotactic ablative radiotherapy for central lung tumours: a systematic review. *Radiother Oncol.* 2013;106(3):276-282. doi:10.1016/j.radonc.2013.01.004
26. McMillan MT, Ojerholm E, Verma V, et al. Radiation Treatment Time and Overall Survival in Locally Advanced Non-small Cell Lung Cancer. *Int J Radiat Oncol Biol Phys.* 2017;98(5):1142-1152. doi:10.1016/j.ijrobp.2017.04.004
27. Stahl JM, Ross R, Harder EM, et al. The Effect of Biologically Effective Dose and Radiation Treatment Schedule on Overall Survival in Stage I Non-Small Cell Lung Cancer Patients Treated With Stereotactic Body Radiation Therapy. *Int J Radiat Oncol Biol Phys.* 2016;96(5):1011-1020. doi:10.1016/j.ijrobp.2016.08.033
28. Alite F, Stang K, Shaikh MP, et al. Local Control Dependence on Consecutive Versus Nonconsecutive Fractionation in Lung SBRT: A Propensity Score Matched Analysis. *Int J Radiat Oncol Biol Phys.* 2015;93(3). doi:10.1016/j.ijrobp.2015.07.1676
29. Faruqi S, Giuliani ME, Raziiee H, et al. Interrater reliability of the categorization of late radiographic changes after lung stereotactic body radiation therapy. *Int J Radiat Oncol Biol Phys.* 2014;89(5):1076-1083. doi:10.1016/j.ijrobp.2014.04.042
30. Yorke E, Gelblum D, Ford E. Patient safety in external beam radiation therapy. *AJR Am J Roentgenol.* 2011;196(4):768-772. doi:10.2214/AJR.10.6006
31. Ogłuszka M, Orzechowska M, Jędraszka D, Witas P, Bednarek AK. Evaluate Cutpoints: Adaptable continuous data distribution system for determining survival in Kaplan-Meier estimator. *Comput Methods Programs Biomed.* 2019;177:133-139. doi:10.1016/j.cmpb.2019.05.023
32. Li C, Li X, You J, et al. Impact of radiation source activity on short- and long-term outcomes of cervical carcinoma patients treated with high-dose-rate brachytherapy: A retrospective cohort study. *Gynecol Oncol.* 2020. doi:https://doi.org/10.1016/j.ygyno.2020.08.037
33. Ishwaran H, Kogalur UB, Blackstone E, Lauer M. Random survival forests. *Ann Appl Stat.* 2008;2:841-860.
34. Fakiris AJ, McGarry RC, Yiannoutsos CT, et al. Stereotactic body radiation

therapy for early-stage non-small-cell lung carcinoma: four-year results of a prospective phase II study. *Int J Radiat Oncol Biol Phys*. 2009;75(3):677-682. doi:10.1016/j.ijrobp.2008.11.042

35. Willers H, Azzoli CG, Santivasi WL, Xia F. Basic mechanisms of therapeutic resistance to radiation and chemotherapy in lung cancer. *Cancer J*. 2013;19(3):200-207. doi:10.1097/PPO.0b013e318292e4e3

36. Lin Q, Sun X, Zhou N, Wang Z, Xu Y, Wang Y. Outcomes of stereotactic body radiotherapy versus lobectomy for stage I non-small cell lung cancer: a propensity score matching analysis. *BMC Pulm Med* 2019;19:98. <https://doi.org/10.1186/s12890-019-0858-y>.

37. Liu X, Gao M, Cheng Z, Cai Z, Yu L, Niu G, et al. Stereotactic body radiotherapy compared with video-assisted thoracic surgery after propensity-score matching in elderly patients with pathologically-proven early-stage non-small cell lung cancer. *Precis Radiat Oncol* 2022;6:279–88. <https://doi.org/10.1002/pro6.1175>.

38. Navarra P, Ascolese AM, Mancosu P, et al. Volumetric modulated arc therapy with flattening filter free (FFF) beams for stereotactic body radiation therapy (SBRT) in patients with medically inoperable early stage non small cell lung cancer (NSCLC). *Radiother Oncol*. 2013;107(3):414-418. doi:10.1016/j.radonc.2013.04.016

39. Hubley E, Pierce G. The influence of plan modulation on the interplay effect in VMAT liver SBRT treatments. *Phys medica an Int J devoted to Appl Phys to Med Biol Off J Ital Assoc Biomed Phys*. 2017;40:115-121. doi:10.1016/j.ejmp.2017.07.025

40. Ong CL, Dahele M, Slotman BJ, Verbakel WFAR. Dosimetric impact of the interplay effect during stereotactic lung radiation therapy delivery using flattening filter-free beams and volumetric modulated arc therapy. *Int J Radiat Oncol Biol Phys*. 2013;86(4):743-748. doi:10.1016/j.ijrobp.2013.03.038

41. Keall PJ, Mageras GS, Balter JM, et al. The Management of Respiratory Motion in Radiation Oncology Report of AAPM Task Group 76. *Med Phys*. 2006;33(10):3874-3900.

42. Edvardsson A, Scherman J, Nilsson MP, et al. Breathing-motion induced interplay effects for stereotactic body radiotherapy of liver tumours using flattening-filter free volumetric modulated arc therapy. *Phys Med Biol*. 2019;64(2):25006. doi:10.1088/1361-6560/aaf5d9

43. Bortfeld T, Jokivarsi K, Goitein M, Kung J, Jiang SB. Effects of intra-fraction motion on IMRT dose delivery: statistical analysis and simulation. *Phys Med Biol*. 2002;47(13):2203-2220. doi:10.1088/0031-9155/47/13/302

44. Netherton T, Li Y, Nitsch P, et al. Interplay effect on a 6-MV flattening-filter-free linear accelerator with high dose rate and fast multi-leaf collimator motion treating breast and lung phantoms. *Med Phys*. 2018;45(6):2369-2376. doi:10.1002/mp.12899

45. Price RA, Veltchev I, Lin T, Eldib A, Chen L, Jin L, et al. Evaluating suggested stricter gamma criteria for linac-based patient-specific delivery QA in the conventional and SBRT environments. *Physica Medica* 2022;100:72–80. <https://doi.org/10.1016/j.ejmp.2022.06.005>.
46. Rajasekaran D, Jeevanandam P, Sukumar P, Ranganathan A, Johnjothi S, Nagarajan V. A study on the correlation between plan complexity and gamma index analysis in patient specific quality assurance of volumetric modulated arc therapy. *Reports of Practical Oncology & Radiotherapy* 2015;20:57–65. <https://doi.org/10.1016/j.rpor.2014.08.006>.
47. Park JM, Kim J, Park S-Y, Oh DH, Kim S-T. Reliability of the gamma index analysis as a verification method of volumetric modulated arc therapy plans. *Radiation Oncology* 2018;13:175. <https://doi.org/10.1186/s13014-018-1123-x>.
48. Hussein M, Clark CH, Nisbet A. Challenges in calculation of the gamma index in radiotherapy - Towards good practice. *Phys medica*. 2017;36:1-11. doi:10.1016/j.ejmp.2017.03.001
49. Woon W, Ravindran PB, Ekayanake P, S V, Lim YY, Khalid J. A study on the effect of detector resolution on gamma index passing rate for VMAT and IMRT QA. *J Appl Clin Med Phys*. 2018;19(2):230-248. doi:10.1002/acm2.12285
50. Chen Q, Nian H, Zhu Y, Talbot HK, Griffin MR, Harrell FEJ. Too many covariates and too few cases? - a comparative study. *Stat Med*. 2016;35(25):4546-4558. doi:10.1002/sim.7021
51. Bellera CA, MacGrogan G, Debled M, de Lara CT, Brouste V, Mathoulin-Pélissier S. Variables with time-varying effects and the Cox model: Some statistical concepts illustrated with a prognostic factor study in breast cancer. *BMC Med Res Methodol* 2010;10:20. <https://doi.org/10.1186/1471-2288-10-20>. 50.
52. Rulli E, Ghilotti F, Biagioli E, Porcu L, Marabese M, D’Incalci M, et al. Assessment of proportional hazard assumption in aggregate data: a systematic review on statistical methodology in clinical trials using time-to-event endpoint. *Br J Cancer* 2018;119:1456–63. <https://doi.org/10.1038/s41416-018-0302-8>.