# Pharmacokinetic Analysis of Hypoxia 18F-Fluoromisonidazole Dynamic PET in Head and Neck Cancer

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This study used pharmacokinetic analysis of <sup>18</sup>F-labeled fluoromisonidazole (18F-FMISO) dynamic PET to assist the identification of regional tumor hypoxia and to investigate the relationship among a potential tumor hypoxia index (K), tumorto-blood ratio (T/B) in the late-time image, plasma-to-tissue transport rate  $(k_1)$ , and local vascular volume fraction ( $\beta$ ) for head and neck cancer patients. Methods: Newly diagnosed patients underwent a dynamic <sup>18</sup>F-FMISO PET scan before chemotherapy or radiotherapy. The data were acquired in 3 consecutive PET/CT dynamic scan segments, registered with each other and analyzed using pharmacokinetics software. The  $(K_i, k_1, \beta)$  kinetic parameter images were derived for each patient. Results: Nine patients' data were analyzed. Representative images of <sup>18</sup>F-FDG PET (of the tumor), CT (of the anatomy), and late-time <sup>18</sup>F-FMISO PET (of the T/B) and parametric images of  $K_i$  (potentially representing tumor hypoxia) are shown. The patient image data could be classified into 3 types: with good concordance between the parametric hypoxia map  $K_i$  and high T/B, with concordant findings between the parametric hypoxia map and low T/B, and with ambiguity between parametric hypoxia map and T/B. Correlation coefficients are computed between each pair of T/ B,  $K_i$ ,  $k_1$ , and  $\beta$ . Data are also presented for other potential hypoxia surrogate measures, for example,  $k_3$  and  $k_1/k_2$ . Conclusion: There is a positive correlation of 0.86 between the average T/B and average hypoxia index  $K_i$  of the region of interest. However, because of the statistical photon counting noise in PET and the amplification of noise in kinetic analysis, the direct correlation between the T/B and hypoxia of the individual pixel is not obvious. For a tumor region of interest, there is a slight negative correlation between  $k_1$  and  $K_i$ , moderate positive correlation between  $\beta$  and  $K_i$ , but no correlation between  $\beta$  and  $k_1$ .

**Key Words:** hypoxia; head and neck cancer; PET; <sup>18</sup>F-FMISO; pharmacokinetics

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E-mail: hummj@mskcc.org COPYRIGHT © 2010 by the Society of Nuclear Medicine, Inc. ypoxia is known as a prognostic factor for many types of cancer, for example, cervical cancer (1), sarcoma (2), and prostate cancer (3). For locally advanced squamous cell carcinoma of the head and neck, tumor hypoxia has been reported to be an important determinant of treatment failure (4-7).

The direct measurement of tumor partial oxygen pressure (pO<sub>2</sub>) can be made using the Eppendorf electrode (Eppendorf AG) or OxyLite probe (Oxford Optronics). With these methods, tumor hypoxia is usually identified as pO<sub>2</sub> values less than or equal to 10 mm Hg. However, pO<sub>2</sub> probe measurement is invasive and may result in sampling error because of the limited number of points that can be feasibly obtained, limited accessibility with some tumors, and an inability to distinguish hypoxic tumor from necrotic tissue. Another method for studying tumor hypoxia is immunohistochemical analysis. However, this method depends on the availability of tumor specimens and again may result in sampling errors associated with the small fraction of tissue analyzed.

PET is a noninvasive approach to image function, such as metabolism, proliferation, blood flow, and hypoxia. PET provides a 3-dimensional quantitative image of the whole tumor, is repeatable during treatment, and may distinguish viable hypoxic tumor from a nonviable necrotic area. Multiple radiotracers have been developed for the purpose of hypoxia imaging: for example, <sup>18</sup>F-labeled fluoromisonidazole (<sup>18</sup>F-FMISO) (8–11). The drug diffuses into and out of the cell in the presence of oxygen but mostly becomes irreversibly bound to macromolecules in the cell under hypoxic conditions (12,13). Although part of the hypoxic binding of the drug is reversible, this effect is small and thus is ignored in this article.

Previous clinical studies on static <sup>18</sup>F-FMISO PET for head and neck cancer patients were performed at the University of Washington. Analysis of the <sup>18</sup>F-FMISO image 90–140 min after injection led to an operational definition of

tumor hypoxia corresponding to a tumor-to-blood activity concentration ratio of greater than 1.2–1.6 (9,14,15). However, the physiologic clearance of <sup>18</sup>F-FMISO from highly perfused normal tissue may result in the same tumor-to-blood ratio (T/B), which is comparable to hypoxic tumor at the time of patient imaging. At 90–180 min after injection, the activity concentration in normal tissue continually decreases as a function of time, whereas in hypoxic tumor it increases as a function of time. The possibility of a cross-over point between the decline in activity from a well-perfused tissue region and the increasing activity from a predominantly hypoxic one can result in an ambiguity in the interpretation of single-time-point imaging.

Therefore, we investigated a kinetic approach, to compare with imaging at a single late time point for identifying regional tumor hypoxia. For this purpose, <sup>18</sup>F-FMISO data were obtained using a dynamic acquisition protocol for use in pharmacokinetic compartmental modeling. The compartmental analysis tool we used was based on a generic 2-tissue model implemented in Philips Voxulus software. Prior investigations using this generic tool for dynamic <sup>18</sup>F-FMISO PET data have been published by Thorwarth et al. (16) and applied for head and neck cancer patients (17). To test the accuracy and precision of this voxelwise approach to compartmental modeling, we used a mathematic dynamic phantom to simulate the <sup>18</sup>F-FMISO kinetics based on predefined compartmental transfer rate constants, initially with no statistical noise (to verify accuracy) and then with gaussian noise comparable to the clinical PET data observed in our patient study group (to verify precision). The results of this simulation study and the impact of the accuracy of the compartmental model by the shift error in the plasma input function are reported by Wang et al. (18).

The aim of this study was to investigate the correlation between 2 potential parametric measures of hypoxia,  $k_3$  and  $K_b$ , and T/B in a late-time <sup>18</sup>F-FMISO image; plasmato-tissue transport rate as a surrogate for local blood perfusion; and local vascular volume fraction in pretreatment patients with head and neck cancer.

#### **MATERIALS AND METHODS**

#### **Patient Selection and Imaging Protocol**

Nine newly diagnosed patients with local squamous cell carcinoma head and neck cancers with no prior chemotherapy or radiotherapy were included in this study. There were 8 men and 1 woman (age range, 50–66 y). The disease stage ranged from I to IV for tumors and I to III for lymph node involvement but without distant metastasis. All patients underwent routine whole-body <sup>18</sup>F-FDG PET/CT to identify the sites of disease. Within 1–3 d, a dynamic <sup>18</sup>F-FMISO PET/CT scan was performed. The <sup>18</sup>F-FMISO imaging protocol was approved by Memorial Sloan-Kettering Cancer Center's Institutional Review Board (IRB 04-070) originally in 2004 and subsequently amended to perform dynamic PET in 2007. All patients were required to provide written informed consent.

#### <sup>18</sup>F-FMISO PET/CT Data Acquisition and Processing

The PET/CT data were acquired on the GE Discovery STE in 2-dimensional mode. Patients were placed supine and fastened with a head-and-shoulder immobilization mask, in the radiotherapy treatment position. A low-dose CT scan (140 kVp, 10 mA) was acquired, with the tumor positioned in the center of a single 15-cm axial field of view. The patient was injected intravenously with 370 MBq of <sup>18</sup>F-FMISO as a bolus, simultaneous with the start of the dynamic PET scan acquisition. The dynamic PET data were acquired in 1-min time frames continuously from 0 to 5 min and in 5-min time frames continuously from 5 to 30 min. The patient was then removed from the scanner and instructed to return at 90 min for a second PET/CT scan and again at 180 min for a third scan. The second and third scans were acquired first with the same low-dose CT, followed by a dynamic PET scan in two 5-min time frames. These 3 low-dose CT scans were used for both PET attenuation correction and registration of 3 series of PET images. The PET data were reconstructed using Fourier rebinning and a 2-dimensional ordered-subset expectation maximization algorithm with attenuation and scatter corrections into 14 dynamic frames  $(128 \times 128 \times 47)$ , with 4.69-mm transverse pixel size and 3.27-mm axial pixel size). A local coregistration technique as implemented in Pinnacle (Phillips) was used to derive the rigid transformation matrix from the second and third CT scans with respect to the first one, which was then applied to the second and third sets of dynamic PET data to generate a registered dynamic series for pharmacokinetic analysis.

#### **Pharmacokinetic Analysis**

The Philips Voxulus software package, which is also included in the research version of Philips' radiation treatment planning software Pinnacle and preclinical software package Imalytics, was used for pharmacokinetic analysis. In pharmacokinetic analysis, the time-varying activity concentration can be partitioned into 3 compartments (19):  $C_p(t)$ ,  $C_1(t)$ , and  $C_2(t)$ , as shown in Figure 1. This partition is only conceptual, not physical. The  $C_p(t)$ ,  $C_1(t)$ , and  $C_2(t)$  are the activity concentration (units are Bq/mL) as a function of time t (unit in min) after injection in the plasma compartment, reversible tissue compartment, and trapped tissue compartment, respectively; and  $k_1$ ,  $k_2$ , and  $k_3$  are the kinetic reaction rate constants (in units of 1/min) between compartments. The rate constants  $k_1$  and  $k_2$  represent the <sup>18</sup>F-FMISO transport rate from the plasma compartment to the reversible tissue compartment and the diffusion rate out of the reversible tissue compartment. At early times,  $k_1$  can be used as a surrogate for blood perfusion when the extraction is close to unity. The  $k_3$ indicates the <sup>18</sup>F-FMISO retention rate constant, from the reversible to the trapped compartment, and is a potential surrogate measure of hypoxia. Another more direct and stable surrogate measure of hypoxia may be the influx rate constant  $K_i = \frac{k_1 k_3}{k_2 + k_3}$  (18).  $K_i$  indicates the direct retention rate constant from the plasma compartment to the trapped compartment without going through the reversible compartment. At late times, if the plasma compartment contribution approaches zero, the activity concentration is proportional to  $K_i$ . The selection of a plasma input function is

FIGURE 1. Generic irreversible 1-plasma, 2-tissue compartmental model.

$$\begin{array}{|c|c|c|}\hline C_p(t) & \stackrel{k_1}{\longrightarrow} & C_1(t) & \stackrel{k_3}{\longrightarrow} & C_2(t) \\\hline \downarrow k_2 & & & & \\\hline \end{array}$$

a requisite step in pharmacokinetic analysis. The plasma input function can be obtained from the arterial blood sampling or from an arterial blood region in the PET images. The plasma compartment in each voxel is only a fraction of the plasma input function. Thus, a vascular volume fraction  $\beta$  (a scale from 0 to 1), compared with the plasma input function, is used to indicate the ratio of vascular volume in tissue at each voxel. For a multiple-voxel tumor region, Voxulus was able to estimate each kinetic parameter as a kinetic image of voxels within the region.

#### **Image-Based Plasma Input Function**

In this study, we used an image-based plasma input function (20-22) instead of blood sampling. The input function region was selected on the common carotid artery from the summed dynamic image at the first 3 min after injection. The diameter of the common carotid artery for a 55-y-old population is about 6.3  $\pm$ 0.6 mm for men and 5.6  $\pm$  0.5 mm for women (23), which is about 1-2 voxels in a PET image. About 20 voxels with highest image intensity in the coronal view of the common carotid artery region were selected for the input function, without correction for the partial-volume effects. The measured input function was then fitted with an analytic formula to capture the correct shape of the input function. The coarse initial frame duration of 1 min was selected to provide adequate statistics when performing voxelby-voxel compartmental analysis. Simulation studies have been conducted to examine the impact of the sharpness of the input function peak and the effect of image noise on the estimation of the rate constants  $k_1$ ,  $k_2$ ,  $k_3$ ,  $\beta$ , and  $K_i$ , and that work has been published separately (18).

#### Region-of-Interest (ROI) Definition and Analysis

Each patient's tumor ROI was defined on the patient's <sup>18</sup>F-FDG PET image with a standardized uptake value greater than 50% of local maximum standardized uptake value, independent of the <sup>18</sup>F-FMISO late-time image or kinetic parametric image. This ROI was then used to derive the average kinetic parameters and other tumor-related statistics from the patient's <sup>18</sup>F-FMISO PET images. The last 10-min time frame of the <sup>18</sup>F-FMISO PET, corresponding to approximately 3 h after injection, was used as the late-time image. The T/B was defined as the ratio of average activity concentration in the tumor ROI versus the plasma input function region at late time. The tumor noise at the late-time imaging was defined as the SD of the voxel intensities within the tumor ROI versus its mean.

## **Scatter Plot and Pearson Correlation Coefficient**

Scatter plots were used as a method to visualize the relationship between 2 kinetic parameters of interest for a set of data points corresponding to tumor ROI or to the whole image space. The Pearson correlation coefficient was computed between 2 kinetic parameters to indicate the strength and direction of a linear relationship between these 2 parameters.

#### **RESULTS**

Table 1 shows the summary of the pharmacokinetic analysis of the 9 head and neck cancer patients. Patients 1–3 have estimated values for the tumor ROI hypoxia index  $K_i$  greater than 0.003 (1/min), whereas the rest of the patients have  $K_i$  less than 0.003 (1/min), suggesting lower hypoxic volumes within patients 4 through 9.

#### Correlation of Tumor Hypoxia and T/B at Late Time

For patients 1–9, the correlation coefficients for individual voxels within the tumor ROI of <sup>18</sup>F-FMISO T/B at a late time and  $K_i$  were 0.46, 0.24, 0.07, 0.10, 0.34, -0.14, 0.30, 0.12, and 0.40. The coefficients for individual voxels suggest that there is no consistent correlation between T/B and  $K_i$  for individual voxels within the tumor ROI. Most of the variation of the T/B and  $K_i$  in the ROI could be due to the statistical noise present in the PET data measurement and propagated through the image reconstruction and pharmacokinetic parameter estimation. Figure 2 displays the scatter plot of the average T/B versus  $K_i$  over the tumor ROI, with each plus sign corresponding to a patient whose number is indicated next to it. In this scatter plot, T/B and  $K_i$  are positively correlated, with a correlation coefficient of 0.86. The dashed line indicates the linear regression between these 2 parameters, where  $K_i = 0.00325 \times T/B - 0.00286$ .

Figure 3 shows the images from patient 2, who exhibited a good concordance between the regions of high T/B in the late-time image and the parametric  $K_i$  image believed to be representative of hypoxia. Figure 3A displays the transverse, coronal, and sagittal views of <sup>18</sup>F-FDG PET, <sup>18</sup>F-FMISO PET at a late time, CT, and the parametric images  $k_{I}$ ,  $K_{i}$ , and  $\beta$ . The tumor ROI is shown in the CT images. Both the late-time <sup>18</sup>F-FMISO PET image and the  $K_i$  image (a hypoxia measure) had high voxel intensities in this ROI, and the  $K_i$  image had higher tumor-to-background contrast than the late-time <sup>18</sup>F-FMISO image. The surrogate perfusion rate constant  $k_1$  image had low voxel intensity in this ROI. There was no clear trend of voxel intensity in the vascular volume fraction β in this ROI, because part of the ROI was close to the common carotid artery. Figure 3B displays the measured and fitted average activity concentration in the tumor ROI as a function of time. Both curves had an early peak at 1-2 min and then decayed and reached a minimum at 20 min; then both curves gradually increased at a later time because more <sup>18</sup>F-FMISO was trapped in the tumor ROI. The estimated average kinetic parameters are also shown in the plot, with  $k_1 = 0.06$  (1/min),  $K_i = 0.0042$ (1/min),  $\beta = 0.25$ , and the average T/B at a late time = 1.74.

Figure 4 shows images from patient 9 and represents a negative finding for hypoxia by PET. This case shows a good concordance between the general low T/B in a late-time image and the corresponding low  $K_i$  values. At the tumor ROI, shown in the CT image in Figure 4A, the average T/B in the late-time image is 1.19, which is not high compared with other regions in the late-time image. The average  $K_i$  in tumor ROI is 0.0009 (1/min), which is in the medium range, compared with other regions. The average  $k_I$  in the tumor ROI is 0.26 (1/min), which is slightly higher than other regions. The average  $\beta$  in the tumor ROI is 0.10, which is also in the medium range, compared with other regions. The tumor ROI time–activity curve, as shown in Figure 4B, had an early peak at 1–2 min similar to that for patient 2, but kept decreasing at a late

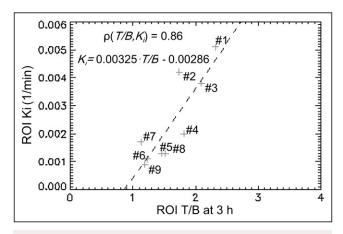
<b>TABLE 1.</b> Summary of Kinetic Analysis of <sup>18</sup> F-FMISO PET in Head and Neck Patients with Squamous Cell Carcinoma											
			Tumor							Tumor	Input
Patient	Tumor		ROI	ROI	ROI	ROI	ROI	ROI k₃	ROI T/B	noise	function
no.	stage	Tumor site	(voxels)	β	$k_1$ (/1 min)	<i>K<sub>i</sub></i> (1/min)	$k_1/k_2$	(1/min)	at 3 h	at 3 h	peak/tail
1	T1N3M0	Pharyngeal	267	0.27	0.19	0.0051	0.65	0.0081	2.32	16%	9.3
2	T3N2M0	Right tonsil	112	0.25	0.06	0.0042	0.42	0.011	1.74	17%	7.3
3	T3N1M0	Right base of tongue	205	0.16	0.30	0.0038	0.86	0.0045	2.09	14%	12.0
4	T4N1M0	Base of tongue	103	0.05	0.16	0.0020	0.94	0.0022	1.81	17%	9.2
5	T2N2M0	Right tonsil	94	0.25	0.39	0.0013	0.85	0.0016	1.46	7%	8.3
6	T1N1M0	Larynx	42	0.15	0.08	0.0011	0.44	0.0026	1.23	11%	6.8
7	T3N2M0	Right pyriform sinus	238	0.04	0.31	0.0017	0.61	0.0031	1.13	9%	4.9
8	T2N2M0	Left base of tongue	114	0.17	0.32	0.0013	0.80	0.0016	1.52	9%	9.4
9	T1N2M0	Right base of tongue	89	0.10	0.26	0.0009	0.73	0.0013	1.19	10%	9.0

time, which was different from patient 2 and suggested little trapping of <sup>18</sup>F-FMISO within the tumor ROI.

However, there were individual cases for which the tumor hypoxia, as reflected by the 2 putative measures that is, late T/B ratio and  $K_i$ —did not result in good intratumoral regional concordance. An example of this is shown in Figure 5, for patient 8, for whom there was an ambiguity between the high intensity voxels in the late T/B image and parametric  $K_i$  image. The T/B in the tumor ROI, as shown in Figure 5A, has moderately high intensity in a region that includes and extends beyond the tumor ROI, with average T/B = 1.36 in the larger region and T/B =1.52 in the tumor ROI. However, only a subregion of the tumor ROI in the  $K_i$  image appeared to be hypoxic. The average  $K_i$  in this subregion was 0.0020 (1/min) and in the tumor ROI was 0.0013 (1/min). The tumor ROI timeactivity curve flats out at a late time in the tumor subregion, as shown in Figure 5B.

# **Correlation of Tumor Hypoxia, Local Blood Perfusion, and Local Vasculature Distribution**

For these 9 patients and over the tumor ROI, the correlation coefficent between the average surrogate perfu-



**FIGURE 2.** Scatter plot of average (T/B,  $K_i$ ) in tumor ROI for 9 patients.

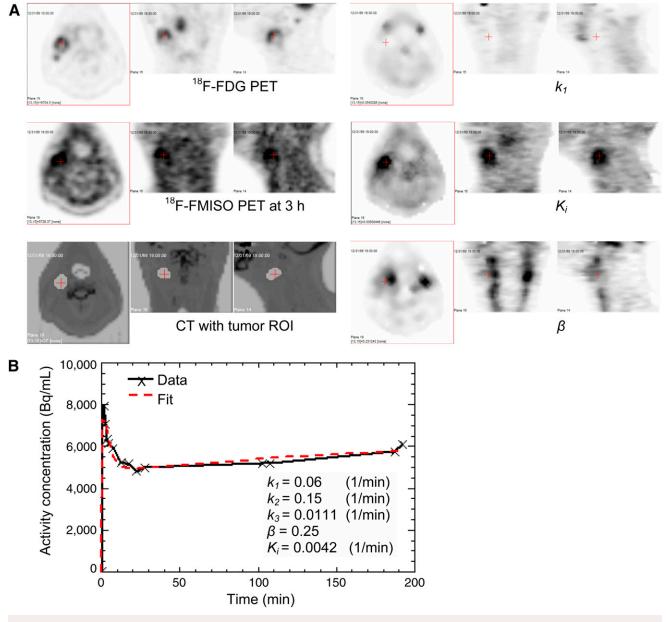
sion rate constant  $k_I$  and the average hypoxia retention rate constant  $K_i$  was -0.33 (slight negative correlation), between the average vascular volume fraction  $\beta$  and average  $K_i$  it was 0.54 (moderate positive correlation), and between the average  $\beta$  and average  $k_I$  it was -0.08 (no correlation).

# **Threshold to Identify Tumor Hypoxia**

The existence of a good correlation between average T/B and  $K_i$  in tumor ROI raises the question of whether there is any threshold to identify hypoxia based on a (T/B,  $K_i$ ) scatter plot. To investigate this, 2 (T/B,  $K_i$ ) scatter plots were created for individual voxels within the tumor ROI and all voxels within the entire head and neck field of view. Figure 6 displays the scatter plots for patients 2 (top row), 9 (middle row), and 8 (bottom row). The left column is the scatter plot for the whole head and neck region (including the tumor ROI). The center column is the scatter plot for voxels corresponding to the tumor ROI only. The scatter plot in the right column is for all of the voxels within the entire head and neck region, but in which the intratumor voxels representing predicted tumor hypoxia are displayed as red triangles and tumor normoxia is represented by the green squares.

For patient 2 (top row), there appears to be significant hypoxia within the whole tumor ROI volume. From Figure 6 (top left), we see that most normoxic tissues lie in the lower left corner of the scatter plot, where dotted points are clustered densely together, with T/B less than 1.6 and  $K_i$  less than 0.00275. The hypoxic tumor ROI values lie in the upper-right corner where the sparse points are, as shown in Figure 6 (top center and top right), where most  $K_i$  are greater than 0.00275 and most triangles lie above the red dashed linear regression line. Therefore, for this patient, the hypoxic tumor region can be well separated from the normoxic normal tissues.

Patient 9 is shown in the middle row, where the whole tumor ROI is normoxic (no hypoxia). The whole head and neck region was densely clustered together. As shown in Figure 6 (middle left), most T/B are less than 1.4 and  $K_i$  is less than 0.00225, and there are almost no sparsely clustered regions. The tumor ROI values lie in the lower right corner of the scatter plot, as shown in Figure 6



**FIGURE 3.** Concordance between high values of parametric hypoxia index  $K_i$  and high T/B in late-time image for patient 2. Transverse, coronal, and sagittal views are shown for <sup>18</sup>F-FDG PET, <sup>18</sup>F-FMISO PET at 3 h, CT fused with tumor ROI,  $k_1$ ,  $K_i$ , and  $\beta$  (A) and measured and fitted average activity concentration in tumor ROI as function of time (B).

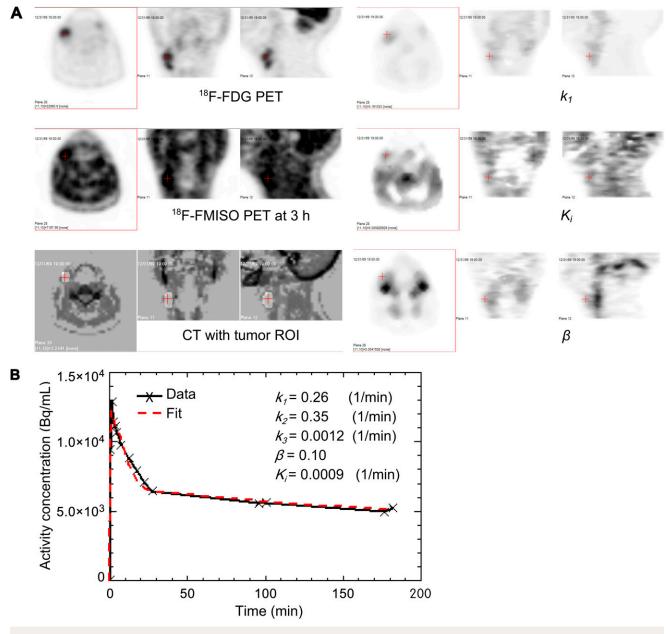
(middle center and middle right), and right below the dashed linear regression line.

Patient 8 is shown in the bottom row, where only a subtumor region is predicted to be hypoxic based on the parametric image data (Fig. 5). The whole head and neck region is densely clustered together in the lower left corner, with most T/B less than 1.4 and  $K_i$  less than 0.002, as shown in Figure 6 (bottom left). The whole tumor ROI has a T/B greater than 1.3 and is divided into 2 subregions by the linear regression line, as shown in Figure 6 (bottom center and bottom right). The subtumor region that is hypoxic (shown with red triangles) is above the linear regression line, and the

subregion that is normoxic (shown with green squares) is mostly below the linear regression line. From these 3 examples, we can deduce that using T/B alone may not be an adequate threshold to identify hypoxia. Instead, we propose that using the  $(T/B, K_i)$  scatter plot in combination with  $K_i$  and linear regression line results in a more robust criterion.

# **DISCUSSION**

Imaging patients at a single late time point after injection is simple and convenient for both clinical staff and the patients. Furthermore, it follows the paradigm of other

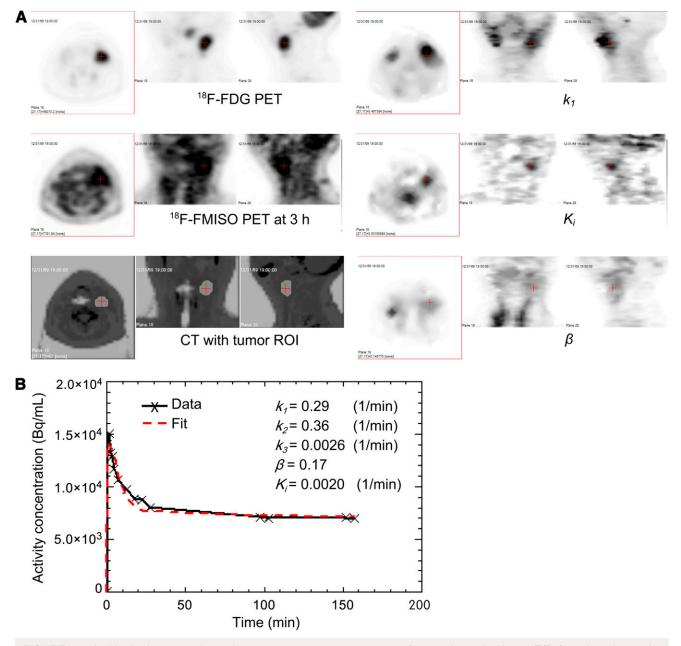


**FIGURE 4.** Concordance between parametric  $K_i$  values suspected to exhibit no or low level of tumor hypoxia and low T/B in late-time image for patient 9. Transverse, coronal, and sagittal views are shown for <sup>18</sup>F-FDG PET, <sup>18</sup>F-FMISO PET at 3 h, CT fused with tumor ROI,  $k_1$ ,  $K_i$ , and  $\beta$  (A) and measured and fitted average activity concentration in tumor ROI as function of time (B).

routine nuclear imaging protocols such as <sup>18</sup>F-FDG PET, where the time selected for imaging after injection considers the balance between competing factors such as tumor uptake, clearance from blood, and count statistics of a decaying radioisotope. For <sup>18</sup>F-FMISO PET, it is necessary to image the patients at least 2.5 h after injection, because of the slow improvement (rise) in contrast between hypoxic and normoxic tissue. For <sup>18</sup>F-labeled hypoxia tracers, count statistics are of concern because of the necessity of late-time-point imaging, if images representing tumor hypoxia are to be achieved. As a consequence, fluctuations in the individual intratumor voxel intensities

can affect the interpretation of which voxels are considered positive versus negative when threshold values are assigned to define regional hypoxia.

In this study, we observed a positive correlation (0.86) between the ROI T/B and the average  $K_i$  for that ROI, a potential hypoxia measurement for the tumor. However, because of the statistical photon counting noise in PET and the amplification of noise in kinetic analysis, the direct voxel-to-voxel correlation between T/B and  $K_i$  serving as a hypoxia index is not obvious. In addition to the statistical fluctuations for each voxel over time, there are registration uncertainties of about 1 CT voxel between the 3 separate

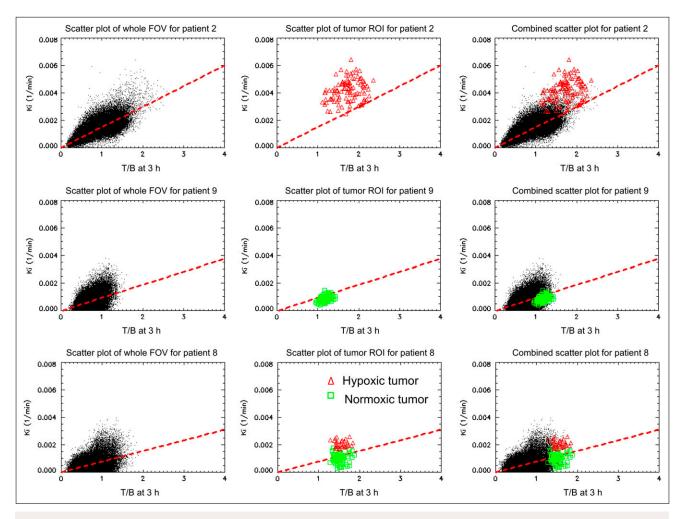


**FIGURE 5.** Ambiguity in concordance between 2 surrogate measures of tumor hypoxia  $K_i$  and T/B from late-time-point imaging for patient 8. Transverse, coronal, and sagittal views are shown for <sup>18</sup>F-FDG PET, <sup>18</sup>F-FMISO PET at 3 h, CT fused with tumor ROI,  $k_1$ ,  $K_i$ , and  $\beta$  (A) and measured and fitted average activity concentration in tumor sub-ROI as function of time (B).

images segments from 0 to 30, 90 to 100, and 180 to 190 min. This degree of registration accuracy can be achieved when patients are immobilized in a head-and-shoulder immobilization mask (24) but would incur larger errors for tumors at other sites (except the brain) of the body. A simulation study to investigate the effect of different errors on the resultant parameter estimation, accuracy, and precision on voxel-based compartmental analysis has been published by our group (18).

In terms of tumor ROI, there is only a slight negative correlation between surrogate perfusion rate  $k_I$  and tumor

hypoxia measure  $K_i$  (-0.33), a moderate positive correlation of vascular volume fraction  $\beta$  and  $K_i$  (0.54), but no correlation between  $\beta$  and  $k_I$  (-0.08). The slight negative correlation between perfusion and tumor hypoxia represented by the surrogate parameters  $k_I$  and  $K_i$ , respectively, is consistent with findings in studies by Gagel et al. (11) and Thorwarth et al. (17). However, these correlation studies need to be repeated on a larger group of patients to have any statistical significance. We also need to correlate the hypoxia, perfusion, and vascular volume fraction with treatment outcome with a larger population of patients.



**FIGURE 6.** Scatter plots of individual voxels (T/B,  $K_i$ ) at both whole head and neck region and tumor ROI for patients 2 (top row), 9 (middle row), and 8 (bottom row). Whole head and neck region is shown on left, tumor ROI in center, and both regions on right.  $\bullet$  = whole head and neck region; red triangles = putative hypoxic tumor region; green squares = putative normoxic tumor region.

The analysis performed in our study is from a pilot study consisting of 9 head and neck patients. Although small, the results of our study concur with the findings reported by Thorwarth et al. (17) that a single T/B threshold in the latetime <sup>18</sup>F-FMISO image may not be adequate and sufficient to represent the spatial distribution of hypoxia within tumors. This may be inconvenient for those seeking a simple PET procedure of inject, wait, and scan with which to identify tumor hypoxia. The clearance of <sup>18</sup>F-FMISO from highly perfused tissue may result in T/B values comparable to the low-level trapping of <sup>18</sup>F-FMISO in hypoxic regions at the 1.5- to 3-h time of PET after injection. Later imaging times can result in significant image noise and variation in voxelized T/B. The consequence of this is that voxel-based subregional analysis of tumors may exhibit high T/B but with low  $K_{i}$  leading to the ambiguity between the 2 potential surrogate PET measures of tumor hypoxia observed in this study. Potential errors, resulting from the misinterpretation of such regions, can be overcome if a composite criterion,

which includes the additional use of a parametric hypoxia measure such as  $(T/B, K_i)$ , is used to identify regional tumor hypoxia. Most hypoxic tumor voxels, when defined by this dual metric, are likely to locate at the sparse region of the  $(T/B, K_i)$  scatter plot, separated from the normal tissue cluster region, at a high  $K_i$  value and above the linear regression line. This  $(T/B, K_i)$  clustering technique might be useful to define biologic target volume in radiotherapy for functional image-based dose escalation.

The focus of this study was to investigate the potential of  $k_3$ ,  $K_i$ , and late-time-point T/B as voxelized image maps as surrogate measures for the intratumor distribution of hypoxia. There were also other potential surrogate measures of tumor hypoxia, such as  $k_1/k_2$ , that were calculated (Table 1) but not further discussed in this article.

# CONCLUSION

Different surrogate measures of intratumor hypoxia may be good concordance for a fraction of patients, but

ambiguities can arise, in particular between the measures obtained by compartmental analysis ( $k_3$  and  $K_i$ ) and simple T/B from late-time-point imaging. The parametric image maps derived in this work are not at this point proven surrogate measures of hypoxia but rather are a fit of the trapping rate constants, based on a compartmental model of <sup>18</sup>F-FMISO binding to a hypoxia compartment within the tumor tissue. The results of our analysis need to be validated against independent and more direct techniques such as image-guided partial oxygen pressure (pO<sub>2</sub>) probe measurement or immunohistochemisty. This pilot study did not include clinical pO<sub>2</sub> measurements or tumor biopsy. It is part of our future plans to conduct such validation measurements in experiments performed in an animal tumor model.

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