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FDG uptake heterogeneity evaluated by fractal analysis improves the differential diagnosis of pulmonary nodules

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Purpose: The present study aimed to determine whether fractal analysis of morphological complexity and intratumoral heterogeneity of FDG uptake can help to differentiate malignant from benign pulmonary nodules.

Materials and methods: We retrospectively analyzed data from 54 patients with suspected non-small cell lung cancer (NSCLC) who were examined by FDG PET/CT. Pathological assessments of biopsy specimens confirmed 35 and 19 nodules as NSCLC and inflammatory lesions, respectively. The morphological fractal dimension (m-FD), maximum standardized uptake value (SUVmax), and density fractal dimension (d-FD) of target nodules were calculated from CT and PET images. Fractal dimension is a quantitative index of morphological complexity and tracer uptake heterogeneity; higher values indicate increased complexity and heterogeneity.

Results: The m-FD, SUVmax and d-FD significantly differed between malignant and benign pulmonary nodules (p < 0.05). Although the diagnostic ability was better for m-FD than SUVmax or d-FD, the difference did not reach statistical significance. Tumor size correlated significantly with SUVmax (r = 0.51, p < 0.05), but not with either m-FD or d-FD. Furthermore, m-FD combined with either SUVmax or d-FD improved diagnostic accuracy to 92.6% and 94.4%, respectively.

Conclusion: The d-FD of intratumoral heterogeneity of FDG uptake can help to differentially diagnose malignant and benign pulmonary nodules. The SUVmax and d-FD obtained from FDG-PET images provide different types of information that are equally useful for differential diagnoses. Furthermore, the morphological complexity determined by CT combined with heterogeneous FDG uptake determined by PET improved diagnostic accuracy.

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1 Introduction

Pulmonary nodules have recently been differentiated using F-18 fluorodeoxyglucose (FDG) positron emission tomography (PET) [1]. The maximum standardized uptake value (SUVmax) is a popular clinical method of evaluating the degree of FDG uptake in tumors. The SUVmax reflects the highest amount of FDG uptake, but not its distribution throughout tumors. The uptake of FDG is heterogeneous in some types of tumor [2]. Factors that might contribute to heterogeneous intratumoral FDG uptake such as necrosis, cellular proliferative activity, blood flow, microvessel density and hypoxia are also regarded as features of tumors [3]. Thus, characterization based on FDG uptake heterogeneity might help to distinguish benign from malignant pulmonary nodules.
Fractal analysis is one method of quantifying tumor complexity and heterogeneity on various types of images [4–8]. Nagao et al. developed a method of quantifying the heterogeneity of nuclear medicine images using density fractal dimension and found it useful for evaluating cerebral blood flow distribution on SPECT [9] and the heterogeneity of 99mTc-technegas distribution in the lung [10]. Here, we applied fractal analysis to FDG-PET images to determine the fractal dimension of pulmonary tumor heterogeneity. To our knowledge, this is the first effort to determine whether fractal analysis applied to FDG-PET/CT can differentiate malignant and benign pulmonary nodules.

2. Materials and methods

2.1. Patients

We retrospectively analyzed FDG PET/CT data from 54 patients (37 men, and 17 women; mean age, 70 ± 9 years; age range, 31–88 y) with suspected non-small cell lung cancer (NSCLC). Thirty-five and 19 biopsy specimens of nodules were confirmed as NSCLC and inflammatory lesions, respectively (Table 1). This clinical study was approved by the ethics committee of our institution (no. 25–54). This study was retrospective, and its results did not influence further therapeutic decision-making.

2.2. FDG PET/CT

Patients fasted for at least 6 h before being injected with 4 MBq/kg FDG and then whole body image acquisition started at a mean of 60 min later from the top of the skull to the mid-thigh using an Aquiduo PET/CT scanner (Toshiba, Japan). Emission data were acquired for 2–3 min per bed position. The PET images were reconstructed using an iterative algorithm (attenuation-weighted ordered-subsets expectation maximization: 4 iterations, 14 subsets) with an 8-mm Gaussian filter, a 128 × 128 matrix (3.9 mm/pixel) and 81 slices (2 mm/slice). Whole-body CT scanning proceeded under the following parameters: 120 kV; auto exposure control system (noise level: SD 10); 512 × 512 matrix; beam pitch, 0.94; 2 mm × 16-row mode.

2.3. Fractal analysis

Fractal geometry is characterized by the relationship between a measure (M) and a scale (ε), expressed as:

\[ M(ε) = k \cdot ε^{−D}, \]

where k is a scaling constant and D is the fractal dimension that is used to detect self-affinity. The morphological fractal dimension (m-FD) is a quantitative index of morphological complexity derived from CT on PET/CT, with higher values corresponding to increasing degrees of complexity. The heterogeneity of FDG distribution is expressed as the density fractal dimension (d-FD), of which higher values correspond to increasing degrees of heterogeneity. In d-FD, the chosen cut-offs were used as the ruler scale ε in the above equation.

The number of voxels containing with radioactivity higher than the corresponding cut-offs are expressed as M(ε), in which M decreases as ε increases and the magnitude of the slope of linear regression between the logarithms of the cut-offs and the numbers of pixels is equal to the fractal dimension.

2.4. Quantitative analysis

We evaluated transaxial PET images of maximal cross-sectional diameters of nodules. A circular region of interest (ROI) with the minimal diameter required to cover the entire nodule was created. The highest pixel value was determined as the SUVmax of the nodule. To calculate the d-FD, the cut-offs for maximal image intensity in the nodule were set at 30 levels ranging from 40% to 100% at intervals of 2% and then the number of pixels exceeding the cut-off was calculated. The m-FD was calculated from the extracted boundary of the lesion derived from CT images using the box-counting method [11].

2.5. Statistical analysis

We compared m-FD, SUVmax and d-FD between malignant and benign pulmonary nodules using an unpaired t test. Relationships between tumor size and m-FD, SUVmax and d-FD were evaluated using Pearson’s correlation coefficients. Values plotted nearest the upper left corner from receiver operating characteristics analyses (ROC) were considered to indicate the best diagnostic accuracy. Sensitivity, specificity and accuracy were calculated using appropriate cut-offs. Diagnostic accuracy was also compared using areas under ROC curves (AUC) and ROC curves were compared using critical z testing. All data were statistically analyzed using software (SPSS Inc., USA) and p < 0.05 was considered to indicate statistical significance.

3. Results

Fig. 1 compares the m-FD, SUVmax and d-FD between malignant and benign pulmonary nodules. The m-FD and d-FD were significantly lower and the SUVmax was significantly higher in malignant, than in benign nodules (all p < 0.05). Fig. 2 shows relationships between tumor size and m-FD, SUVmax and d-FD. Tumor size significantly correlated with SUVmax (p < 0.05), but not with either m-FD or d-FD.

The optimal cut-off values for differentiating malignant from benign pulmonary nodules were 1.183, 4.24 and 0.0267 for m-FD, SUVmax and d-FD, respectively. The diagnostic accuracy of SUVmax (68.5%) and d-FD (77.8%) on PET was better that that of m-FD on CT (64.8%). The diagnostic accuracy tended to be higher for d-FD than for SUVmax, but the difference did not reach statistical significance. Fig. 3 shows the outcomes of the ROC analyses for each index. The AUCs of m-FD, SUVmax and d-FD were 0.597, 0.79 and 0.827, respectively, and did not significantly differ (Table 2).
Fig. 1. Comparison of values for m-FD, SUVmax and d-FD between malignant and benign pulmonary nodules. Values for m-FD and d-FD are significantly lower, whereas that of SUVmax is significantly higher in malignant, than benign nodules (all \( p < 0.05 \)). B, benign; M, malignant.

Table 2

<table>
<thead>
<tr>
<th></th>
<th>CT m-FD</th>
<th>PET SUVmax</th>
<th>PET d-FD</th>
<th>CT + PET m-FD + SUVmax</th>
<th>CT + PET m-FD + d-FD</th>
</tr>
</thead>
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<td>AUC</td>
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</tr>
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<td>Sensitivity</td>
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<td>94.2%</td>
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<td>72.7%</td>
<td>73.7%</td>
<td>89.5%</td>
<td>94.2%</td>
</tr>
<tr>
<td>Accuracy</td>
<td>62.9%</td>
<td>68.5%</td>
<td>77.8%</td>
<td>92.6%</td>
<td>94.4%</td>
</tr>
</tbody>
</table>

4. Discussion

Fractal analysis evaluates the spatial pattern of irregular objects so that the morphological complexity and spatial heterogeneity can
be quantified and assigned numerical values [12,13]. Intratumoral heterogeneity in FDG-PET images has been evaluated using textural analysis [14–16], the coefficient of variance (COV) [2], cumulative SUV-volume histograms (CSH) [17] and the area under the CSH (AUC–CSH) [2,18]. The advantage of fractal analysis is the accurate quantitation of highly sensitive, reproducible values that can be used to evaluate the specific characteristics of tumors [7,19]. Repeated fractal analysis of the same image can theoretically result in assigning an agreement value [7].

Although the m-FD, SUV_{max} and d-FD could differentiate between malignant and benign, the diagnostic ability was higher for both SUV_{max} and d-FD than for m-FD. Metabolic information obtained from FDG-PET imaging is generally superior to morphological information obtained from CT for differential diagnoses of pulmonary nodules [20]. We found that the diagnostic ability of d-FD was better than that of the SUV_{max}, but not significantly. The SUV_{max} is considered to reflect the most biologically aggressive area, but this value is influenced by the spatial resolution of the images [21]. We identified a positive correlation between tumor size and the SUV_{max}. On the other hand, d-FD did not correlate with tumor size. These results indicate that the SUV_{max} and d-FD obtained from FDG-PET images provide different information and are equally useful for differential diagnosis [22].

The d-FD was lower for malignant, than benign pulmonary nodules. Yu et al. performed a textural analysis of head and neck cancer and found lower heterogeneity in tumors and nodes than in normal tissue [23]. Higher FDG uptake heterogeneity in benign pulmonary nodules is thought to reflect metabolically diverse components including inflammatory tissues and surrounding normal tissues such as vessels, bronchi and pleura. In the same context, van Velden et al. [18] proposed a quantitative index of intratumor FDG uptake heterogeneity with which to evaluate NSCLC responses to treatment, whereas Tixier et al. [14] showed that the intratumoral heterogeneity of FDG uptake could predict responses of esophageal cancer to radio-chemotherapy.

The usefulness of the combination morphological complexity and heterogeneity of FDG uptake has not been reported. The combination of m-FD and d-FD significantly improved diagnostic accuracy associated with a significant reduction in false-positive findings. Yu et al. also reported that co-registered FDG-PET/CT-based analysis was useful for the differential diagnosis of head and neck cancer [23]. These findings suggest a complementary relationship between morphological complexity and the heterogeneity of FDG uptake. Thus, the combination of m-FD and d-FD might improve the accuracy of diagnosing pulmonary nodules.

This study had some limitations. The FD may be influenced by various histological types of tumors and inflammatory conditions. Further studies of more patients with various types of tumors are required. Image noise and a partial volume effect due to the limited spatial resolution of PET might have influenced the results [18]. Therefore, the FD should be calculated from PET/CT images acquired under identical conditions using the same equipment.

5. Conclusions

The intratumoral heterogeneity of FDG uptake evaluated by fractal analysis was useful for discriminating benign, from malignant pulmonary nodules. The SUV_{max} and intratumoral heterogeneity determined from PET images provided different types of information that are equally useful for differential diagnosis. Furthermore, combining the morphological complexity of tumors on CT images with the heterogeneity of FDG uptake on PET images improved diagnostic accuracy.

Conflicts of interest

The authors declare that they have no conflict of interest.

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References


