Identify the radiotherapy-induced abnormal changes in the patients with nasopharyngeal carcinoma

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Abstract

Radiotherapy (RT) is the standard treatment for nasopharyngeal carcinoma, which often causes inevitable brain injury in the process of treatment. The majority of patients has no abnormal signal or density change of the conventional magnetic resonance imaging (MRI) and computed tomography (CT) examination in the long-term follow-up after radiation therapy. However, when there is a visible CT and conventional MR imaging changes, the damage often has been severe and lack of effective treatments, seriously influencing the prognosis of patients. Therefore, the present study aimed to investigate the abnormal changes in nasopharyngeal carcinoma (NPC) patients after RT. In the present study, we exploited the machine learning framework which contained two parts: feature extraction and classification to automatically detect the brain injury. Our results showed that the method could effectively identify the abnormal regions reduced by radiotherapy. The highest classification accuracy was 82.5 % in the abnormal brain regions. The parahippocampal gyrus was the highest accuracy region, which suggested that the parahippocampal gyrus could be most sensitive to radiotherapy and involved in the pathogenesis of radiotherapy-induced brain injury in NPC patients.

Keywords

nasopharyngeal carcinoma (NPC); radiotherapy (RT); injury; classification; structural MRI (sMRI);

Introduction

Nasopharyngeal carcinoma (NPC) is an endemic disease and rare in most regions of the world. However, it exists much more frequently in Southeast Asia and China. In Southern China, NPC is one of the most common malignant tumor and the incidence of NPC is approximately 30–80 per 100,000 per year [1]. Because of the special lesion location, nasopharyngeal carcinoma is not suitable for surgical treatment and radiotherapy plays a crucial part in NPC therapy. However, brain injury caused by radiation therapy is a serious complication, which has a serious impact on the prognosis and quality of life of the patients. Thus, it is necessary to understand the effect of radiation therapy on brain structure and find neural biomarkers to facilitate clinical diagnose, treatment, and prevention [2].

Previous studies have found that radiation therapy for NPC resulted in multiple regions abnormalities,
such as temporal lobe necrosis [3, 4], precuneus, cuneus, lateral occipital cortex [5], the vermis, hippocampus, cerebellum lobule, middle occipital lobe, and insula [6, 7]. However, it was difficult for doctors to identify the abnormal changes after radiotherapy (RT) based on original image features. There was a lack of automatic classification method for NPC patients before and after RT based on the big data of sMRI. Therefore, we hoped that the machine learning framework with sMRI could improve the present situation on brain radiation therapy.

Materials and Methods

Participants

The total participants included 103 right-handed NPC patients (80 males and 23 females, mean ±SD= 47.5 ± 10.8 years, age from 23 to 82 years). All the patients were recruited from the Sichuan Cancer Hospital in Chengdu, China, and they all had no other disease. Each subject performed two MRI scans, i.e. before (pre-RT) and after radiotherapy (post-RT).

The research was approved by the Ethics committee of Sichuan Cancer Hospital & Institute, Chengdu, China. Written informed consent was obtained from all subjects.

MRI acquisition

The sMRI images of the whole brain were obtained by the rapid acquisition of T1 weighted sequences with a high-resolution three-dimensional magnetic Spin echo using a Siemens 1.5 MR scanner (TR = 750 ms, TE = 11 ms, FOV = 75x75 cm, flip angle = 150°, matrix size = 256x192, voxel size = 1x1x1 mm³, slice thickness = 3 mm, layer space = 3.6 mm, the continuous acquisition = 36 layers covering the whole brain.

The voxel-based morphometry (VBM) toolbox of the Statistical Parametric Mapping software (SPM8) was used to perform the data-preprocessing. Firstly, the structured images were registered into the standard MNI space according to 12 radiation parameters. Secondly, the spatially normalized images were segmented into gray matter, white matter and cerebrospinal fluid using a priori tissue probability maps (TPM) provided by International Consortium for Brain Mapping (ICBM). Thirdly, the segmented gray images were further smoothed with a kernel of 8 mm to remove the influence of noise and increase the validity of statistical tests of the posterior parameters.

Methods

The methods mainly included two parts: 1) feature extraction and feature selection with the principal component analysis (PCA); 2) classification with support vector machine (SVM). Details were shown in Figure1. A 3×3×3 cube was defined and vectorized. Leave-one-out cross-validation (LOOCV) was used to evaluate the general classification performance. Image data were divided into two sessions: one was a test session whose size were 1×27 was leaving out by turns in each round of LOOCV, and the other was a training session whose size were 205×27. Then, PCA was used to extract the features of the training session. LIBLINEAR was an open source library for large-scale linear classification that inherits many features of the popular SVM library LIBSVM. In this paper, we used the LIBLINEAR toolbox [8] to perform the classification task.
Results and Discussion

As shown in Figure 2, the final accuracy for distinguishing the pre-RT and post-RT was 82.5 %. The interested brain regions should be met: 1) the clusters must contain more than 50 continuous voxels. 2) The accuracy of all the voxels in the cluster must be greater than 70 %. The final classification accuracy was the greatest accuracy of the voxels in all the interested brain clusters, and the regions in the interested brain clusters were the most distinguishable brain regions. In this study, we identified some interested brain regions which mainly included the bilateral parahippocamp, the bilateral middle temporal gyrus, the right middle occipital gyrus, the left rolandic operculum, the bilateral cerebellum, the vermis, the right hippocampus, the right thalamus, the insula, the right precuneus, and the left cuneus. Details were shown in the Table 1.

In this study, we used the machine learning framework based on feature extraction and classification to achieve the automatic classification between the pre-RT and post-RT. The method of PCA+SVM was quite helpful to improve the accuracy of the classification, and validly detected the abnormal brain regions, which widely contained the previous research achievement. As shown in Table 1, our results were consistent with previous studies, suggesting that the method was in favour of identifying the injury caused by RT [5-7, 9]. The classification accuracy was 82.5 %, indicating that the method could well identify the abnormal changes induced by RT. The abnormal brain regions can be considered as biomarkers to detect the radiation-induced injury and help doctors diagnose and monitor the disease.

Some researchers revealed that the parahippocampal cortex and hippocampus were involved in memory [10-13]. The parahippocampal cortex played a role for spatial configuration of object in memory [14]. Hippocampus was associated with attention processes and visuospatial working memory [15]. The temporal lobe necrosis was the most debilitating late-stage complication [16]. Cui et al [17] found that the temporal lobe played critical roles in auditory and language processing, which might be associated with auditory verbal hallucinations (AVHs) for schizophrenia patients. Temporal regions might also be
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**Figure 2.** The classification accuracies of NPC patients before RT and after RT based on the PCA and SVM.

**Table 1.** The most distinguish force of gray matter volume brain regions for the pre-RT and post-RT groups

<table>
<thead>
<tr>
<th>Clusters</th>
<th>Voxels</th>
<th>x</th>
<th>y</th>
<th>z</th>
<th>Accuracy of the peak</th>
</tr>
</thead>
<tbody>
<tr>
<td>ParaHippocampal_L</td>
<td>575</td>
<td>-16.5</td>
<td>-19.5</td>
<td>-24</td>
<td>82.5 %</td>
</tr>
<tr>
<td>Temporal_Mid_R</td>
<td>186</td>
<td>48</td>
<td>-54</td>
<td>13.5</td>
<td>80 %</td>
</tr>
<tr>
<td>Occipital_Mid_R</td>
<td>996</td>
<td>28.5</td>
<td>-78</td>
<td>10.5</td>
<td>80 %</td>
</tr>
<tr>
<td>Rolandic_Oper_L</td>
<td>496</td>
<td>-57</td>
<td>-6</td>
<td>9</td>
<td>77.5 %</td>
</tr>
<tr>
<td>Cerebellum_9_L</td>
<td>177</td>
<td>-12</td>
<td>-45</td>
<td>-51</td>
<td>77.5 %</td>
</tr>
<tr>
<td>Cerebellum_9_R</td>
<td>129</td>
<td>7.5</td>
<td>-51</td>
<td>-40.5</td>
<td>75 %</td>
</tr>
<tr>
<td>Vermis_1_2</td>
<td>716</td>
<td>1.5</td>
<td>-40.5</td>
<td>-22.5</td>
<td>80 %</td>
</tr>
<tr>
<td>Hippocampus_R</td>
<td>183</td>
<td>19.5</td>
<td>-10.5</td>
<td>-9</td>
<td>80 %</td>
</tr>
<tr>
<td>ParaHippocampal_R</td>
<td>60</td>
<td>18</td>
<td>-24</td>
<td>-15</td>
<td>75 %</td>
</tr>
<tr>
<td>Thalamus_R</td>
<td>137</td>
<td>4.5</td>
<td>-13.5</td>
<td>4.5</td>
<td>75 %</td>
</tr>
<tr>
<td>Temporal_Mid_L</td>
<td>480</td>
<td>-45</td>
<td>-70.5</td>
<td>18</td>
<td>77.5 %</td>
</tr>
<tr>
<td>Insula_R</td>
<td>723</td>
<td>30</td>
<td>-28.5</td>
<td>12</td>
<td>77.5 %</td>
</tr>
<tr>
<td>Precuneus_R</td>
<td>495</td>
<td>9</td>
<td>-51</td>
<td>13.5</td>
<td>77.5 %</td>
</tr>
<tr>
<td>Cuneus_L</td>
<td>61</td>
<td>-13.5</td>
<td>-57</td>
<td>18</td>
<td>75 %</td>
</tr>
</tbody>
</table>

Ps. ParaHippocampal_L: left parahippocampal gyrus; Temporal_Mid_R: right middle temporal gyrus; Occipital_Mid_R: right middle occipital gyrus; Rolandic_Oper_L: left Rolandic operculum; Cerebellum_9_L: left cerebellum; Cerebellum_9_R: right cerebellum; Hippocampus_R: right hippocampus gyrus; ParaHippocampal_R: right paraHippocampal gyrus; Thalamus_R: right thalam; Temporal_Mid_L: left middle temporal gyrus; Insula_R: right insula; Precuneus_R: right precuneus; Cuneus_L: left cuneus.

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involved in the Human understanding insight which occupied an important position in understanding the clinical heterogeneity of obsessive-compulsive disorder (OCD) [18]. Glasser and Rilling [19] found that the temporal lobe, especially the left superior and middle temporal gyri, was a pivotal part of language pathway in the brain’s language network.

The cerebellum was vital to the motor control, cognitive and affective regulation [7]. Shen et al. [20] reported that the cerebral microbleeds (CMBs) observed in the cerebellum were related to cognitive dysfunction.

The insula participated in numerous cognitive processing, including goal-directed cognition, conscious awareness, autonomic regulation, interoception and somatosensation [21].

The precuneus was found to be connected with a wide range of cognitive processes, involving reflective and self-related processing [22, 23], awareness and conscious information processing [24, 25], episodic memory [26-28], and visuospatial processing [29, 30].

The cuneus was significantly involved with visual processing [30, 31]. Collignon et al. [32] demonstrated that auditory-spatial processing mainly recruited the right cuneus and the right middle occipital gyrus, and the specific occipital regions in visuospatial / motion processing for sighted individuals.

The thalamus was believed to regulate and coordinate cortical activity both within and across functional regions, such as motor and visual cortices [33].

Wu et al. [34] found that the NPC patients after radiotherapy demonstrated significant changes in cognition which mainly referred to short term memory, delayed recall, language, attention, orientation, visuo-spatial and executive function. In addition, the patients showed a relatively high rate of general intelligence impairment, which might be related to a longer post-RT interval of mean 4.3 years and bigger total dosage of mean 70.7Gy. Tang et al [35] reported that NPC patients with radiation-induced brain injury (RI) exhibited negative emotions, impaired cognitive function and quality of life (QOL). The accumulated studies showed that the radiation-induced functional impairments included the disorder of short-term memory [36], personality changes [37] and motor abilities [38], a marked anterograde memory impairment for verbal material [39], neuropsychological impairments in recent memory, immediate and delayed verbal recall, and immediate visual recall [40].

Hence, we inferred that these abnormal brain regions might be correlated with the dysfunction of patients with NPC after radiotherapy, which might reveal a potential mechanism for radiation-induced impairment. Furthermore, it was clear that the machine learning framework proposed in the current study could provide an effective reference for identifying the brain injury induced by radiotherapy in clinical diagnosis.

Conclusions

In summary, this present study proposed an automatic classification method for pre-RT and post-RT. The experimental results showed that the method could effectively identify the abnormal changes between pre-RT and post-RT. Furthermore, these changes might be considered as an underlying biomarker to detect the radiation-induced impairment.
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References


