

Predicting Treatment Response in Patients with Hepatocellular Carcinoma Treated with Y90 Radioembolization Using Deep Learning

William Wagstaff, MD, MS, Emory University

Introduction

Because only a minority of patients with hepatocellular carcinoma (HCC) meet criteria for curative surgical/transplant therapy, Yttrium-90 radioembolization (Y90-RE) has become an alternative, established therapy. During Y90-RE, radioactive particles are injected into the tumor-feeding vessels, increasing patient survival when cytotoxic doses are achieved. Response is often determined on follow-up imaging at 3-6 months, during which, the tumor can progress or metastasize if not adequately treated. This study aims to predict which patients will have complete treatment response using a deep convolutional neural network (DNN), and in doing so, highlight patients that may benefit from earlier follow-up or additional therapy.

Hypothesis

We hypothesize that a DNN can accurately classify Y90-RE treatments in patients with HCC as complete response/dead tumor or incomplete response/residual disease.

Methods

After IRB approval, a retrospective analysis was conducted in patients with HCC who received Y90-RE between 12/2014 and 1/2019 at a single institution. 77 patients with 103 lesions met the inclusion criteria: three or fewer tumors, pre- and post-treatment MRI, and no prior Y90-RE. Lesions were divided into complete (n=57) or incomplete response (n=46), based on reports from 3-month post-treatment MRI. Lesions were hand-segmented on pre-treatment arterial phase MRI to determine lesion location and on post-treatment arterial phase MRI to determine slice-wise response. Immediate post-Y90-RE Bremsstrahlung SPECT was used as a proxy for treatment area and dose. Lesions were divided by MRN into a 90% training set for 5-fold cross-validation, and a 10% hold-out test set. The final test model was an average ensemble of the models and thresholds from the best epochs across all 5 folds (Figure 1). Slice-wise classification results were compared against the current standard of care for post-Y90-RE analysis, partition modeling, using sensitivity, specificity, accuracy, and area under the receiver operating characteristic curve (ROC-AUC).

Results

Five-fold cross-validation results for the ensemble model were sensitivity 0.73±0.29, specificity 0.70±0.14, accuracy 0.75±0.09, and ROC-AUC 0.79±0.12; and for the partition model were sensitivity 0.36±0.34, specificity 0.80±0.40, accuracy 0.63±0.04, and ROC-AUC 0.62±0.09. Test set results for the ensemble model were sensitivity 0.32, specificity 0.85, accuracy 0.65, and ROC-AUC 0.66; results for the partition model were sensitivity 0.03, specificity 0.99, accuracy 0.62, and ROC-AUC 0.57 (Figure 2).

Conclusion

The deep learning algorithm had a 16% higher ROC-AUC for predicting immediate, postradioembolization treatment response in patients with HCC compared to the current best method. This algorithm has the potential to highlight patients that may benefit from earlier follow-up or additional therapy.

Figure(s)



Figure 1. The top row demonstrates arterial phase imaging (first), segmentation of the HCC (second), bremsstrahlung SPECT overlayed on the arterial phase imaging (third), and post-treatment segmentation of the residual enhancing component of the HCC (fourth). The bottom row demonstrates the input to the neural network (timm-skresnext50_32x4d; input size = $192 \times 192 \times 6$ with 3 grouped lesion maps and 3 grouped bremsstrahlung maps, batch size=16, gradient accumulation=4, dropout=0.1, epochs=3), with the output as a value between 0 and 1 converted to 1 or 0 based on a validation threshold.



Figure 2. Receiver operating characteristic curve for predicting slice-wise treatment response after Y90 radioembolization therapy in patients with hepatocellular carcinoma.

Keywords

Applications; Artificial Intelligence