

Predicting Response to Bevacizumab in Colorectal Cancer by Integrating Radiomics to Clinical and Genomic Features

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Background

In our retrospective analysis, Notch signalling pathway was associated with resistance to anti-vascular endothelial growth factor (VEGF) therapy in patients with metastatic colorectal cancer (mCRC).

We tested whether radiomics might select treatment-naïve mCRC patients responding to Bevacizumab, beyond clinical and genomic (Notch Intracellular Cleaved Domain (NICD)/JAG1 expression) parameters.

Methods

consecutive mCRC patients treated with first-line bevacizumab were retrospectively selected.

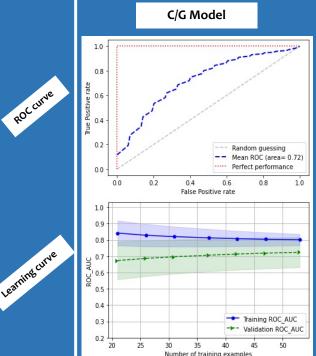
Immunohistochemistry analysis of tissue microarrays assessed NICD, JAG1, CD44, CD3, CD4, CD8, CD20, DLL3 and DLL4 expression. Abdominal CT scans were imported into a dedicated software for tumor segmentation and extraction of 852 radiomic features (RFs), which were included into machine learning-based predictive models. Pre-processing of RFs included redundant features elimination and standardization; L2 penalized logistic regression with Monte-Carlo cross-validation were implemented for wrapper-based feature selection and model training/test. Three models were developed: clinical/genomic (C/G), radiomic (R) and the comprehensive integrated model (I), which were

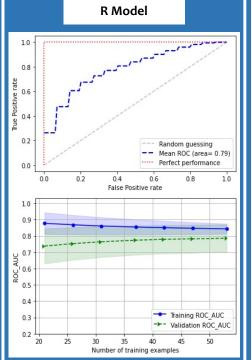
Results

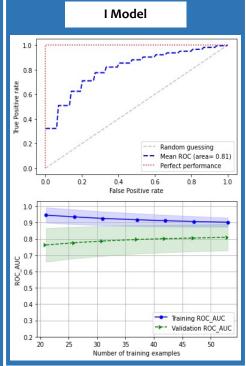
NICD and JAG1 expression was associated with response to Bevacizumab (p<0.05). Using likelihood-ratio test as inclusion criteria, the selected variables were 5 for both C/G and R models, then aggregated into the I model. C/G features included NICD expression, number of involved sites, primitive location, resection of metastases and performance status, while selected RFs belonged to both first- and higher-orders classes. ROC-AUC and accuracy were 0.724 (95%Cl:0.722-0.727) and 0.669 (95%Cl:0.666-0.671), 0.786 (95%Cl:0.784-0.788) and 0.710 (95%Cl:0.708-0.713), o.810 (95%CI:o.808-o.812) and o.743 (95%CI:o.741-o.745) for C/G, R and I model, respectively.

Conclusions

The integration of clinical, genomic and radiomic features showed the highest performance in predicting response to Bevacizumab.







compared based on ROC-AUC and accuracy metrics.

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