

# Prediction of Response to Immunotherapy in Cancers via Transfer Learning

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## Abstract

To date, immunotherapies such as immune checkpoint inhibitors (ICIs) have emerged as a leading treatment for metastatic cancer, significantly improving patient survival while causing relatively few side effects. However, the objective response rate for ICIs remains low, approximately 25% for metastatic urothelial carcinoma (mUC) and renal cell carcinoma (mRCC), and ~40% for melanoma, underscoring the urgent need for predictive response biomarkers. Several state-of-the-art signatures have been revealed in top-tier journals, highlighting the importance of this field. As the number of genes (~20,000) far exceeds the sample sizes of typical training sets (generally  $\leq 300$ ), we first developed feature selection procedures to reduce the number of features to a few hundred. We then trained multiple machine learning classifiers using the selected genes and the IMvigor210, IMmotion150, and Gide cohorts, which includes RNA-seq and clinical data from 298, 77, and 91 patients with mUC, mRCC, and melanoma, respectively. Notably, our predictor LogitDA using the identified gene signatures achieved a prediction AUC of 0.75, 0.83, and 0.71~0.75 in independent cohorts, PCD4989g (mUC, mRCC), and three melanoma cohorts, respectively. Moreover, our signatures outperformed (most of) six state-of-the-art signatures, PD-L1 IHC, and five tumor microenvironment signatures, including IFN- $\gamma$ , T-effector, and T-cell exhaustion signatures in mUC (mRCC and melanoma). From our signatures, we identified key prognostic biomarkers in mUC, mRCC, and melanoma, respectively.

Keywords: biomarker, cancer, machine learning, regression, prediction