

In silico modeling of combination systemic therapy for advanced renal cell carcinoma

Ritesh R Kotecha ^(b), ¹ Dennis J Hsu ^(b), ^{1,2} Chung-Han Lee, ¹ Sujata Patil, ³ Martin H Voss ^(b)

ABSTRACT

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¹Department of Medicine, Memorial Sloan Kettering Cancer Center, New York, New York, USA ²School of Medicine, University of Pittsburgh, Pittsburgh, PA, USA

³Department of Epidemiology and Biostatistics, Memorial Sloan Kettering Cancer Center, New York, New York, USA

Correspondence to Dr Martin H Voss; vossm@mskcc.org

Therapeutic combinations of VEGFR tyrosine kinase inhibitor plus immune checkpoint blockade now represent a standard in the first-line management of patients with advanced renal cell carcinoma. Tumor molecular profiling has shown notable heterogeneity when it comes to activation states of relevant pathways, and it is not clear that concurrent pursuit of two mechanisms of action is needed in all patients. Here, we applied an in silico drug model to simulate combination therapy by integrating previously reported findings from individual monotherapy studies. Clinical data was collected from prospective clinical trials of axitinib, cabozantinib, pembrolizumab and nivolumab. Efficacy of two-drug combination regimens (cabozantinib plus nivolumab, and axitinib plus pembrolizumab) was then modeled assuming independent effects of each partner. Reduction in target lesions, objective response rates (ORR), and progression-free survival (PFS) were projected based on previously reported activity of each agent, randomly pairing efficacy data from two source trials for individual patients and including only the superior effect of each pair in the model. In silico results were then contextualized to register phase III studies of these combinations with similar ORR, PFS, and best tumor response. As increasingly complex therapeutic strategies emerge, computational tools like this could help define benchmarks for trial designs and precision medicine efforts. Summary statement: In silico drug modeling provides meaningful insights into the effects of combination immunotherapy for patients with advanced kidney cancer.

INTRODUCTION

Combination vascular endothelial growth factor receptor (VEGFR) tyrosine kinase inhibitor (TKI) plus immuno-oncology (IO) therapy constitutes a first-line standard for metastatic renal cell carcinoma (mRCC), with survival benefit over TKI monotherapy.^{1 2} IO monotherapy similarly achieves less striking efficacy compared with combinations. Transcriptomic and immunogenomic data highlight interpatient tumor heterogeneity and specific activation states of targeted molecular pathways pertinent for individual drug activity.³ To evaluate how these drugs combine, we performed in silico modeling⁴ integrating monotherapy clinical data to simulate TKI/IO combinations.

METHODS

Objective response rates (ORR) and progression-free survival (PFS) data were collected from seven individual mRCC prospective studies conducted in TKI-naïve and immune checkpoint inhibitor naïve patients: monotherapy trials of axitinib, cabozantinib, nivolumab, and pembrolizumab; and, registration studies of axitinib plus pembrolizumab and cabozantinib plus nivolumab.¹²⁵⁻⁹ Monotherapy data were digitally measured using Plot Digitizer (V.2.6.9) and a custom Python script was written to extrapolate probability density function from each PFS curve. Each in silico patient was randomly assigned a PFS, response (yes/no), and maximal tumor reduction for each of the two drugs using published monotherapy data. Combination therapy was modeled under the assumption of independent treatment effects, hypothesizing that solely the more active of the two agents was determining the extent of benefit. Thus, for ORR, PFS and best response, each in silico patient was assigned the best values of the two individual therapies. Each trial was run 1000 times, unless otherwise stated, with a median PFS resultant from each simulation. Empiric 95% CI estimates were calculated from each run.

RESULTS

Monotherapy source data including ORR, PFS and overall tumor volume change were digitally measured from available clinical trials (see online supplemental figure 1 for measurements). To first test the model, we calculated single agent ORR and a PFS distribution for each of the four agents (online supplemental figure 2). Next, using this in

Progression-Free Survival Estimates



Figure 1 Forest plot of progression-free survival (PFS) of combination TKI/IO therapy. In silico combination (red) and published registration data (blue) reported, with 95% CI estimates. IO, immuno-oncology; TKI, tyrosine kinase inhibitor.

silico monotherapy data, we modeled outcomes of the two combination therapies. The in silico median PFS for axitinib/pembrolizumab and cabozantinib/nivolumab were 17.9 months (95% CI 15.0 to 19.1) and 16.7 months (95% CI 14.4 to 19.3), respectively. Figure 1 contextualizes these findings with PFS data from the published

registration studies. Best radiographic responses in target lesions are summarized in figure 2 (together with data from published registration trials, online supplemental figure 3). In silico ORR for axitinib/pembrolizumab was 75.6% (95% CI 70.6 to 80.5) with 17.1% of patients achieving 100% target regression (95% CI 12.9 to 21.3).



Figure 2 (A–B) Waterfall plot of best response of combination therapy. In silico combination (red) and published registration data (blue) are superimposed for combination cabozantinib plus nivolumab and axitinib plus pembrolizumab.

For in silico cabozantinib/nivolumab, the model estimated an ORR of 70.4% (95% CI 64.4% to 75.4%) with 10.6% achieving 100% target regression (95% CI 7.4 to 14.4).

DISCUSSION

As more complex therapeutic strategies emerge for patients with mRCC (eg, triplicate therapy), clinical tools which estimate efficacy thresholds are needed for clinical trial design and for benchmarking combination strategies. TKI/IO combinations have revolutionized outcomes for patients with mRCC, and superiority has been proposed to stem from the immunomodulatory effects of TKI which may promote IO activity. To estimate the effects of combining agents, other computational models have previously been applied to assess the contribution of individual drug activity with IO therapy.¹⁰ Here, we performed in silico modeling of TKI/IO combinations by integrating monotherapy clinical trial data. Through random pairing of response and 'best of two' read-outs (which ignore additive treatment effects), our model produced ORR and PFS data comparable to TKI/ IO registration data. Undoubtedly, differences in study populations, dosing levels, and our inability to model how treatment-related toxicities affect drug exposure and treatment duration limit the accuracy of these projections. In addition, as the model relies on selection of the 'best-of-two' treatment response, it is limited in simulating progressive disease as it underestimates response in patients with tumors harboring cross-resistance to both TKI and IO therapies. Nevertheless, these findings highlight how some patients may derive therapeutic benefit from early concurrent exposure to independent mechanisms of action with these agents. As the field looks forward to building on these doublet backbone therapeutic regimens, incorporating in silico modeling may allow for an enhanced appreciation of how these agents combine and may allow for early identification of combinations which harbor negative interacting effects when clinical data deviate from modeled simulations. Lastly, these results call for further translational efforts which integrate biomarkers to adaptively match therapies with tumor phenotypes dynamically.

The following software/libraries were used

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Twitter Dennis J Hsu @DennisJHsu

Contributors Authors confirm contribution to the paper as follows: study contribution and design: RRK, DH, MHV. Data collection and analysis; interpretation of results; manuscript preparation: all authors. All authors reviewed and approved the final version of the manuscript.

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ORCID iDs

Ritesh R Kotecha http://orcid.org/0000-0002-0223-3479 Dennis J Hsu http://orcid.org/0000-0003-4200-1098 Martin H Voss http://orcid.org/0000-0003-0551-5807

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