Modeling progression and treatment of prostate cancer using the Thales QSP software platform

Cameron Meaney, Ashley Markazi, Ray Joe, Conner Sandefur, Matthew McDaniel, Ryan Suderman Quantitative Systems Pharmacology Solutions, Simulations Plus Inc., Research Triangle Park, NC Contact: <u>cameron.meaney@simulations-plus.com</u>





Validation

Fitting

Responder

Responder

OBJECTIVE

- Metastatic, castration resistant prostate cancer (mCRPC) is an aggressive form of prostate cancer in which patients progress despite and rogen deprivation therapy
- The objective was to create a QSP platform model of mCRPC with an associated virtual population of mCRPC patients by training and validating to clinically relevant measures from many clinical trials simultaneously

METHODS

- A QSP model of mCRPC was developed using Thales, a QSP modeling platform specifically designed to facilitate, model, predict, and aid in the design of clinical trials
- The model includes immune cells, cancer cell killing, cytokine effects, and prostate-specific antigen (PSA) as a biomarker
- Virtual patients had primary and metastatic lesions, each with 3 distinct tumor ulletmicroenvironments varying in size, growth, immune penetration, and drug PK/PD
- Parameters were aligned with existing mCRPC models or estimated from public data lacksquare
- A virtual population of 100 mCRPC patients was generated by fitting and validating data from 9 clinical trials with 11 therapeutic compounds (monotherapy/combination)

- This model and virtual population can be used to evaluate new therapies for treating mCRPC patients
- Key clinical measures the model predicts include tumor diameter, serum PSA, RECIST1.1, ORR, BOR, 50%/90% PSA decreases, time/duration of response, and survival (PFS, OS)

RESULTS



Model Predictions of Overall Response Data

Abiraterone 1000mg qd + Olaparib 300mg bid. Categories are BOR and definitions are based on the RECIST1.1 criteria: PD – at least 20% increase in sum of tumor diameters with absolute increase of at least 5mm; SD – not CR, PR, nor PD; PR – at least 30% decrease in sum of target tumor diameters; CR – no remaining detectable tumors. Bars represent the model predictions as a percentage of the full virtual population, x represent clinical trial endpoint data. 90% confidence intervals of model predictions are also shown. Clinical trial data taken from [1], [2], [3].



Categories are percent of patients who achieved a 90% (left plot) or 50% (right plot) decrease from baseline PSA levels. Bars represent the model predictions as a percentage of the full virtual population, x represent clinical trial endpoint data. 90% confidence intervals of model predictions are also shown. Clinical trial data taken from [5], [6].

- The model is capable of accurately predicting clinical endpoint for various drugs and dosing regimens including BOR, ORR, and PSA decreases
- Note that these results are for a single global population fit to all of these (and other) data simultaneously; hence, the generated virtual population can predict responses to novel compounds • Altogether, data from 9 different clinical trials spanning 11 different therapies with multiple dosing



Non-Responder Responder **Figure 2**: Top – Enzalutamide 160mg qd; middle – Abiraterone 1000mg qd; bottom - Nivolumab 1mg/kg q3w + Ipilimumab 3mg/kg q3w followed by Nivolumab 480mg q4w. Categories are ORR based on the RECIST1.1 criteria – responder is PR or CR, nonresponder is SD or PD. Bars represent the model predictions as a percentage of the full virtual population, x represent clinical trial endpoint data. 90% confidence intervals of model predictions are also shown. Clinical trial data taken from [3], [4], [5].



regimens are simultaneously included as fitting and validation targets to train and validate the global virtual population (see references for some of the therapies and trials); Thales is capable of handling various types of data common in clinical trials natively

• The model also accurately predicts cytokine, immune cell, and PSA levels for healthy individuals, and can be extended to include mCRPC subtypes using biomarkers, disease phases, or genetic mutations

CONCLUSION

- These results provide proof-of-concept that mCRPC progression and treatment can be modeled using a QSP platform model
- The model and virtual population accurately predict clinical endpoint data for many clinical trials and therapeutic regimens simultaneously
- Additional therapies could be added to further train the population, both enhancing model performance and allowing clinically relevant predictions for novel compounds such as optimal firstin-human dose, special populations of interest, or novel therapeutic combinations
- The mCRPC model extends our solid tumor modeling library, which contains a curated set of processes common to many solid tumor indications

REFERENCES

- Oudard S et al. Cabazitaxel Versus Docetaxel As First-Line Therapy for Patients With Metastatic Castration-Resistant Prostate Cancer: A Randomized Phase III Trial-FIRSTANA. J Clin Oncol. 2017
- Ryan CJ et al. Abiraterone in metastatic prostate cancer without previous chemotherapy. N Engl J Med. 2013
- Saad F et al. Olaparib plus abiraterone versus placebo plus abiraterone in metastatic castrationresistant prostate cancer (PROpel): final prespecified overall survival results of a randomised, double-blind, phase 3 trial. Lancet Oncol. 2023
- Beer TM et al; PREVAIL Investigators. Enzalutamide in metastatic prostate cancer before chemotherapy. N Engl J Med. 2014
- Sharma P et al. Nivolumab Plus Ipilimumab for Metastatic Castration-Resistant Prostate Cancer: Preliminary Analysis of Patients in the CheckMate 650 Trial. Cancer Cell. 2020
- Beer TM et al. Randomized, Double-Blind, Phase III Trial of Ipilimumab Versus Placebo in Asymptomatic or Minimally Symptomatic Patients With Metastatic Chemotherapy-Naive Castration-Resistant Prostate Cancer. J Clin Oncol. 2017

