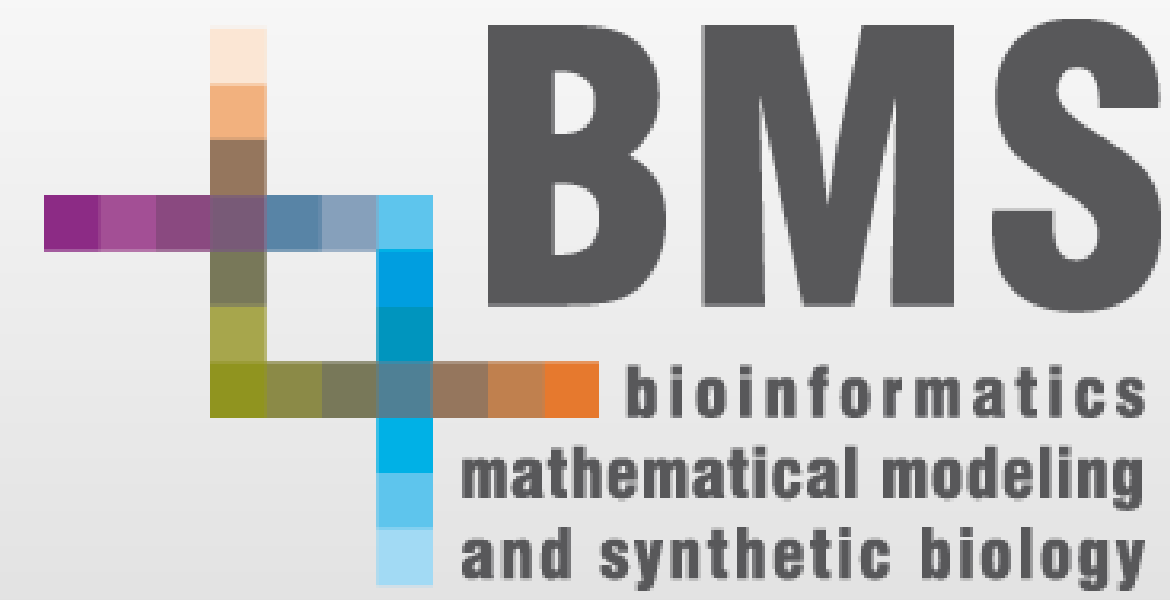


Predicting tumor volume doubling time and progression-free survival curves in cancer patients from patient-derived-xenograft (PDX) models: a translational model-based population approach

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BACKGROUND

Tumor volume doubling time (TVDT):

- time required by a tumor to doubling its volume;
- computed from at least 2 subsequent tumor volume measurements

$$TVDT = \frac{(t_2 - t_1) \ln(2)}{\ln(TV_2 / TV_1)}$$

- essential in clinical cancer management for many medical decisions;
- correlated with response to therapy, tumor progression and patient survival.

Progression-free-survival (PFS):

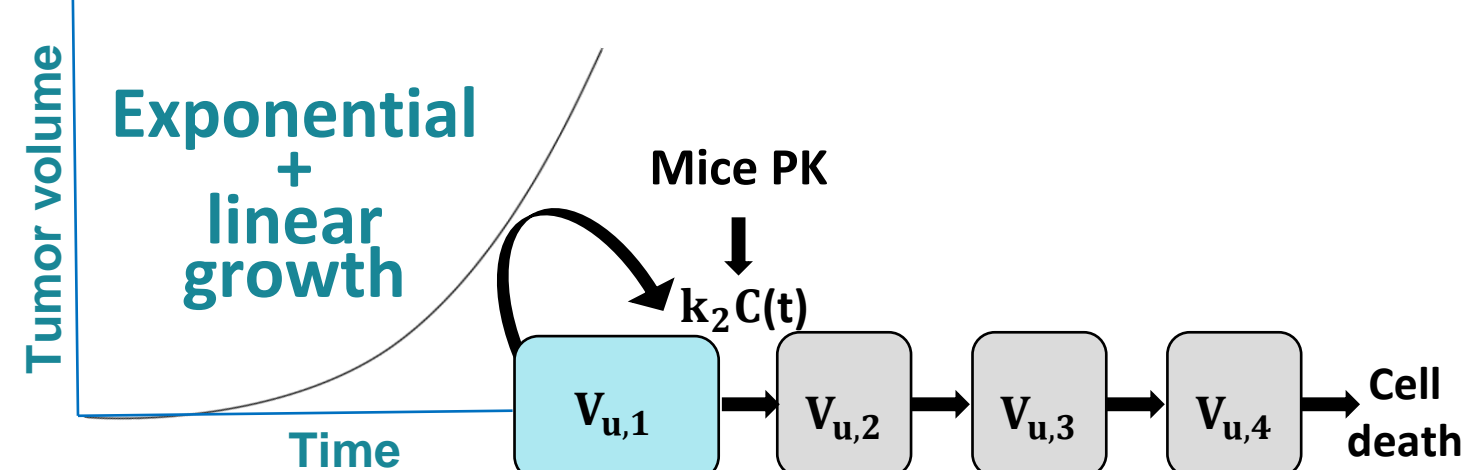
- timeframe between study randomization and tumor progression or death, whichever occurs first;
- inherently linked to tumor growth (TG) dynamics;
- summarized through Kaplan-Meier (KM) curves;
- primary endpoint in early clinical studies: in absence of placebo-controlled arm, quantification of investigated treatment benefit in terms of PFS timespan increment respect to standard therapies.

OBJECTIVE: Developing a translational modeling framework to predict TVDTs and PFS curves in an untreated cancer patient population from TG data in patient-derived xenograft (PDX) mice [1]. Extending the translational population approach to predict PFS curves in pancreatic (PA) and liver (LI) cancer patients treated with Gemcitabine and Sorafenib, respectively.

METHODS

Modeling PDX data

Simeoni TGI model [2]

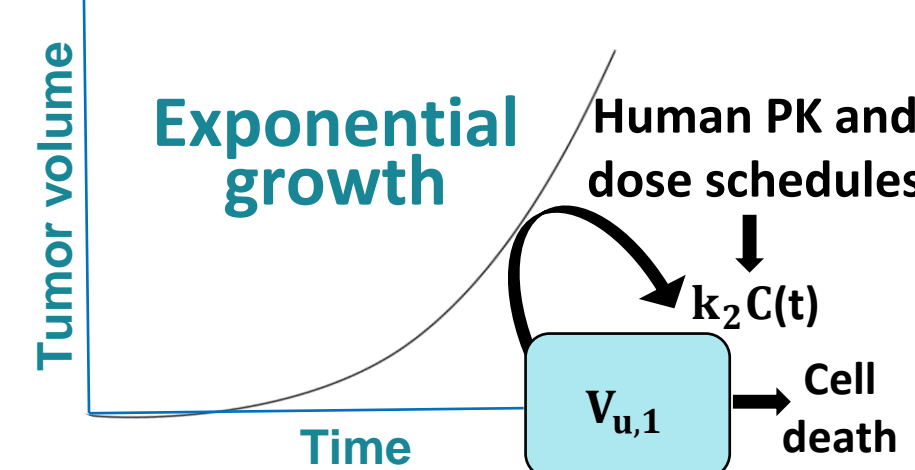


- non-linear mixed effect approach to account for inter-PDX variability
- Identification of log-normal distributions of:
 - exponential growth rate, λ_{Mice}
 - 11 cancer types
 - 25 PDX models for cancer type ($n_{tot}=265$)
 - anticancer drug potency, $k_{2,Mice}$
 - Gemcitabine on PA PDX models ($n=27$)
 - Sorafenib on LI PDX models ($n=24$)

CROWN BIOSCIENCE

Scaling from PDX mice to humans

TGI model in humans



- Exponential growth model with direct killing effect
- Allometric scaling

$$\lambda_{Human} = \lambda_{Mice} \left(\frac{BW_{Human}}{BW_{Mice}} \right)^{-1/3}$$

$$k_{2,Human} = k_{2,Mice} \left(\frac{BW_{Human}}{BW_{Mice}} \right)^{-1/3}$$
- Log-normal distributions where inter-patient variability = inter-PDX variability



Predictions in cancer patients

Under exponential tumor growth assumption

$$TVDT_{Human} = \frac{\ln(2)}{\lambda_{Human}}$$

Considering

- a single spherical mass
- only progression event

20% increase of tumor diameter \longleftrightarrow 72.8% increase of tumor volume



- KM-visual predict check (VPC) plots from 1000 studies, each with 200 patients

RESULTS

Comparison with clinical TVDTs from literature:

- 91% of predicted TVDT medians within the 2-fold of observations; RMSE=0.9244 (Fig.2).
- 83.59% of individual-PDX TVDTs predictions within clinical 80%CI ranges (Fig.2).
- Predicted 80%CI TVDT ranges fell within the observed ranges, with a general underestimation of the clinically observed inter-patient variability.

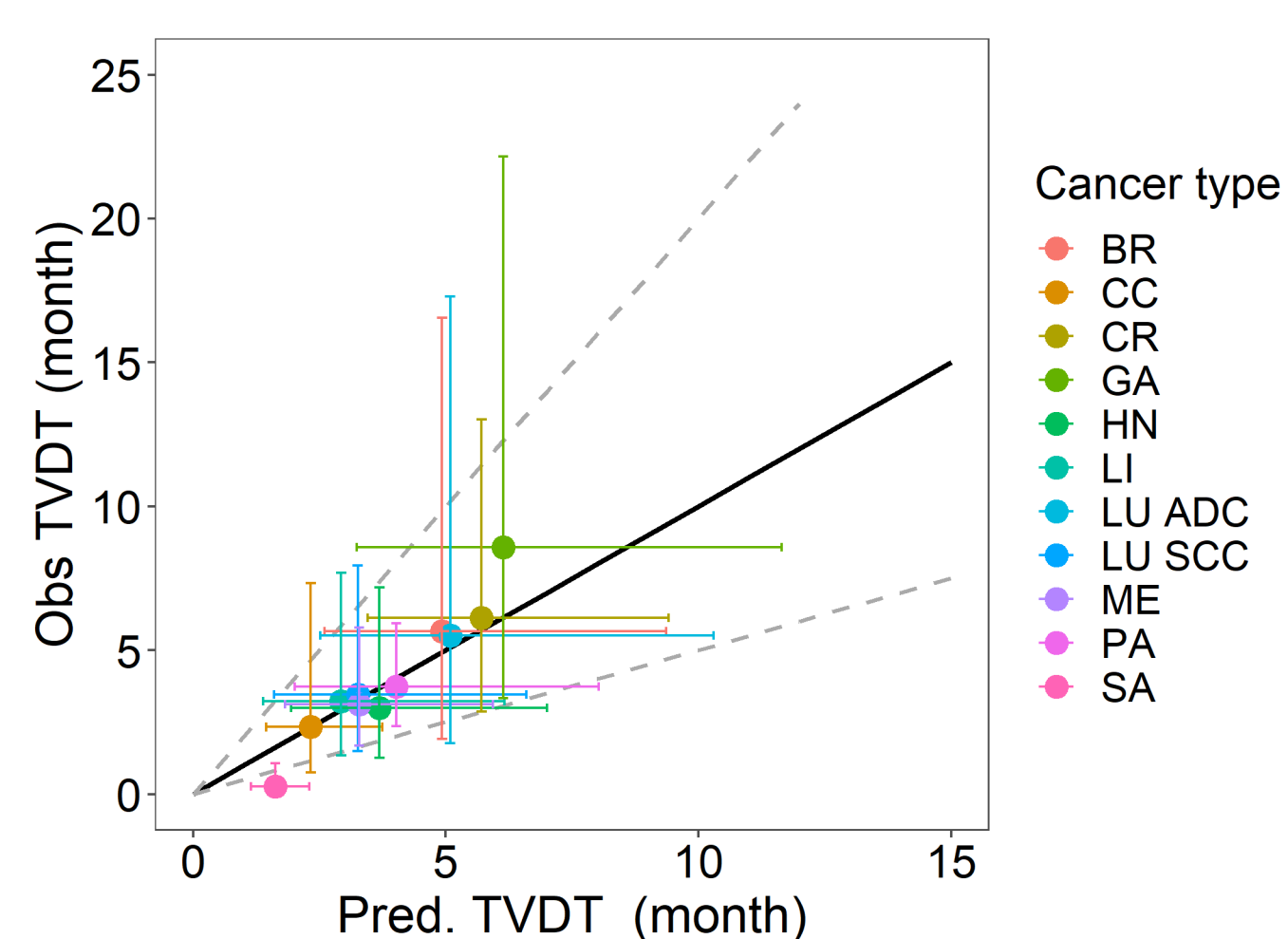


Fig 1: Pred. versus obs. TVDTs stratified by cancer type. Dots = typical values. Bars = 80% CI. Solid line=identity. Dashed lines=2-fold area.

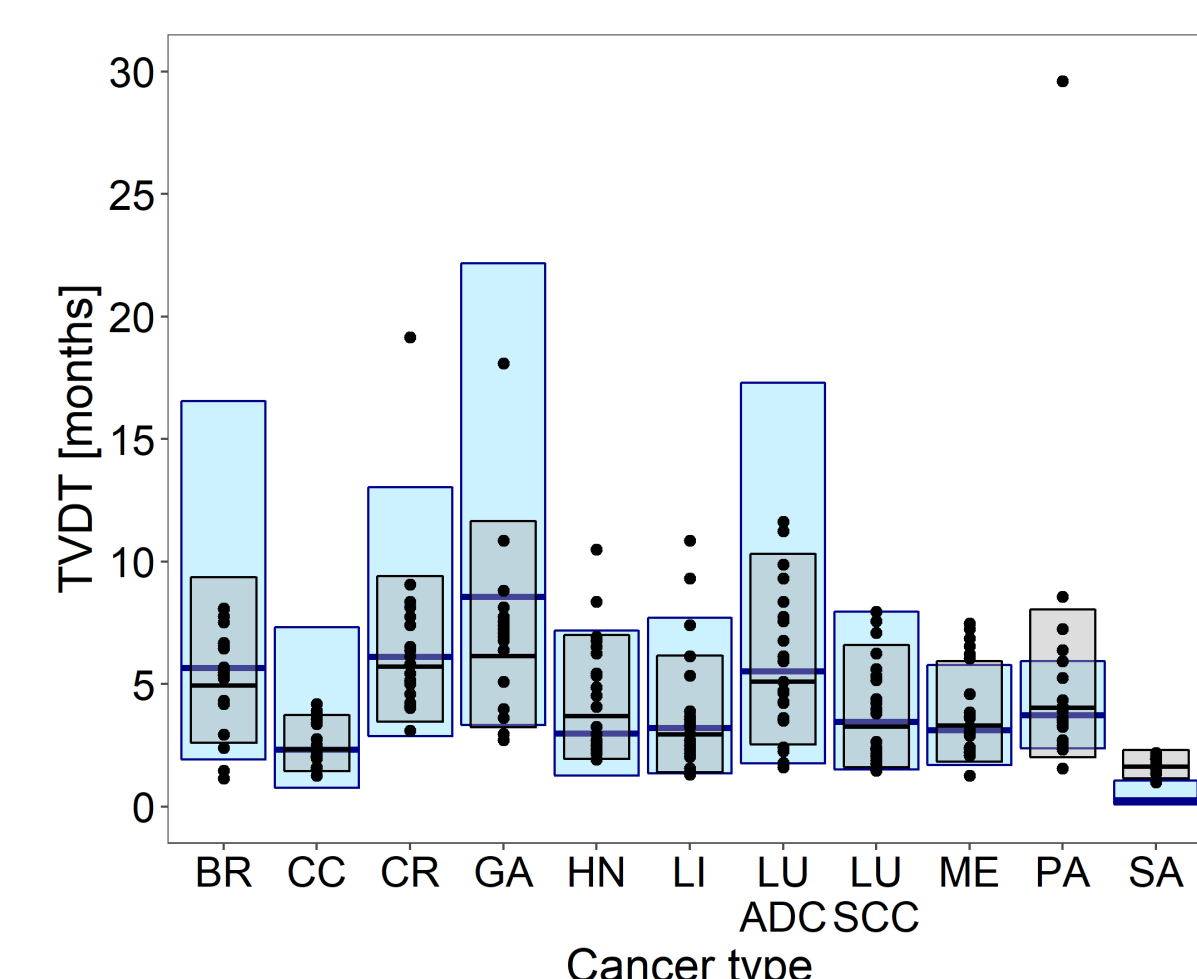
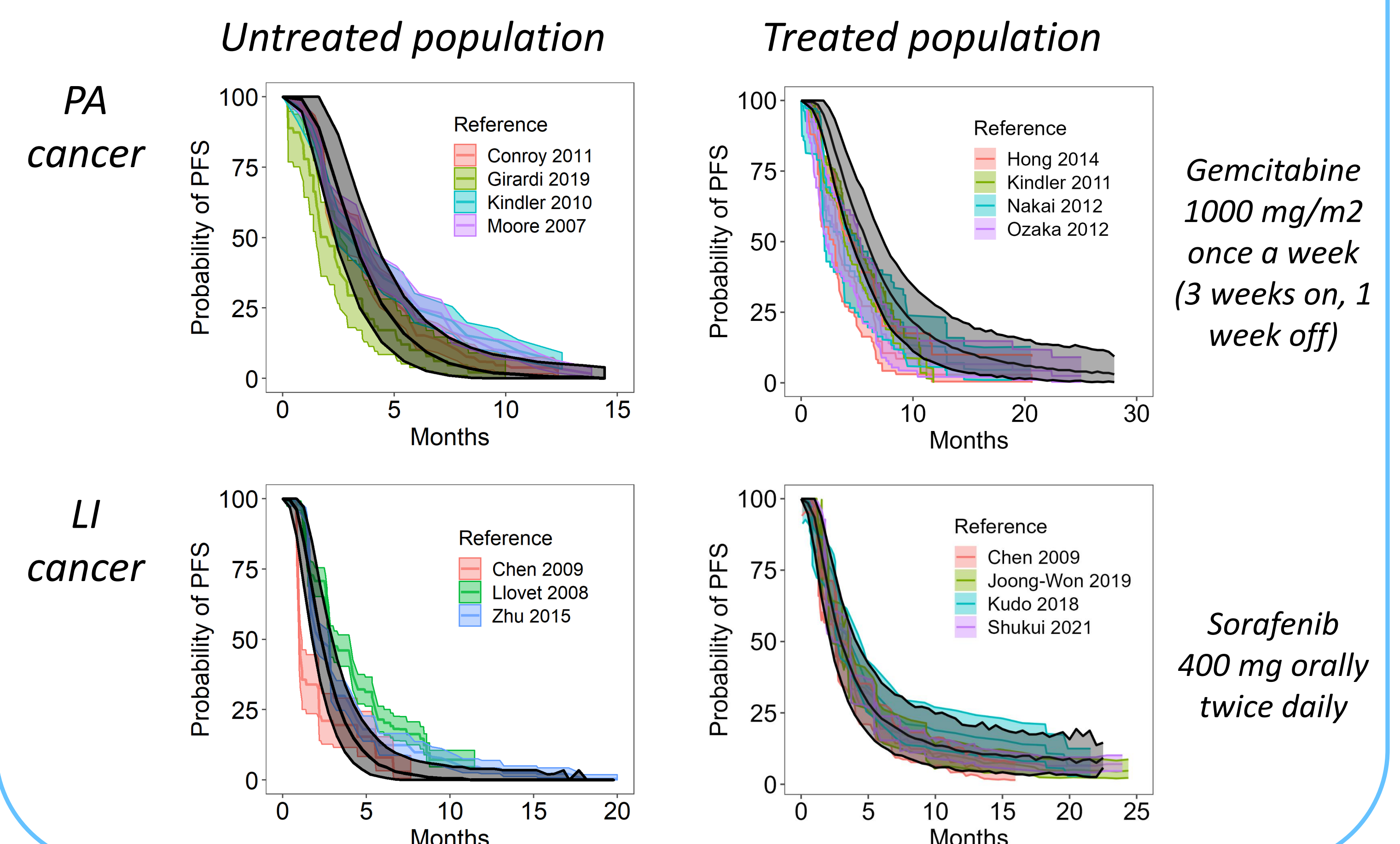


Fig 2: Box plots (medians and 80% CI) of obs (lightblue) and pred (grey) TVDTs. Dots=individual TVDTs predicted from the analysed PDX models.

Comparison with published PFS data from different clinical trials; patient cohorts receiving placebo or active treatments exerting little to no effect were used for comparison with untreated population



CONCLUSIONS: The proposed model-based translational framework provides a powerful tool to i) increase the knowledge on TVDT without the need of further tumor volume measurements; ii) early predict the treatment effect on PFS starting from a panel of TGI studies in PDX mice.

REFERENCES

[1] HuBase database, Crownbio Bioscience Inc., <https://www.crownbio.com/>.

[2] Simeoni M, et al. Predictive pharmacokinetic- pharmacodynamic modeling of tumor growth kinetics in xenograft model after administration of anticancer agents. Cancer Res. 2004; 64:1094–101.