

Modeling the Emergence of Resistance to Chemotherapeutics with Virtual Tumour

PHYSIOMICS
rational therapeutics





AACR Minisymposium: Novel and Integrative Analyses of Cancer Genome Data

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I have the following financial relationship to disclose:

Employee of Physiomics plc

and

I will not discuss off label use and/or investigational use in my presentation.

Chemotherapeutic Resistance: Background



- ➔ Major cause of treatment failure
- ➔ Arises from genetic mutations/epigenetic changes
- ➔ Intrinsic or acquired, to one drug or many simultaneously
- ➔ Inter- and intra-tumour heterogeneity compound the problem
- ➔ Personalized medicine proposed as a way to prevent and overcome resistance – the future of cancer therapy...
- ➔ Emergence of resistant disease can be delayed by novel dosing regimens

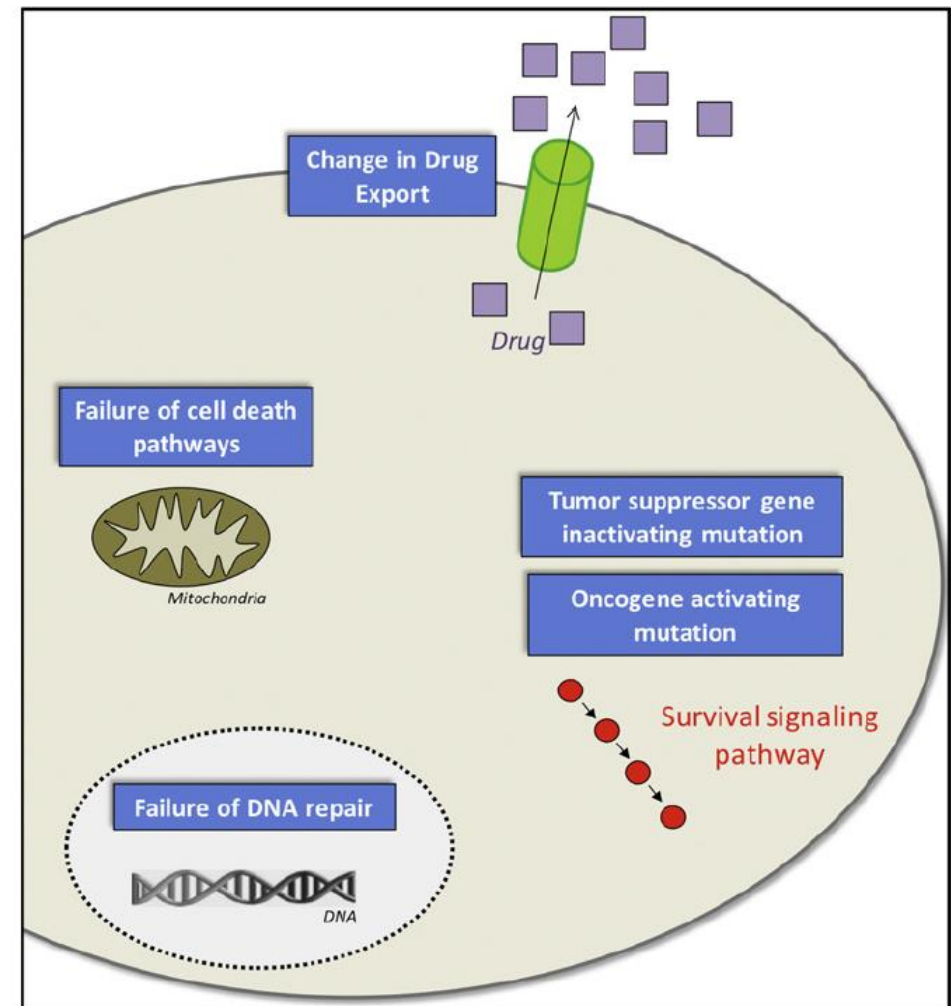


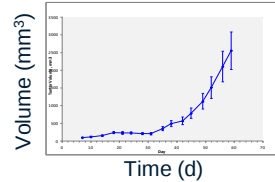
Fig. 1. Overview of drug resistance common mechanisms.

From Rebutti and Michiels (2013) *Biochem. Pharmacol.* 85: 1219-26.

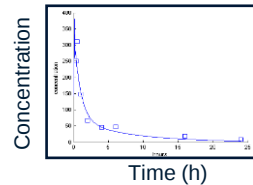
Virtual Tumour



1) Cell line growth data: Control xenograft growth curves

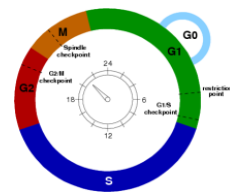
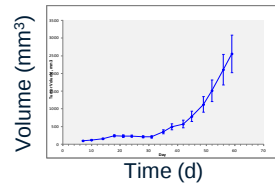


2) Compound PK data: Plasma/tumor time courses



3) Compound PD data:

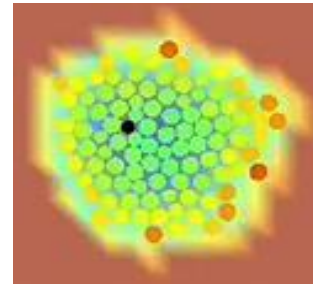
- Xenograft inhibition data (growth curves)
- Biomarker data (cell cycle, cell death, target biomarkers)



Required data
Optional data



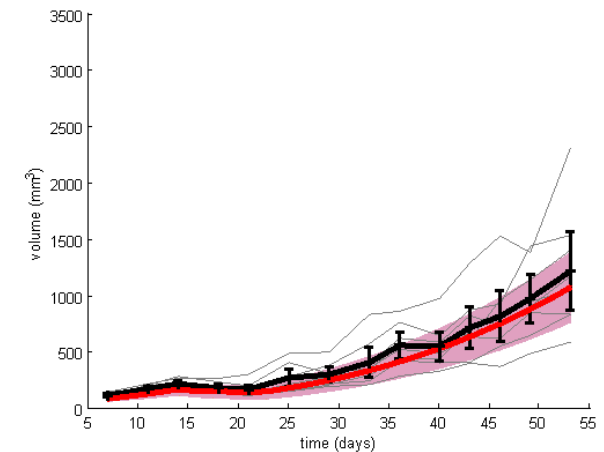
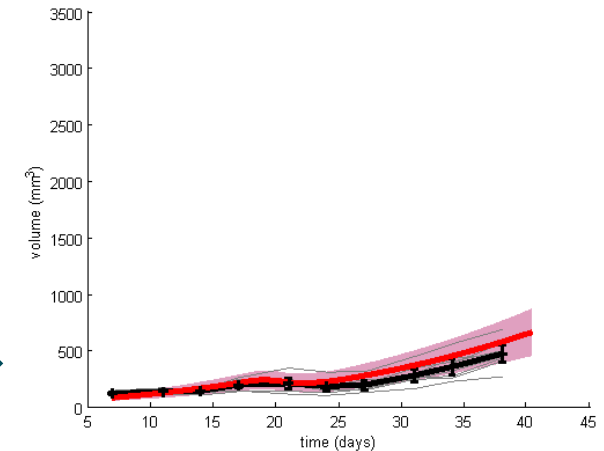
Virtual Tumour



proprietary cell population model



Tumor growth inhibition for selected schedules



Prediction

Mean → Confidence interval

Experiment

Individual mice
Mean and standard error

Case Study

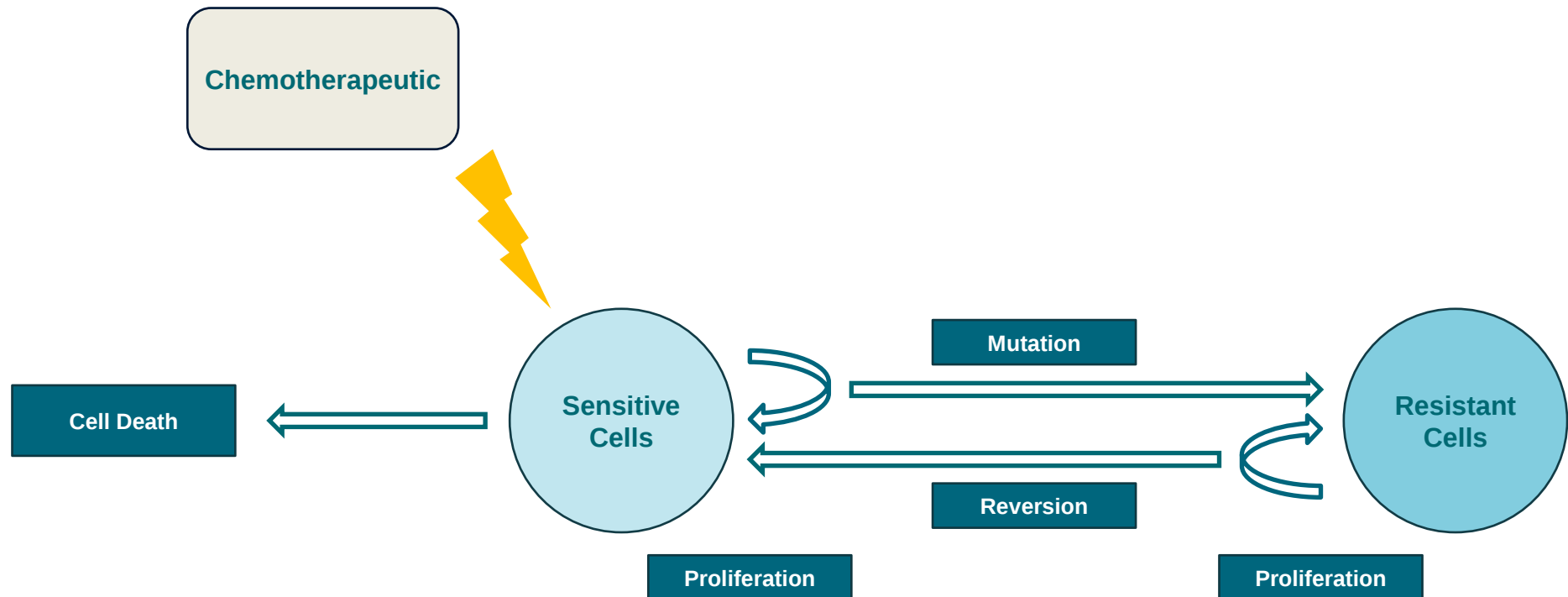
Predicting the efficacy of an optimized regimen to minimize the emergence of chemotherapeutic resistance



ADVANCE:	Validation of the VT resistance module
OBJECTIVE:	To determine whether our technology could accurately predict the efficacy of a novel dosing regimen of vemurafenib in metastatic melanoma PDX in minimizing the emergence of resistance
PARTNER:	Internal – literature based
START POINT:	Published PK and PDX growth inhibition data for calibrating the resistance module, with supportive literature data to determine the structure of the model
DURATION:	Three months
OUTCOMES:	Correctly predicted the efficacy of an optimized vemurafenib regimen for minimizing the appearance of resistance “time to regrowth”

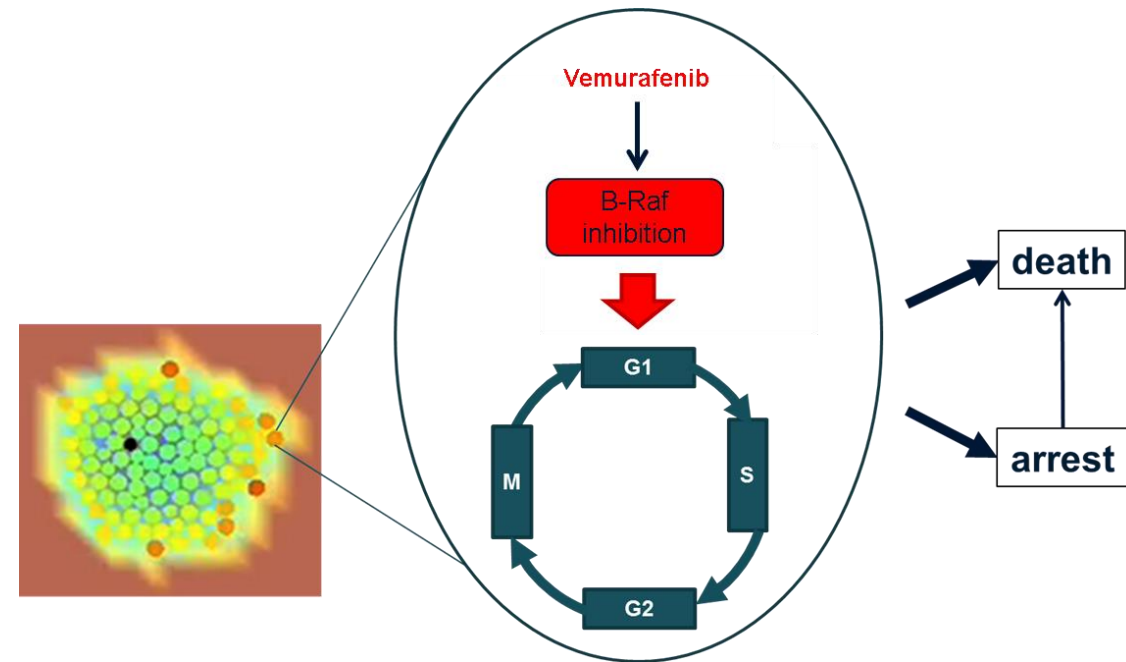
Modeling Drug Resistance

- ➔ A provisional model for chemotherapeutic resistance has been developed as an add-on module for the Virtual Tumour platform



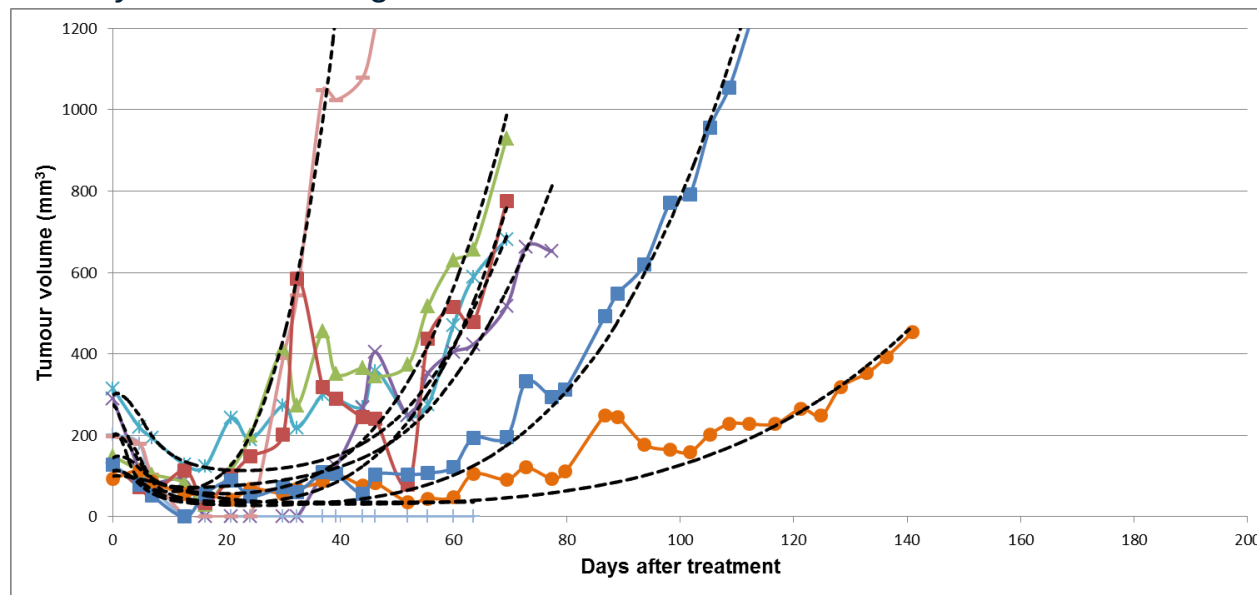
Modeling Vemurafenib Mechanism of Action

- ➔ Each cell in the Virtual Tumour contains set of equations that simulate:
 - ➔ Cell-cycle phase progression and checkpoints
 - ➔ The effect of anti-neoplastic drugs/agents
- ➔ The state-dependent effect of a drug on individual cells is modeled according to our understanding of the mechanism of action from literature and biomarker data
- ➔ Vemurafenib inhibits B-Raf/MEK/ERK pathway, leading to cell cycle arrest in G1 and cell death in sensitive cells
- ➔ Balance for cell arrest and cell death can be estimated using experimental observations and cell-cycle biomarker level measurements in treated tumors (e.g. Topoll and pRB)



Model Comparison with Experimental Data

- ➔ Continuous dosing of HMEX1906 xenografts with vemurafenib results in rapid emergence of resistance (Das Thakur *et al. Modelling vemurafenib resistance in melanoma reveals a strategy to forestall drug resistance. Nature 494, 251-5 (1013).*)
- ➔ This effect can also be simulated using the model
 - ➔ The pre-existing population of resistant cells was fixed, while the growth kinetics and vemurafenib efficacy were calibrated individually for each xenograft

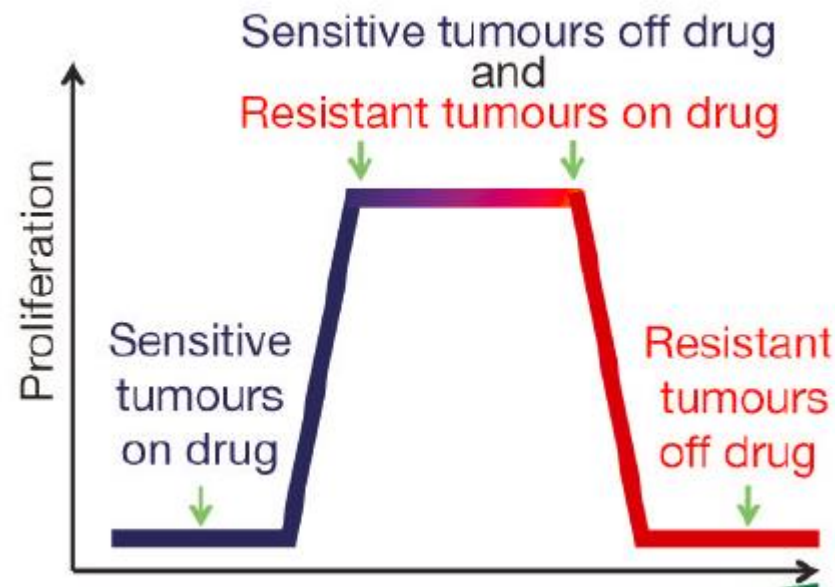


Tumor growth kinetics of naïve HMEX1906 xenografts dosed continuously with vemurafenib (15 mg/kg twice daily). Experimental data (coloured lines and symbols) from Das Thakur *et al. Nature 494, 251-5 (1013)*, overlaid by simulation results (dashed black line).



How to forestall the onset of drug resistance?

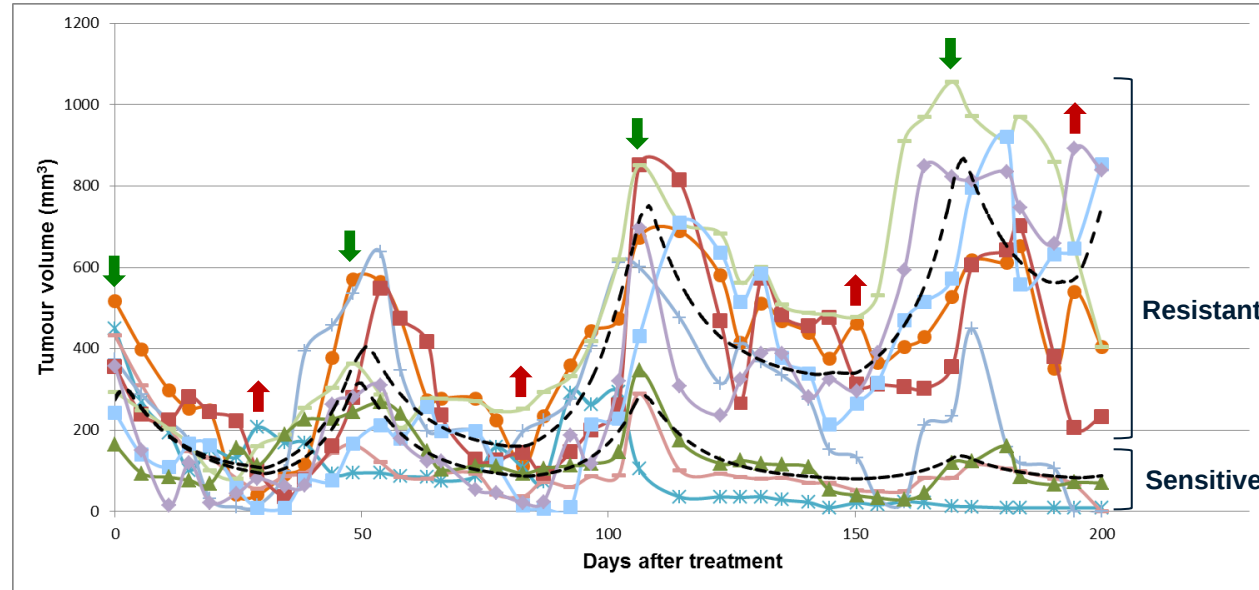
- ➔ Das Thakur et al. suggest to use discontinuous dosing regimens
- ➔ Intermittent dosing significantly delays the onset of drug resistance by exploiting the fitness deficit of drug-resistant cells in the absence of the drug



Das Thakur et al. Nature 494 251-5 (1013).

Model Comparison with Experimental Data

- ➔ Intermittent dosing of HMEX1906 xenografts with vemurafenib delays the emergence of resistance by exploiting the fitness deficit of drug-resistant cells in the absence of the drug
- ➔ The model also reflects this behaviour
 - ➔ The model was calibrated separately for two groups of xenografts, sensitive (no resistant cells) and resistant (pre-existing resistant cells)

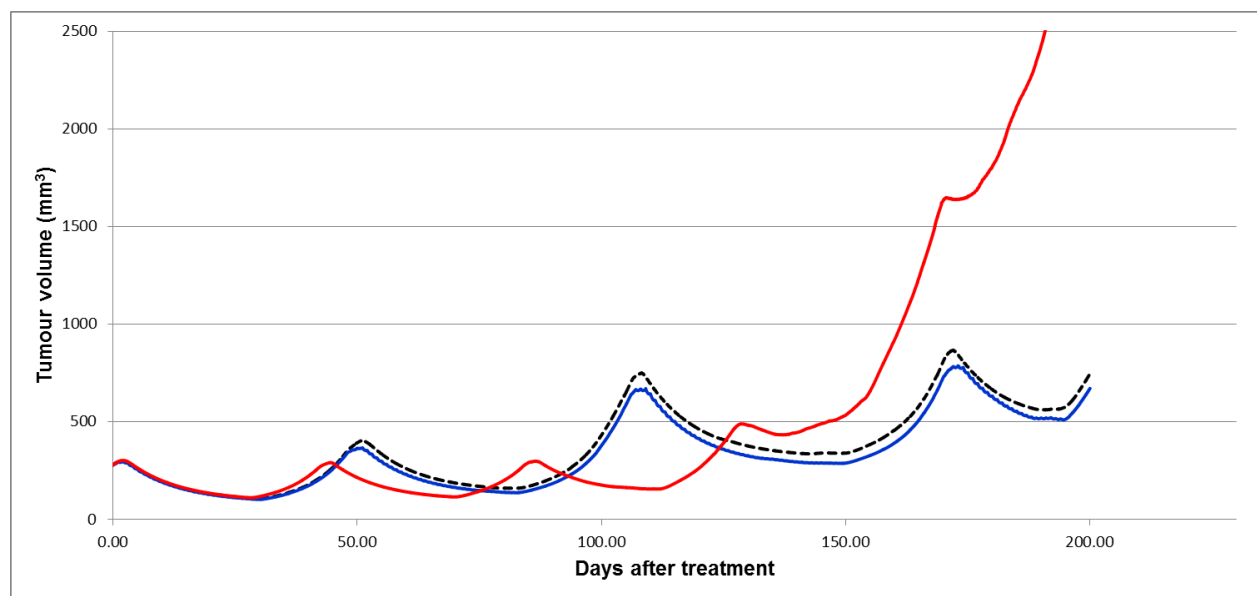


Tumor growth kinetics of naïve HMEX1906 tumors dosed intermittently with vemurafenib (15 mg/kg twice daily, adaptive schedule). Experimental data (coloured lines and symbols) from Das Thakur *et al.* Nature 494 251-5 (1013), overlaid by simulation results (dashed black line).

Investigation of Optimal Dosing Schedules to Delay the Emergence of Resistance



- ➔ The model also confirms that the 'adaptive' dosing regimen (dosing 15 mg/kg twice daily or 30 mg/kg once daily on an intermittent and adaptive on/off schedule) that was employed in the experimental study is close to optimal
- ➔ A strict four-week on two-week off schedule (15 mg/kg twice daily) fails to delay the emergence of resistance



Simulated tumor growth kinetics of naïve HMEX1906 tumors dosed intermittently with vemurafenib. Original adaptive schedule (15 mg/kg twice daily), dashed black line; (30 mg/kg once daily), blue line; strict four-week on two-week off schedule (15 mg/kg twice daily), red line.



Summary and Next Steps

- ➔ **Successfully validated the resistance module of our Virtual Tumour technology**
 - ➔ Using fixed, non-adaptive scheduling of vemurafenib administration leads to the risk of early onset of resistance, which eventually overwhelms the population of drug-sensitive cells
 - ➔ On the other hand, monitoring tumour growth and predicting the optimal time at which treatment should be applied or withdrawn leads to a better outcome
 - ➔ Model can be used to help design better preclinical schedules
- ➔ **Next step 1: Integrate the resistant module into Virtual Tumour Clinical**
 - ➔ Use existing validated translational Virtual Tumour for metastatic melanoma, developed in collaboration with Oxford University (Prof. Mark Middleton)
 - ➔ Accurate translation of preclinical efficacy reduces the number of clinical studies required to find optimal doses and schedules
- ➔ **Next step 2: Patient stratification with Virtual Tumour Clinical to predict time to relapse**
 - ➔ Stratify patient based of pERK expression and link it to PDX data
 - ➔ Optimize regimen for each group of patients, or individual patients

* Mistry, H. *et al.* Virtual Tumour Clinical development, part II: translational modelling of vemurafenib, selumetinib and docetaxel in metastatic melanoma, PAGE Meeting (2015), Alicante, Spain. [Download poster here](#).



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