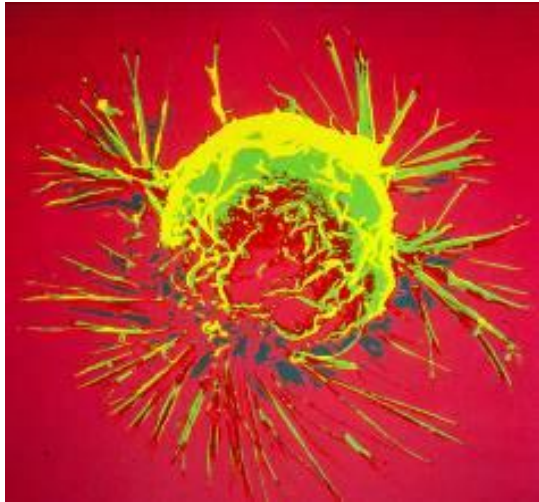


# Translational Modelling of Vemurafenib, Selumetinib and Docetaxel in Metastatic Melanoma with Virtual Tumour Clinical

PHYSIOMICS  
rational therapeutics

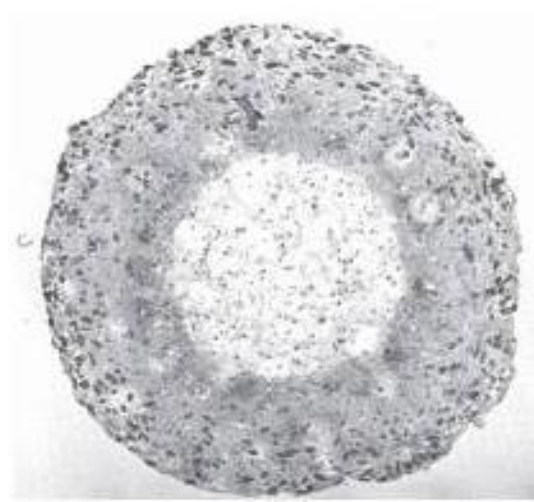


# The Need for a Virtual Tumour: Cancer is a Multi-Scale Phenomenon



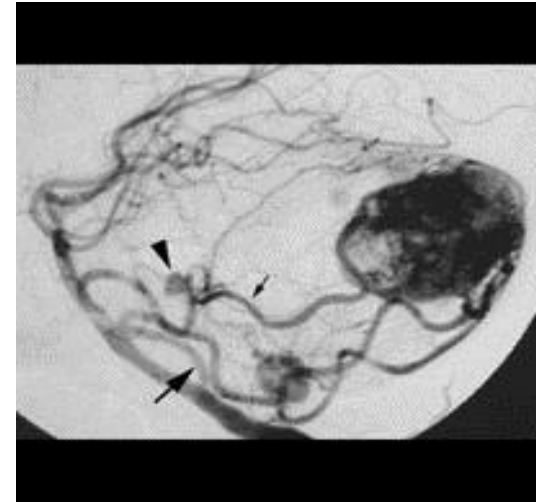
**Figure A:** Electron micrograph of a single breast cancer cell.  
Source: National Cancer Institute.

**10 - 30  $\mu\text{m}$  across**



**Figure B:** Avascular multicellular tumour.  
Source: J. Folkman, M. Hochberg, *J. Exper. Medicine*, 138: 745-753, 1973.

**$10^7$  cells  
1 mm across**



**Figure C:** Angiogram of a patient with a large vascular brain tumour. Arrows and arrowhead point to prominent blood vessels feeding this tumour. Source: Stanford Hospital.

**$10^{11}$ - $10^{12}$  cells  
5-20 cm across**



**Figure D:** Whole-body 18-FDG (fluorodeoxyglucose) imaging of a patient with small cell carcinoma of the lung. Source: Unité d'Imagerie Moléculaire et de Radiothérapie Expérimentale Cliniques Universitaires Saint-Luc Bruxelles.

***Cancer is a multi-scale phenomenon, hence it must be modelled on many levels***



# Virtual Tumour - Background

- ➔ **Physiomics' Virtual Tumour focusses on key tumour dynamics**
  - ➔ Tumour growth / spatial aspect
  - ➔ Individual cell / synchronisation
  - ➔ Predict drug effects on tumour
  - ➔ Does not try to replicate the full complexity of biological systems
- ➔ **Agent-based model, each cell (agent) contains a different instance of the model**
- ➔ **Tumours contain a heterogeneous cell population**



# Virtual Tumour - Background

## ➔ Preclinical

- ➔ Predicts the change in mean tumour volume over time
  - ➔ *Over 35 preclinical studies have confirmed the predictive capability of the model*
- ➔ The model describes the growth of a single tumour

## ➔ Clinical

- ➔ Predict the change in mean tumour diameter over time for all lesions
  - ➔ *From the preclinical work we have learnt that the mean behaviour is predictable*

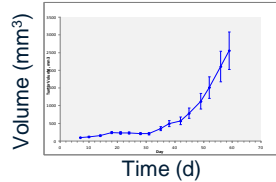
## ➔ Moving from preclinical to clinical setting and vice versa

- ➔ Current pharma approach involves merely matching PK between xenograft and man. We also take into consideration the different tumour growth dynamics.
- ➔ Adjust certain key parameters we have identified as important for reflecting the different tumour growth rates between xenograft and man.

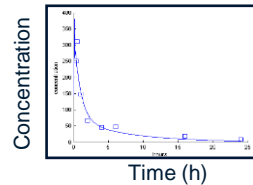


# Virtual Tumour Preclinical Mechanics

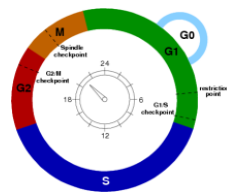
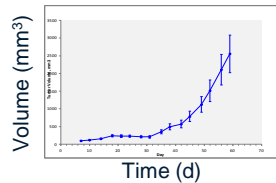
1) Cell line growth data: control xenograft growth curves



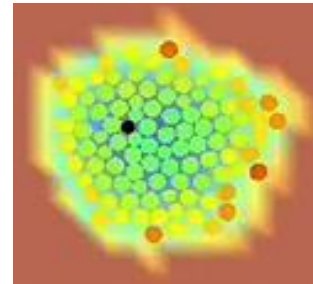
2) Compound PK data: plasma/tumour time courses



3) Compound PD data:  
- xenograft inhibition data (growth curves)  
- biomarker data (cell cycle, cell death, target biomarkers)



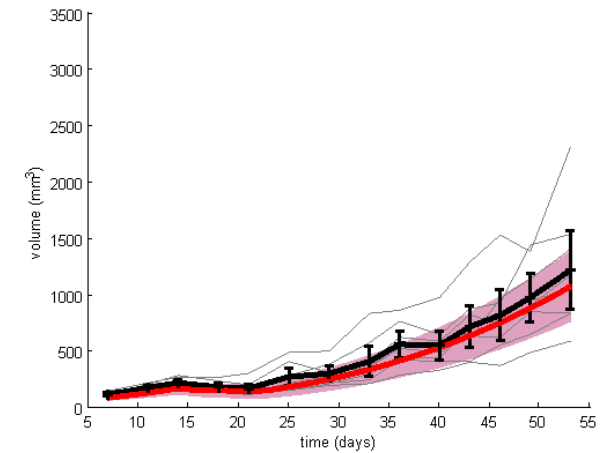
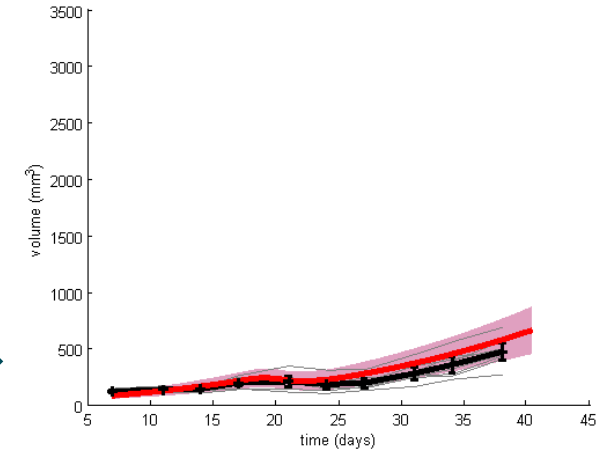
Virtual Tumour



proprietary cell population model



Tumour growth inhibition for selected schedules



Required data  
Optional data

Prediction

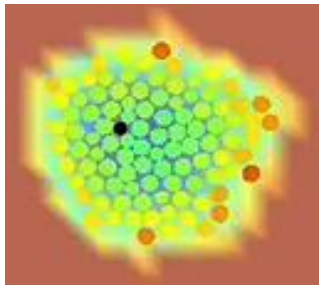
Mean → Confidence interval

Experiment

Individual mice  
Mean and standard error



## Preclinical Virtual Tumour



proprietary cell population model



### ➔ Literature Data across numerous tumour types

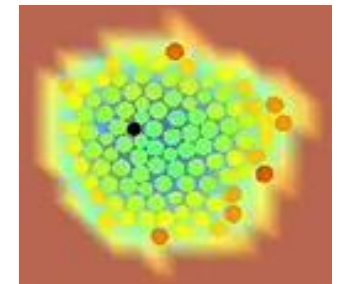
- ➔ Growth and decay rates of clinical tumours.
- ➔ Variability in durations of cell-cycle phases.

### ➔ Key patient data

- ➔ Human PK for drug of interest. Usually from a phase I study.
- ➔ How quickly a lesion shrinks. From clinical trials on other drugs in the same indication.



## Virtual Tumour Clinical



proprietary cell population model



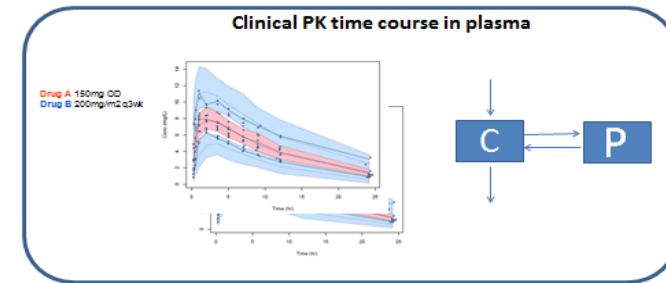


# Virtual Tumour Clinical Development

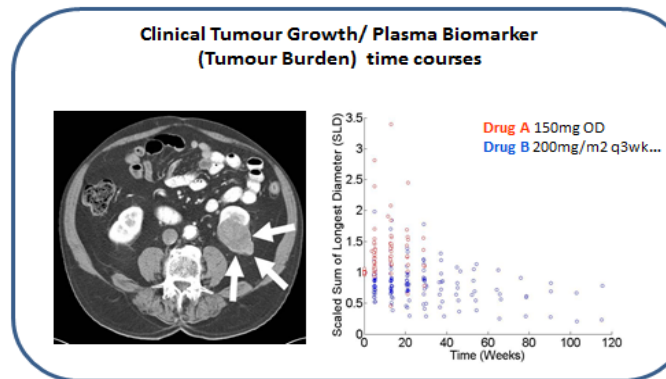
- ➔ Biomedical Catalyst funding award from the UK Technology Strategy Board (July 2013- March 2014)



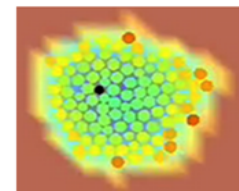
- ➔ NIH collaboration within metastatic castrate-resistant prostate cancer
- ➔ Oxford University clinical centre to look at three cancer types
- ➔ Advanced discussion with large pharma to provide large clinical data sets
- ➔ Early results suggest that the existing preclinical model architecture may be appropriate for making clinical predictions
- ➔ Large unmet need – interest from almost every potential partner



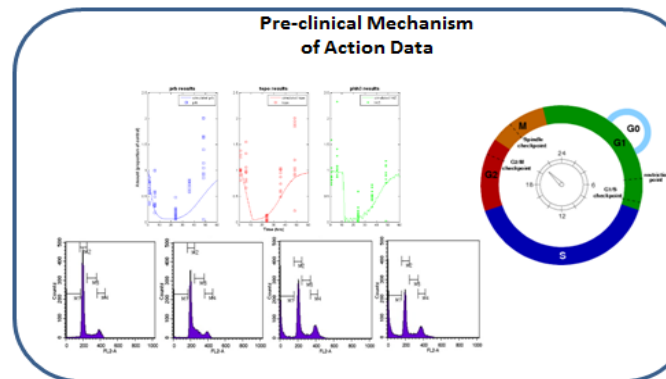
PK Model



Virtual Tumour



proprietary cell population model



PD Model

# Clinical to Preclinical (Back Translation) Metastatic Melanoma







## ➔ **Clinical data**

- ➔ 20 patients where each lesion was monitored over time
- ➔ Total number of evaluable lesions: 69
- ➔ FDA report contains a PK model

## ➔ **Preclinical data**

- ➔ COLO 205 xenograft (colorectal cell line with BRAF V600 mutation) for which we have change in tumour volume for different doses of the drug
- ➔ Literature PK model

## ➔ **Mechanism of action**

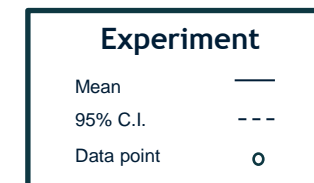
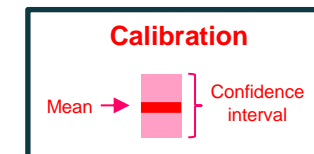
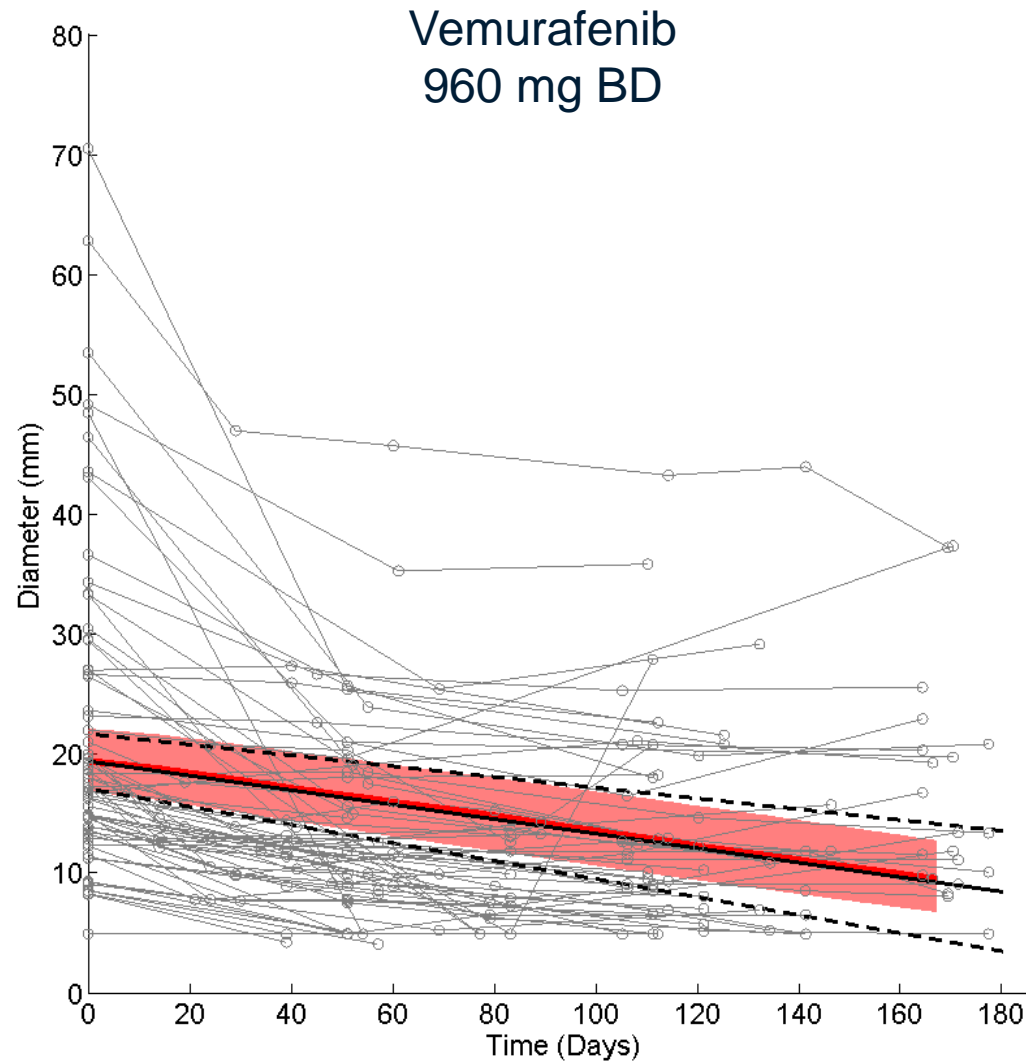
- ➔ B-Raf inhibitor
- ➔ Drug is known to exert its anti-tumour effect through causing G1 arrest



# Modelling Plan

- ➔ **Step 1:** Analyse clinical data using population analysis approach
- ➔ **Step 2:** **Calibrate** Virtual Tumour to the mean clinical signal
  - ➔ Clinical PK model sourced from literature
- ➔ **Step 3:** Switch clinical growth settings for preclinical growth settings and calibrate preclinical model to control growth
- ➔ **Step 4:** **Predict** preclinical monotherapy effects
  - ➔ Preclinical PK model sourced from literature
- ➔ **Step 5:** Compare prediction with actual results

# Steps 1 and 2: Clinical Calibration



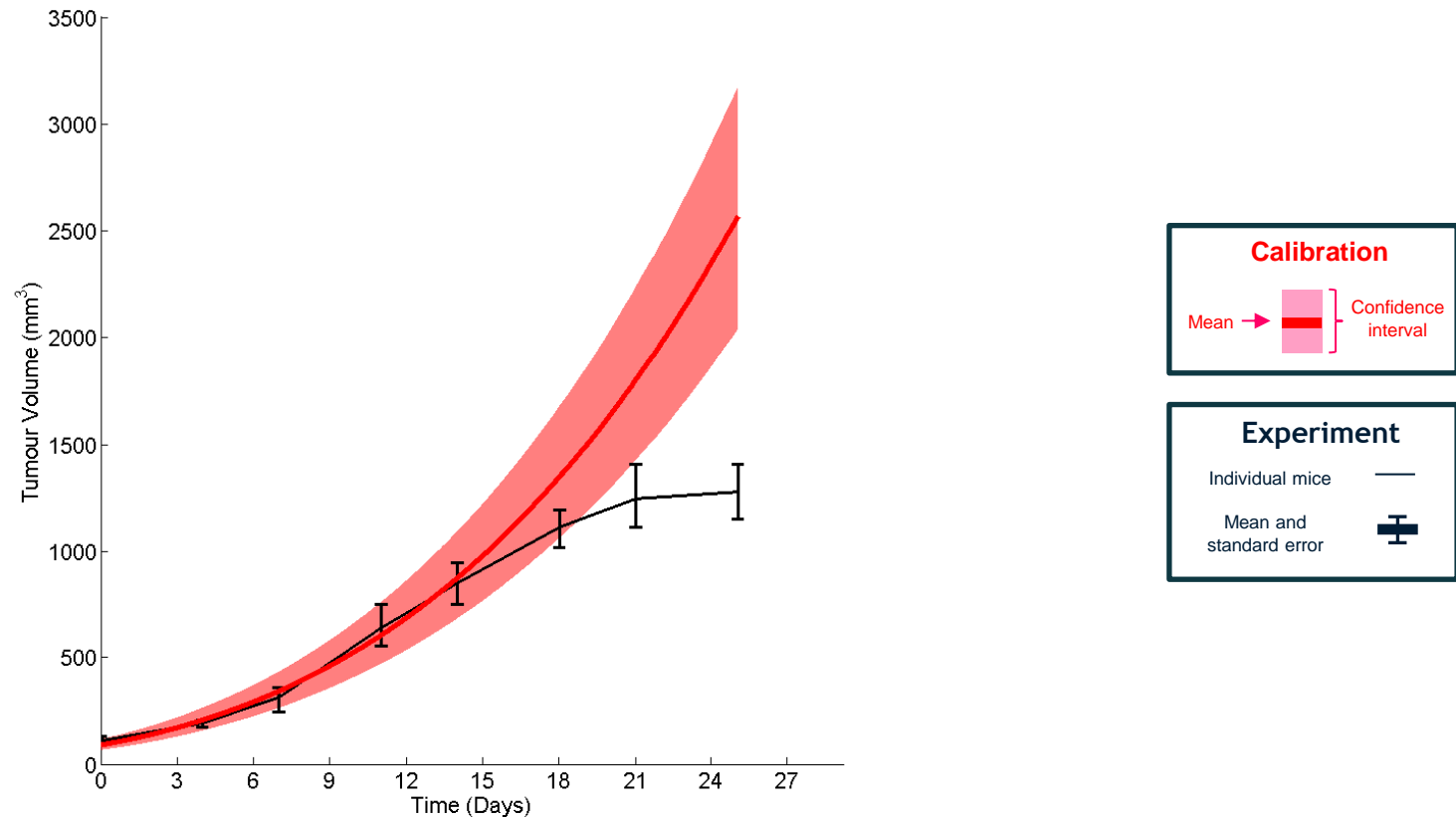
Evolutionary dynamics of cancer in response to targeted combination therapy. eLife. DOI: 10.7554/eLife.00747.001.



# Step 3: Preclinical Control Calibration

➔ **Mouse drop-outs affect the mean behaviour at late time points**

➔ Focus on early dynamics as mice are usually sacrificed once tumour volumes reach a certain size



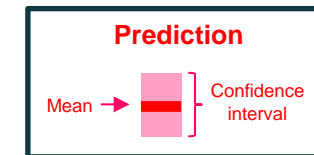
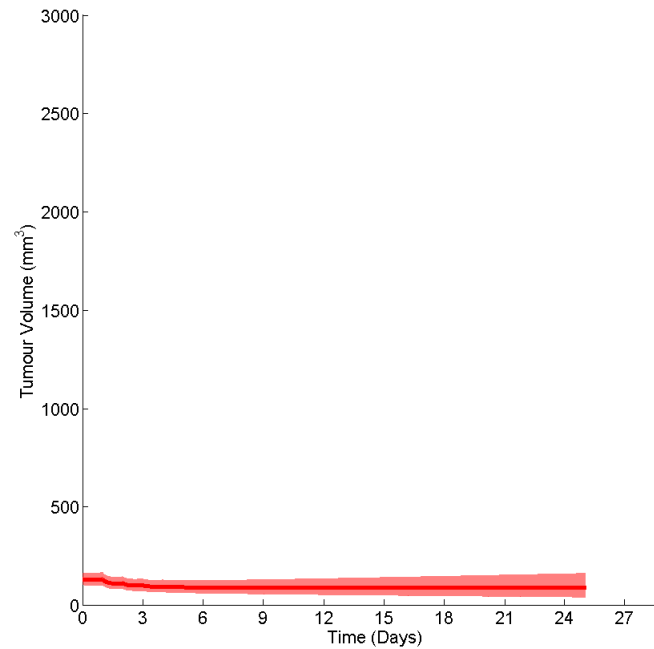
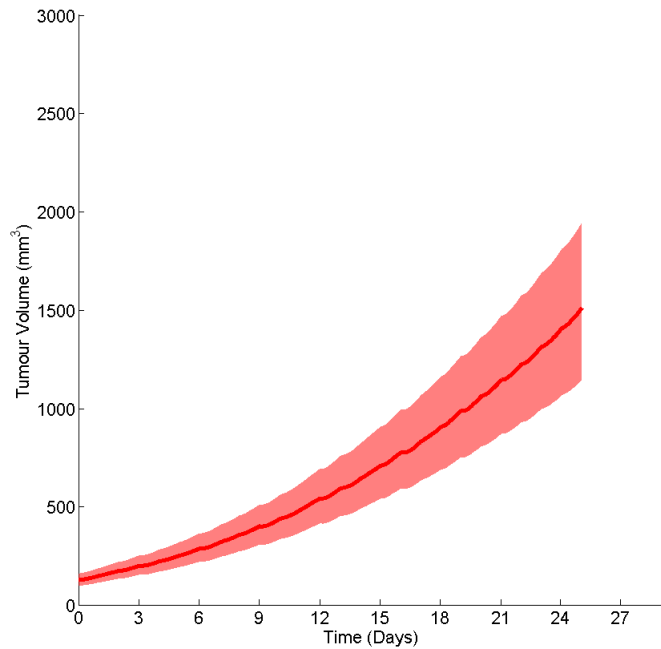
Clinical efficacy of a RAF inhibitor needs broad target blockade in BRAF-mutant melanoma. Nature. 2010. 467(7315): 596-599.



# Steps 4 and 5: Preclinical Prediction

➔ **Monotherapy predictions compare well with experimental observations**

➔ Left panel 6 mg/kg QD, right panel 20 mg/kg QD



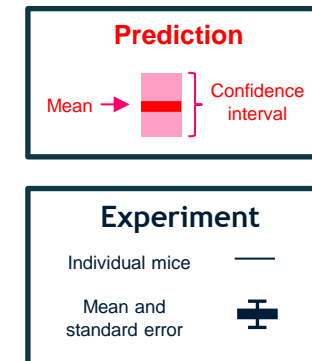
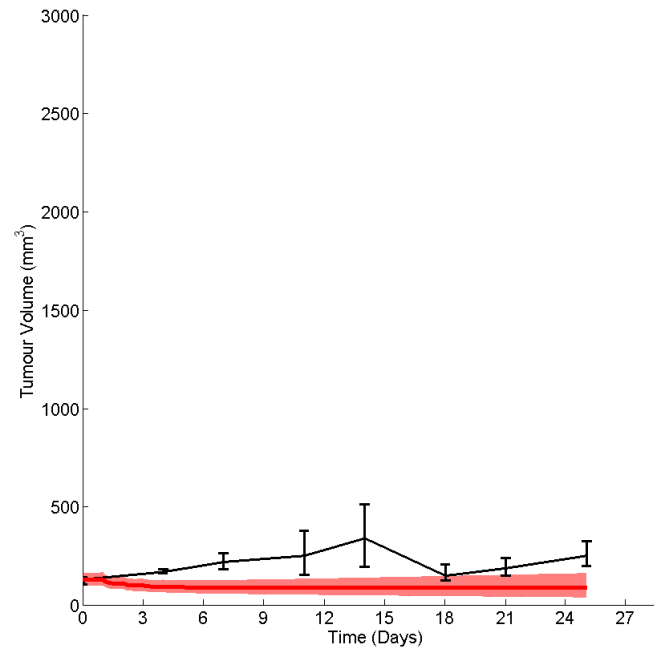
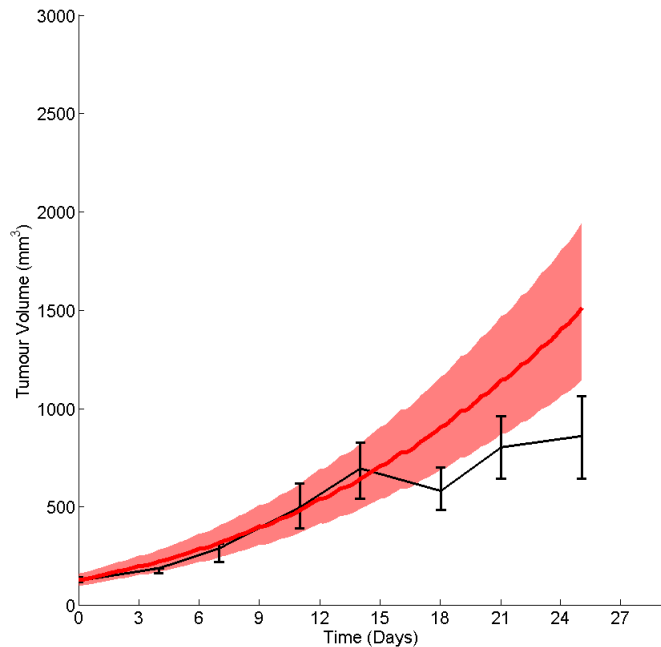
Clinical efficacy of a RAF inhibitor needs broad target blockade in BRAF-mutant melanoma. Nature. 2010. 467(7315): 596-599.





# Steps 4 and 5: Preclinical Prediction

- ➔ **Monotherapy predictions compare well with experimental observations**
  - ➔ Left panel 6 mg/kg QD, right panel 20 mg/kg QD
- ➔ **This was a colorectal cancer xenograft (COLO 205) which had BRAF V600 mutation**
  - ➔ Mutational background more important than tissue type? See later



Clinical efficacy of a RAF inhibitor needs broad target blockade in BRAF-mutant melanoma. Nature. 2010. 467(7315): 596-599



- ➔ **Calibrated Virtual Tumour to monotherapy changes in individual clinical lesions**
- ➔ **Model prediction:**
  - ➔ Captured the preclinical dynamics very well
- ➔ **Successful back-translational validation**
  - ➔ Predicted the effects reasonably well

**We shall now look at a forward translational project in this disease area...**

# Preclinical to Clinical Metastatic Melanoma



# Case Study – Translational Qualification

## Predicting Clinical Efficacy Using Preclinical Data



<b>ADVANCE:</b>	Qualification of the translational capability of the Virtual Tumour
<b>OBJECTIVE:</b>	To determine whether our technology could accurately predict the mean change in tumour size over time in a phase II clinical study of docetaxel vs. docetaxel/selumetinib in BRAF WT metastatic melanoma
<b>PARTNER:</b>	Mark Middleton, Oxford ECMC
<b>START POINT:</b>	Single drug xenograft dose-response data, preclinical and clinical PK
<b>DURATION:</b>	6 weeks
<b>OUTCOMES:</b>	Correctly predicted mean change in tumour size over time in both arms of the study and provided schedule options to ameliorate toxicities



- ➔ **AstraZeneca sponsored randomised phase II study: docetaxel/selumetinib v docetaxel**
  - ➔ 40 patients in each arm
  - ➔ ~100 lesions in each arm
  - ➔ BRAF WT setting
- ➔ **Selumetinib is a MEK inhibitor being investigated in numerous disease areas**
  - ➔ Phase III combination with docetaxel currently ongoing in NSCLC
- ➔ **Trametinib (GSK) MEK inhibitor was approved last year in the BRAF MUT setting**
- ➔ **Literature search was required for:**
  - ➔ Preclinical xenograft and PK
  - ➔ Clinical PK





# Modelling Plan

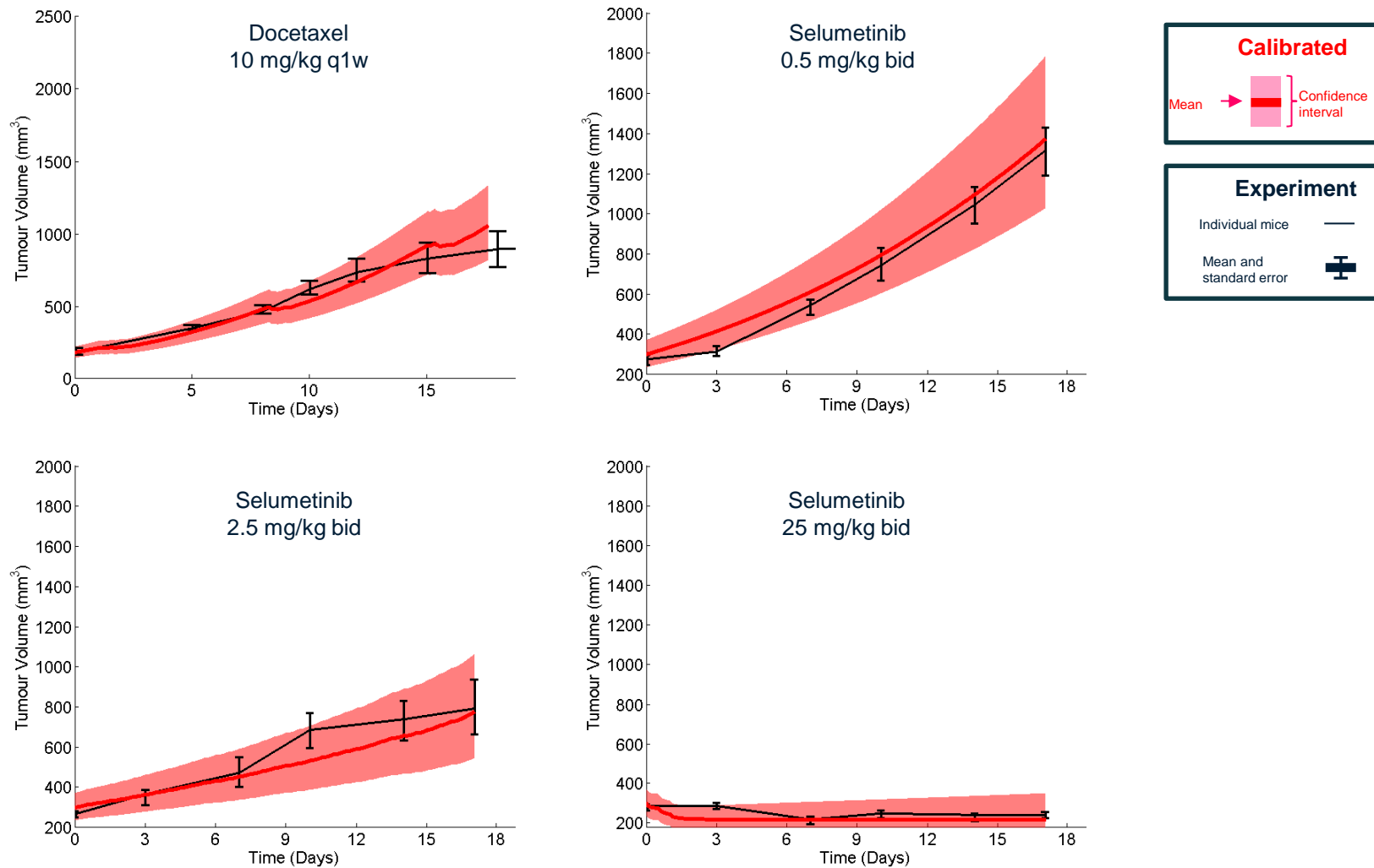
- ➔ **Step 1: Calibrate** Virtual Tumour to preclinical data for each agent
  - ➔ Literature PK and xenograft data sourced from literature
- ➔ **Step 2:** Switch preclinical growth settings for clinical growth settings
- ➔ **Step 3: Predict** the two-arm phase II trial
  - ➔ Clinical PK models sourced from literature
- ➔ **Step 4:** Population analysis of the clinical study
- ➔ **Step 5:** Compare prediction with actual result

# Case Study – Translational Validation

## Step 1: Preclinical Calibration



### ➔ Calibration of the Virtual Tumour to preclinical monotherapy data

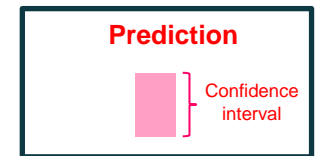
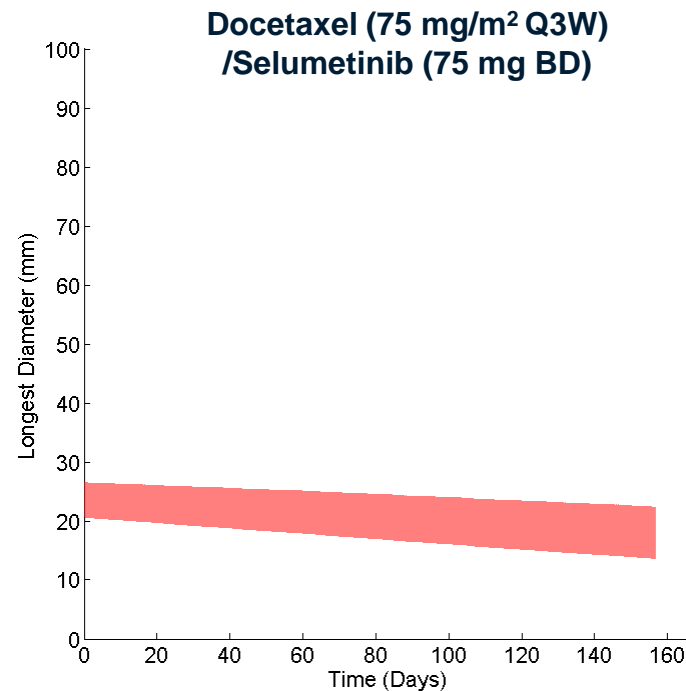
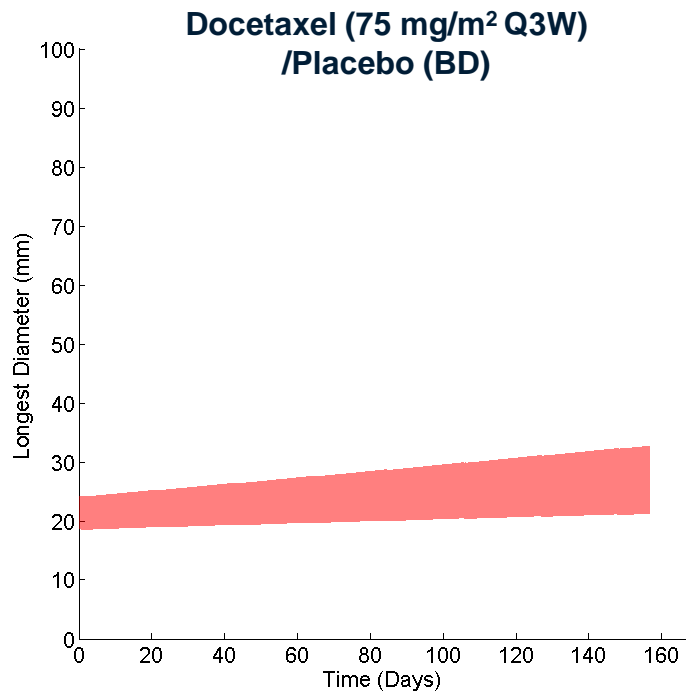


# Case Study: Translational Qualification

## Steps 2 and 3: Prediction



- ➔ **Replace preclinical growth settings with clinical growth settings**
  - ➔ Baseline longest diameters are provided as initial inputs
- ➔ **Replace preclinical PK with clinical PK and simulate predictions**

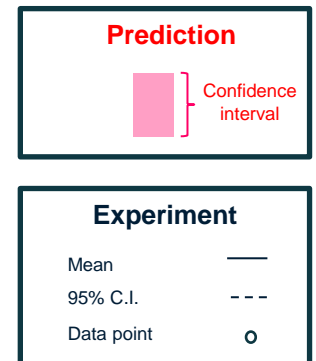
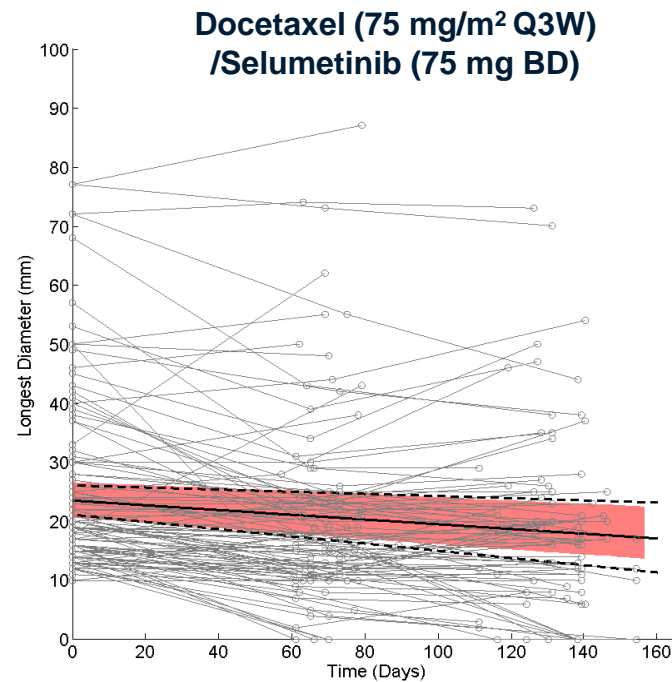
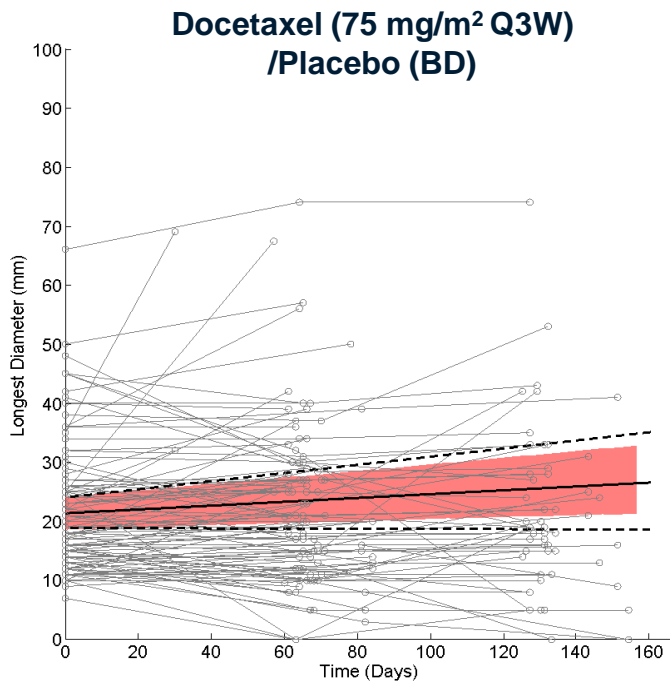


# Case Study: Translational Qualification

## Steps 4 and 5: Qualification



- ➔ Perform a population analysis of the clinical data and overlay the results
- ➔ Accurate predictions for both arms of the study
  - ➔ Final Study Result: overall response rate (ORR) 32% Doc/Mek v 14% Doc ( $p = 0.059$ )

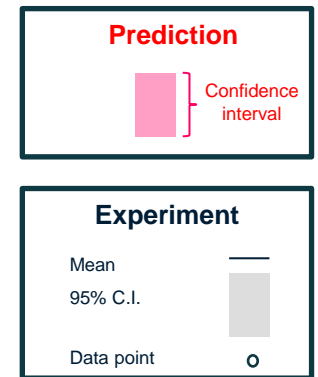
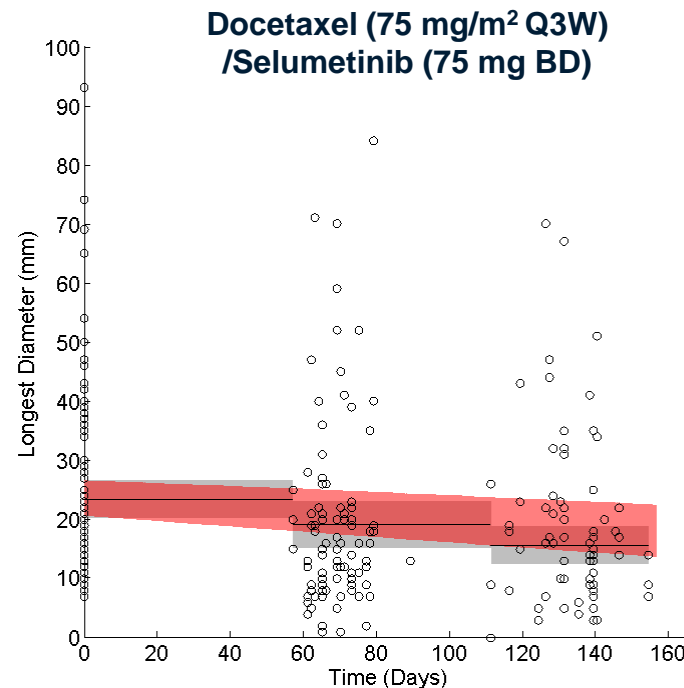
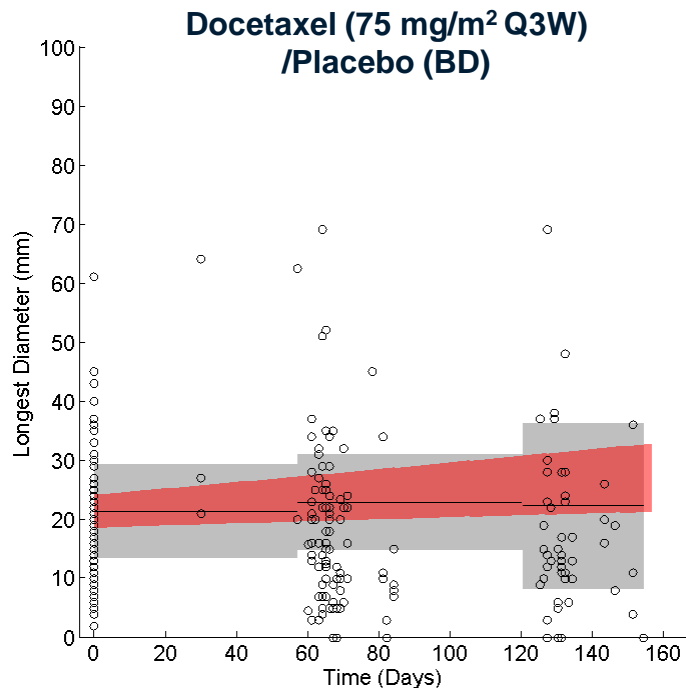


# Case Study: Translational Qualification

## Steps 4 and 5: Qualification



- ➔ **Biostatistics view: bin the data according to three groups and calculate the mean and 95% confidence interval**
- ➔ **Accurate predictions for both arms of the study**
  - ➔ Final Study Result: overall response rate (ORR) 32% Doc/Mek v 14% Doc ( $p = 0.059$ )







➔ Successfully predicted the results of the 2-arm clinical phase 2 trial using monotherapy preclinical efficacy data

➔ Performed further predictions for Oxford's ECMC to look at different regimens e.g.

➔ What happens if we alter the way Selumetinib is given in a day?

➔ Legend:

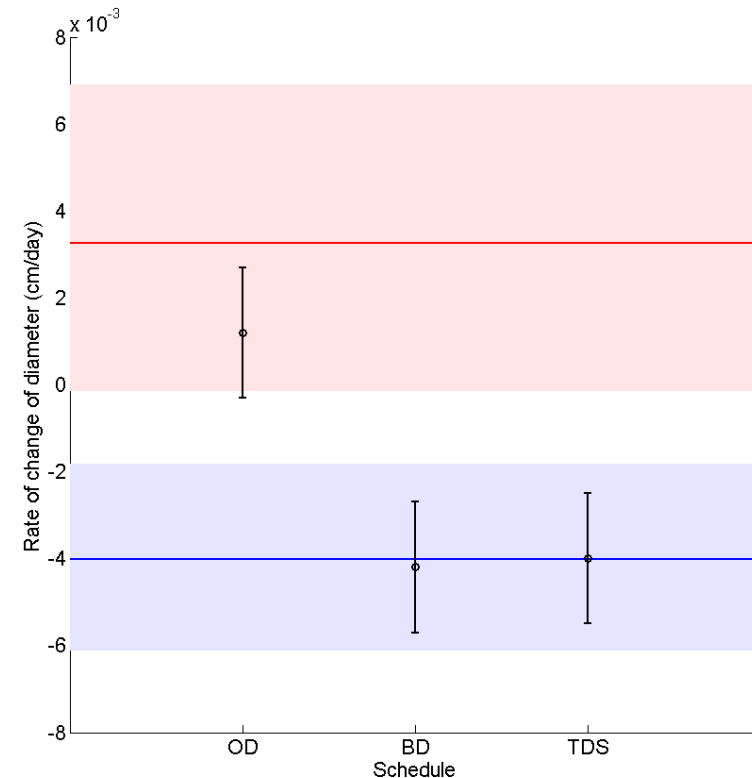
➔ Docetaxel ( $75 \text{ mg/m}^2$ ) mean (red) and 95% C.I. (pink region)

➔ Docetaxel ( $75 \text{ mg/m}^2$ )/Mek ( $75 \text{ mg BD}$ ) mean (blue) and 95% C.I. (light blue region)

➔ Model predictions open circles and C.I.

➔ Total daily dose is 150 mg

➔ No difference between BD and TDS for the same total daily dose.





- ➔ **Successfully predicted the mean change in lesion size for each arm of the phase II trial, using monotherapy preclinical efficacy data and clinical PK data**
  - ➔ Performed further predictions for Oxford ECMC, exploring different dosing regimens and changing docetaxel for paclitaxel
- ➔ **Virtual Tumour Clinical can provide significant cost-savings**
  - ➔ accurate translation of preclinical efficacy reduces the number of clinical studies required to find optimal doses and schedules
- ➔ **Virtual Tumour Clinical could reduce attrition rates**
  - ➔ Optimized regimens can enhance efficacy, increasing the chance of clinical trial success

***Dr Christophe Chassagnole: [cchassagnole@physiomics-plc.com](mailto:cchassagnole@physiomics-plc.com)***