

Modeling Drug Effects and Resistance Development on Tumor Growth Dynamics

PAGE

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1

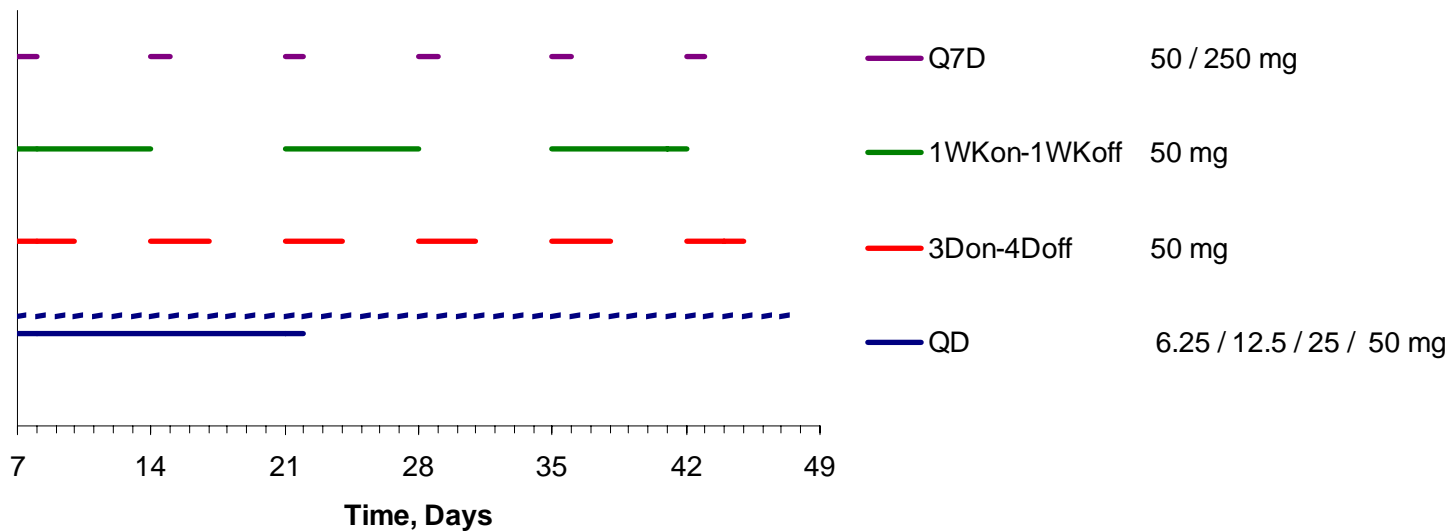


Outline

- Xenograft data used
- Simeoni and KPD model
- PKPD model implementation
- Results and simulations
- Conclusions

Tumor Xenograft data

- Overall sample size
 - ✦ Animals: 347 (nude mice)
 - ✦ Measurements: 1729
- Tumor inoculation on day 0 (U-87MG glioblastoma)
- First dose and observations at day 7 or day 9
- Oral Dosing Regimens:



Animal model used

- Animal model
 - ✦ Human **tumor xenograft** animal model:
 - ❖ human tumor lines
 - ❖ implanted in nude mice
 - ✦ Measure tumor weight/volume

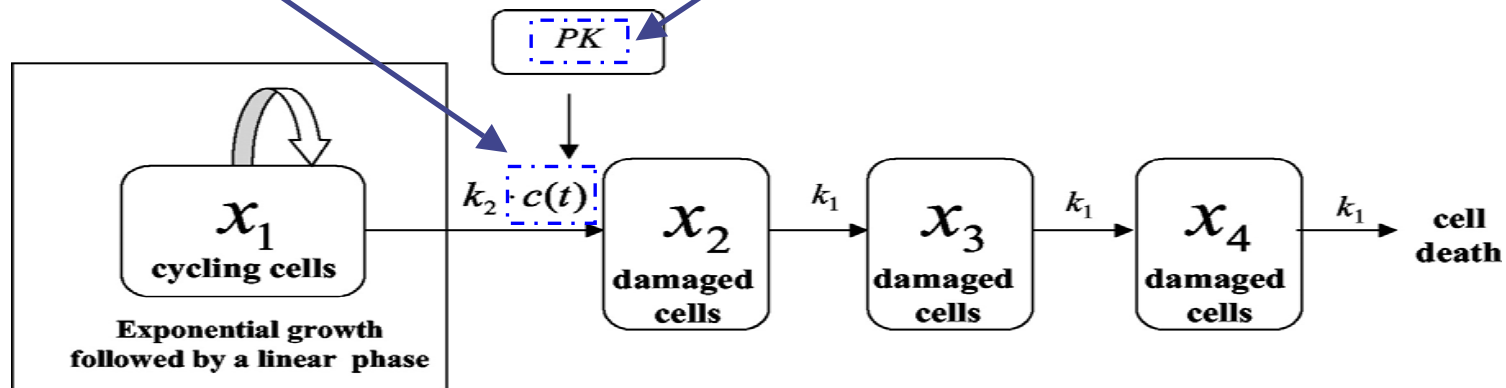


PKPD model used

- Model published by Simeoni.¹
 - ✦ Mechanistic model
 - ✦ Allows:
 - ❖ Re-use in other anti-tumor compounds
 - ❖ Link: dosing - tumor growth
 - ✦ Adaptations to the model

Resistance term

KPD model²

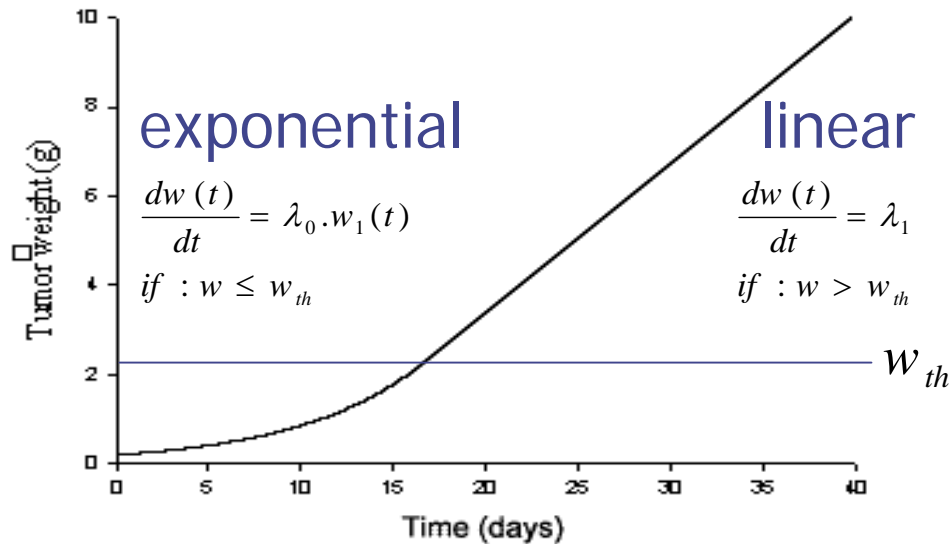
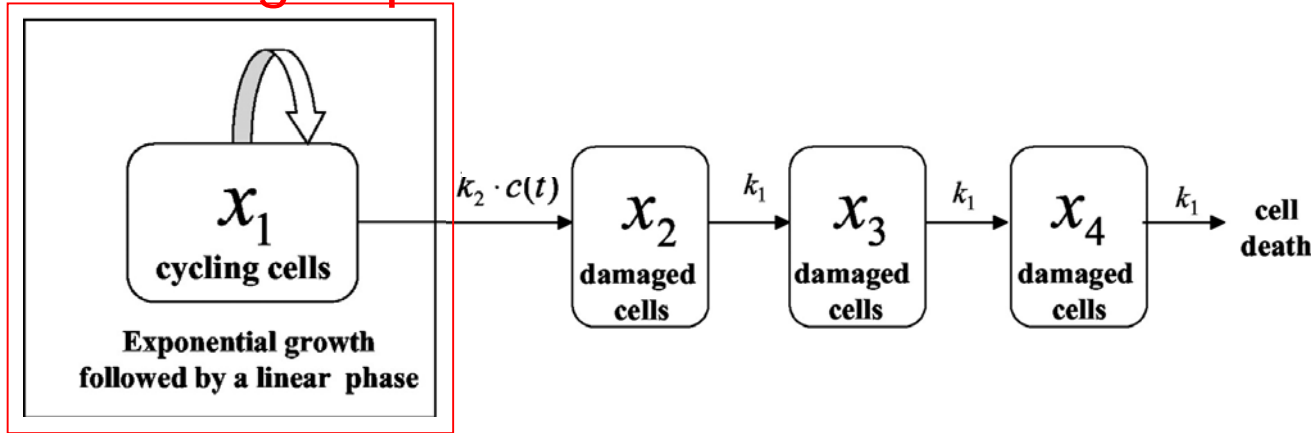


¹ Simeoni et al. CANCER RESEARCH, 2004

² Jacqmin et al. JPKPD, 2007

Simeoni model

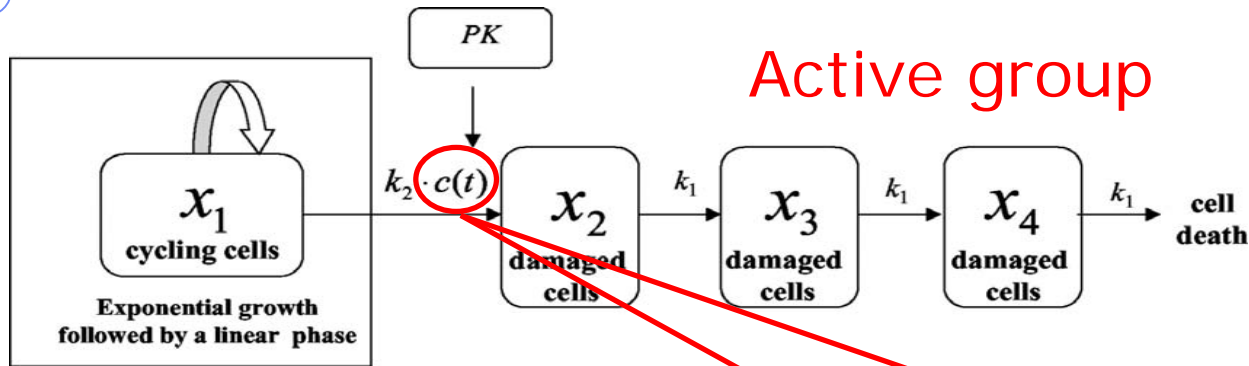
Control group



$$\frac{dw(t)}{dt} = \frac{\lambda_0 \cdot w(t)}{\left[1 + \left(\frac{\lambda_0}{\lambda_1} \cdot w(t)\right)^\Psi\right]^{1/\Psi}}$$

$$w(0) = w_0$$

Simeoni model (cont.)



$$\frac{dx_1(t)}{dt} = \frac{\lambda_0 \cdot x_1(t)}{1 + \left(\frac{\lambda_0}{\lambda_1} \cdot w(t)\right)^{\Psi}}^{1/\Psi} - k_2 \cdot c(t) \cdot x_1(t)$$

$$\frac{dx_2(t)}{dt} = k_2 \cdot c(t) \cdot x_1(t) - k_1 \cdot x_2(t)$$

$$\frac{dx_3(t)}{dt} = k_1 \cdot [x_2(t) - x_3(t)]$$

$$\frac{dx_4(t)}{dt} = k_1 \cdot [x_3(t) - x_4(t)]$$

• KPD model used

• Resistance equation

Simeoni endpoints

- At equilibrium:

$$\frac{\lambda_0 \cdot x_1(t)}{\left[1 + \left(\frac{\lambda_0}{\lambda_1} \cdot w(t)\right)^\psi\right]^{1/\psi}} - k_2 \cdot c(t) \cdot x_1(t) = 0$$

- **Threshold Concentration for Tumor Eradication (TCTE):**

$$TCTE = \frac{\lambda_0}{k_2}$$

- **Threshold Dose Rate for Tumor Eradication PKPD (TDRTE):**

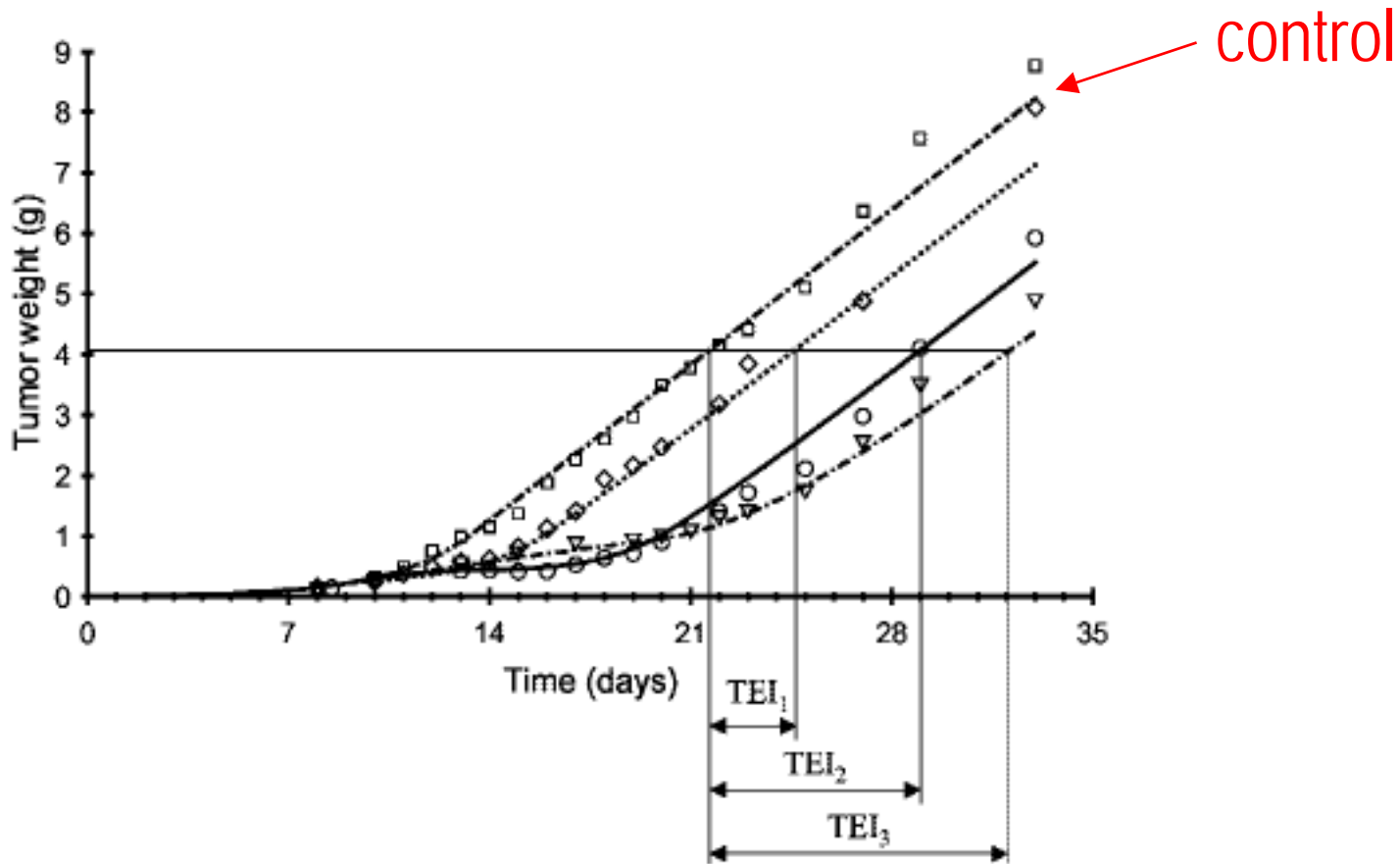
$$TDRTE = \frac{\lambda_0}{k_2} \cdot CL$$

- **Threshold Dose Rate for Tumor Eradication KPD (TDRTE):**

$$TDRTE = \frac{\lambda_0 \cdot k_{PD}}{k_2} \quad Kpd = CL/V$$

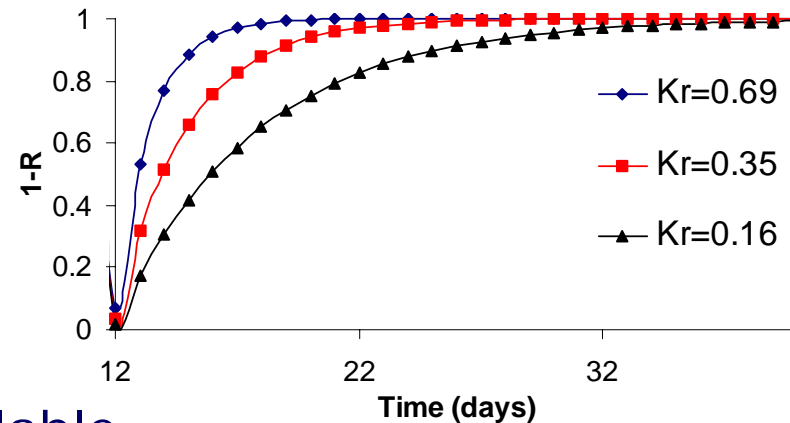
Simeoni endpoints (cont.)

Time Efficacy Index during linear growth:

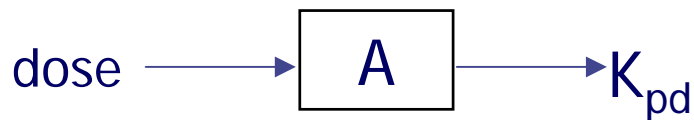


Implementation of the model

- Population approach of the model:
 - One model amongst the different dosing regimens
- Resistance term incorporated
 - If $t \leq t_{lag}$: no resistance
 - If $t > t_{lag}$: $R = \exp^{-k_r \cdot (t - t_{lag})}$

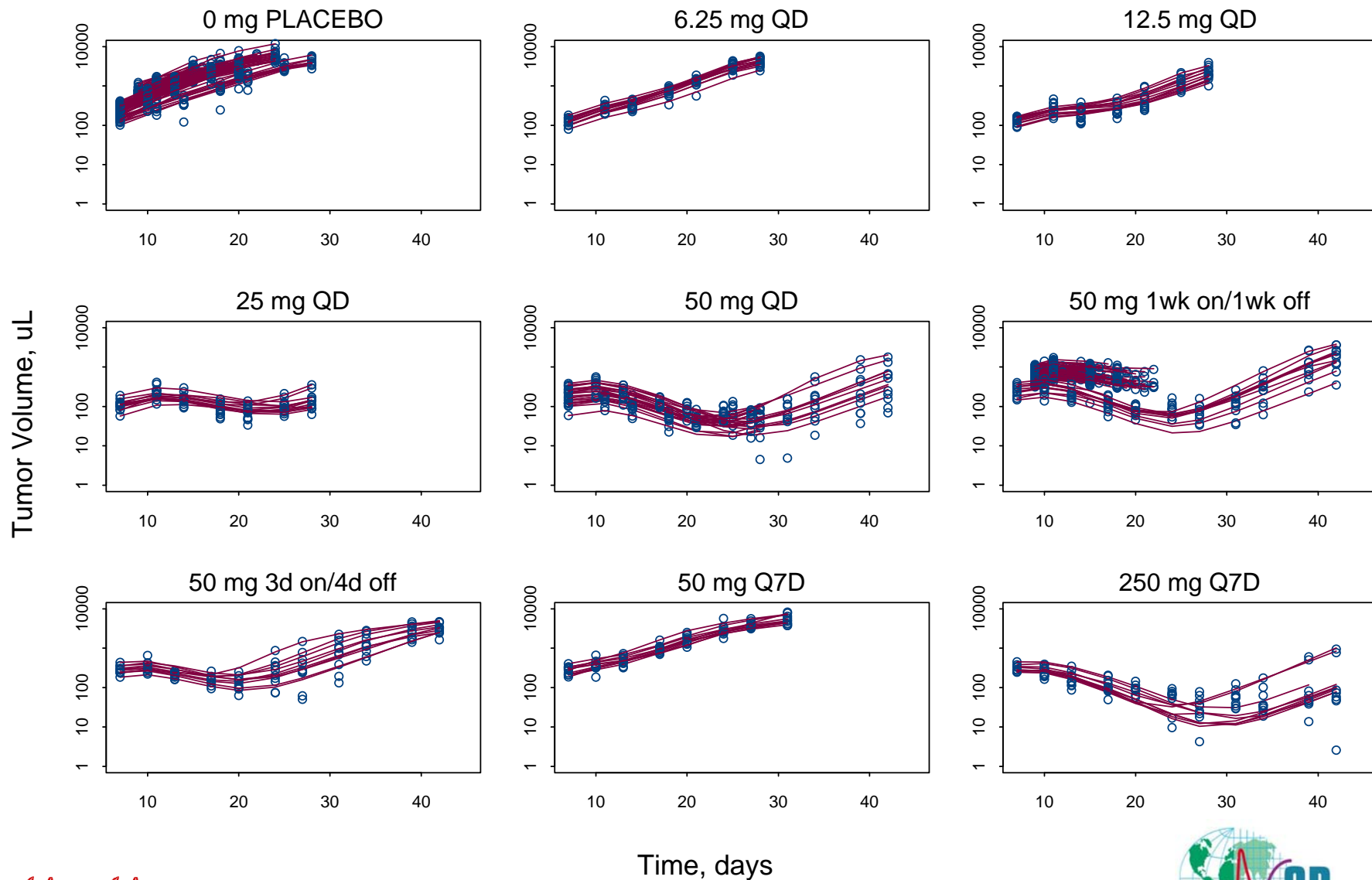


- KPD model
 - Very limited PK of mice available
 - PK in the rats non-linear



=> Growth rate decreased by: $K_2 \cdot K_{pd} \cdot R \cdot X(1)$

Individual predictions - oral dosing



Parameters estimates : U87GM-cells

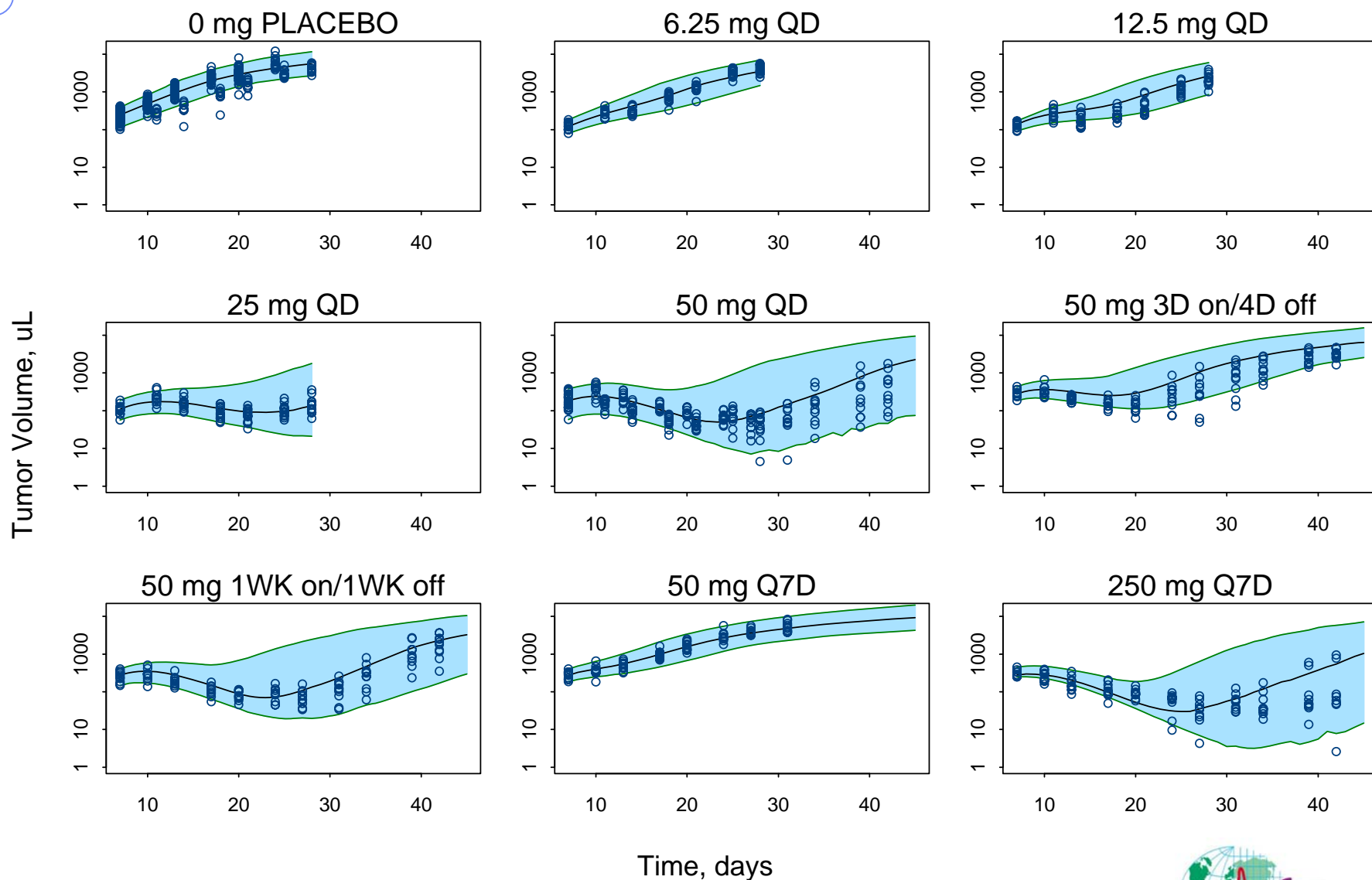
	Mean	SE	BAV
● λ_0 (day ⁻¹)	0.23	5%	17%
● λ_1 (μL.day ⁻¹)	323	2%	45%
● k_1 (day ⁻¹)	0.389	6%	9%
● k_2 (day ⁻¹ .mg ⁻¹)	0.021	10%	/
● k_{PD} (day ⁻¹)	0.993	14%	90%
● T_{Lag} (day)	11.9	3%	/
● k_R (day ⁻¹)	0.78	54%	52%

- **Threshold Daily Dose for Tumor Eradication:**
11 mg (95% CI: 9 - 12 mg)
- **Average time for a cell to die (n/k1):**
8 days (n the number of damage stages)

BAV: Between Animal Variability

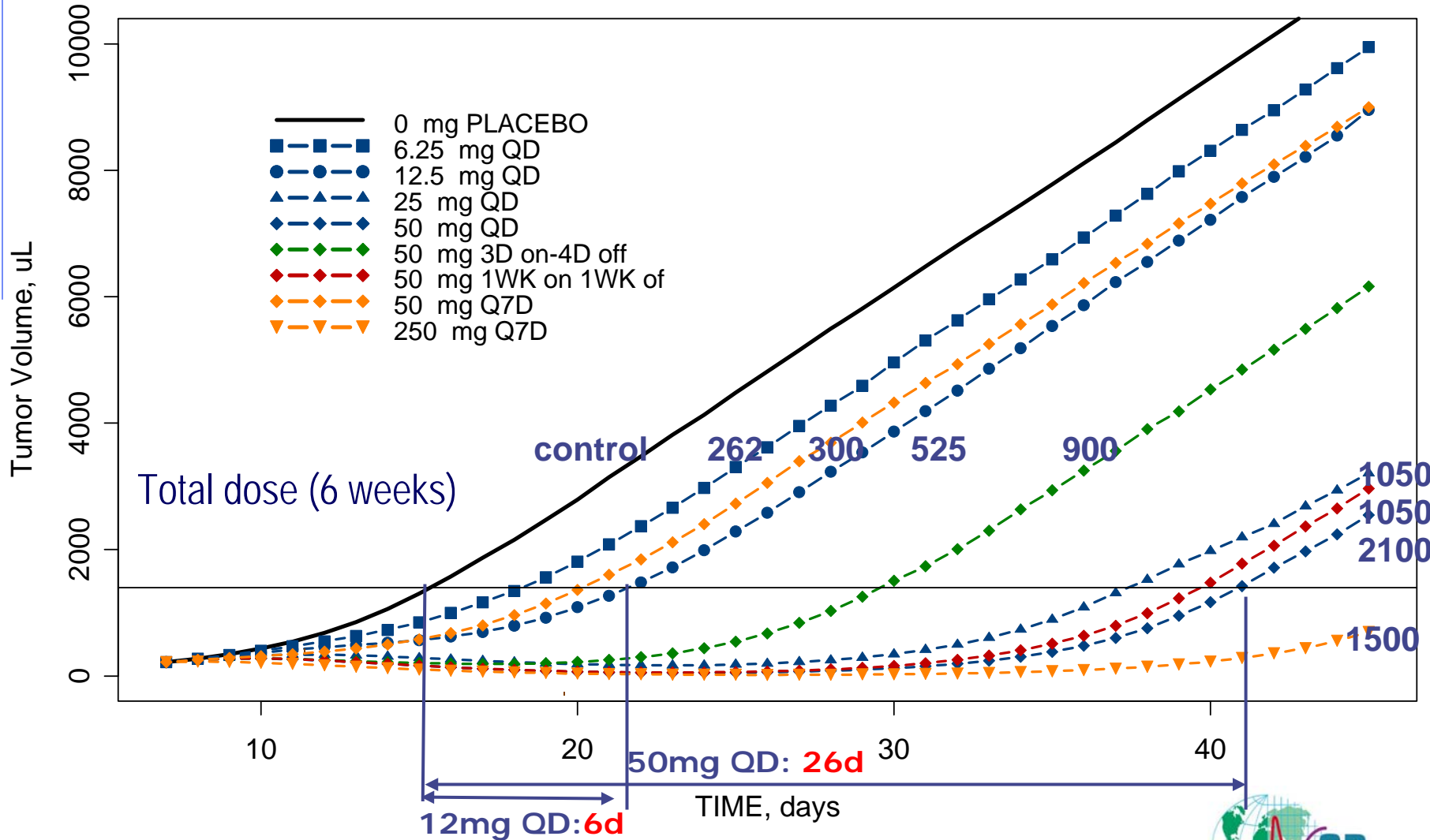
SE: Standard Error on the estimate

Visual predictive check



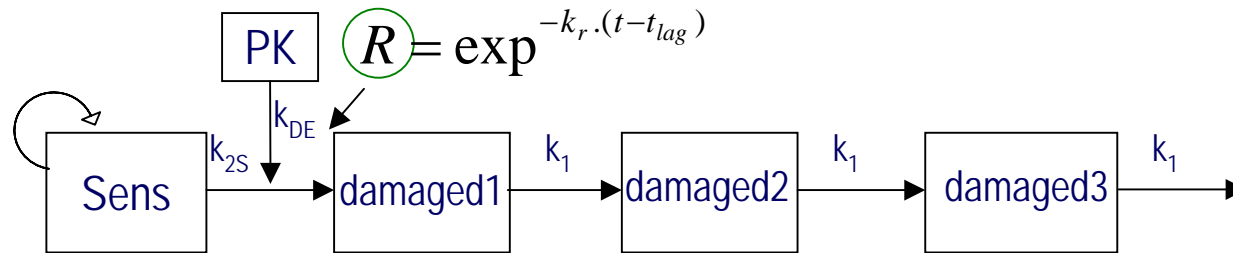
Time-efficacy index

Time Efficacy Index

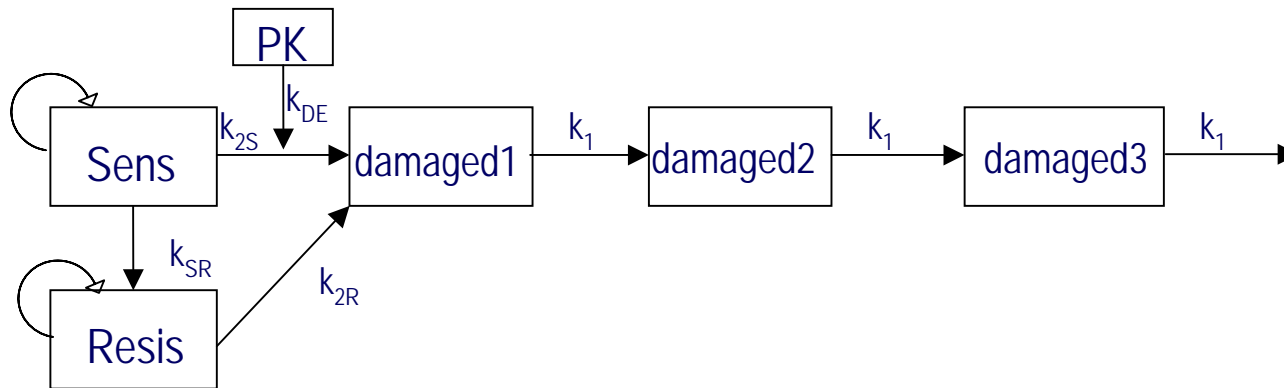


Comparison resistance models

- Empirical resistance



- Mechanistic resistance



Conclusions

- Data is very well described by the model
- KPD model can be used in absence of PK data
- Useful parameters can be derived:
 - ✦ Time efficacy index
 - ✦ Effective dose (if no resistance)
 - ✦ Time when resistance shows up
 - ✦ Time needed for a cell to die
- Model will be used to optimize future experiments