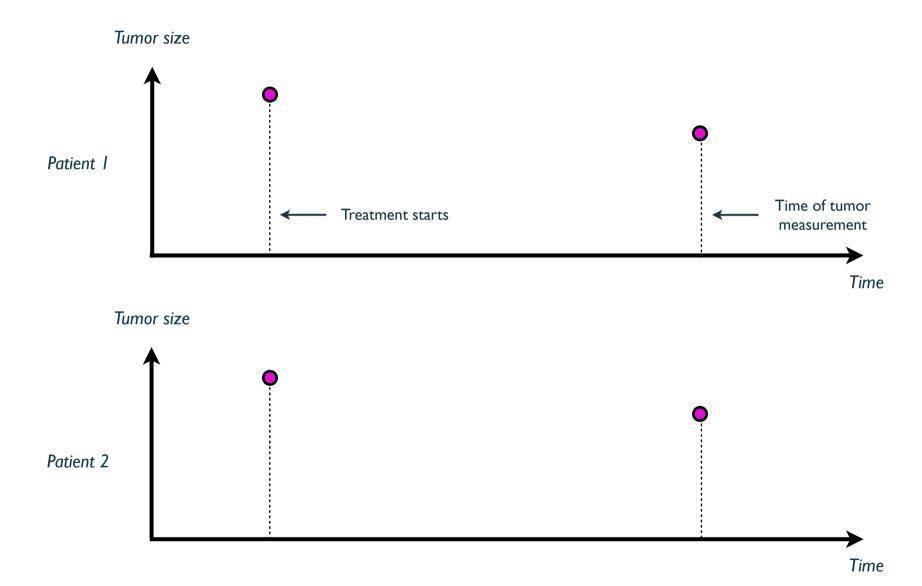
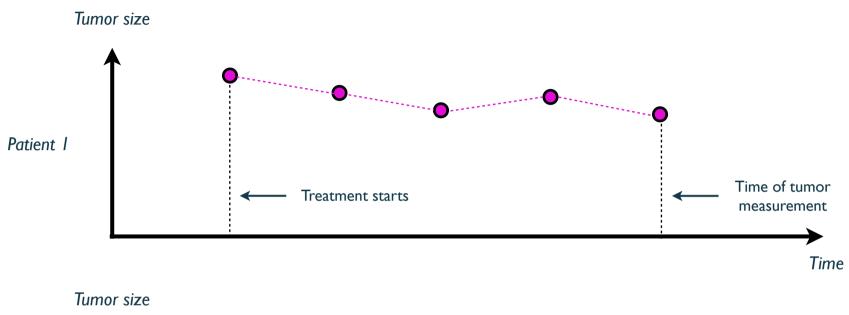
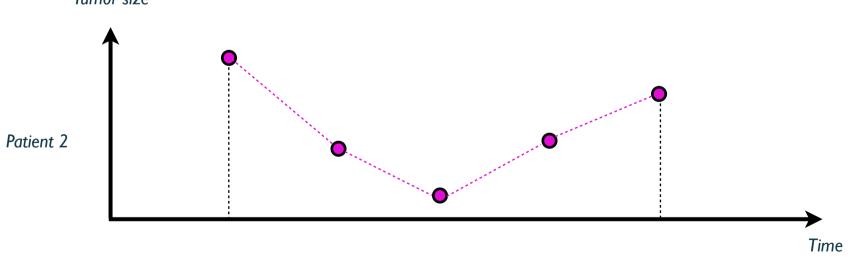


Modeling of efficacy data in clinical oncology

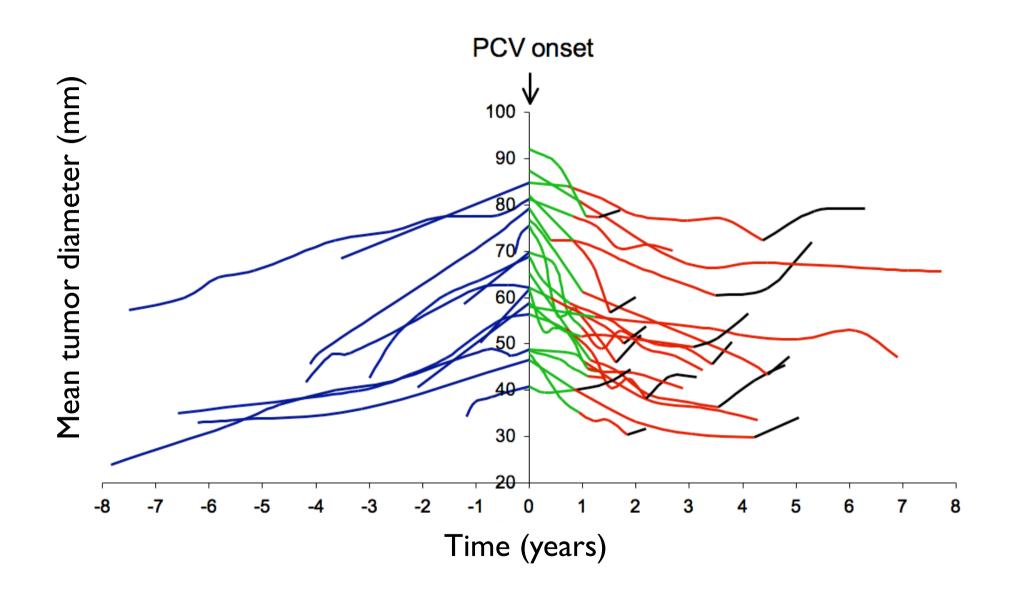
Benjamin Ribba

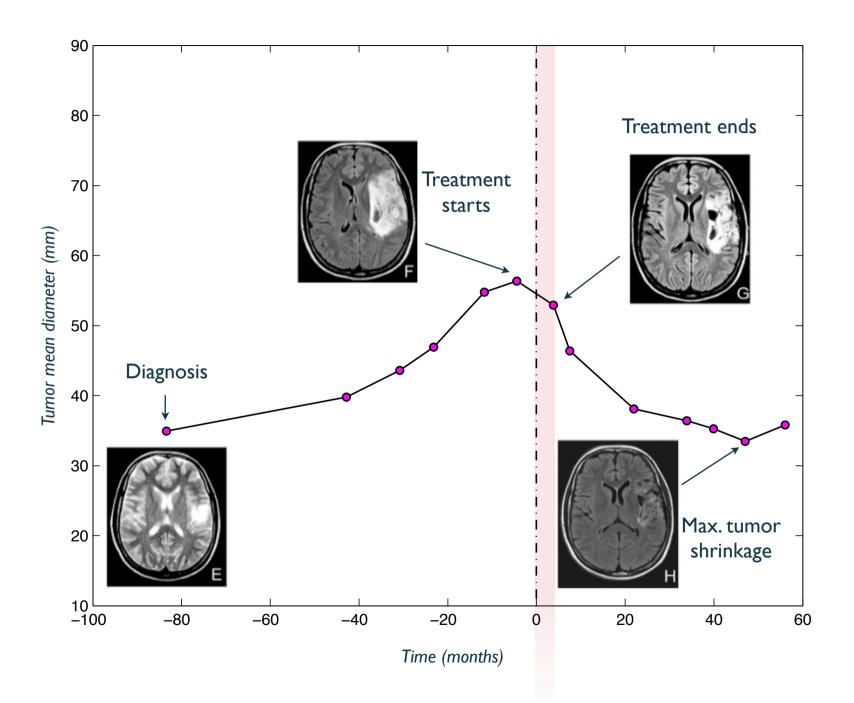






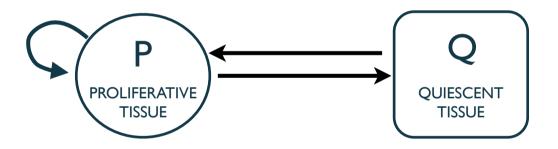
Models and tools to analyze such type of data?

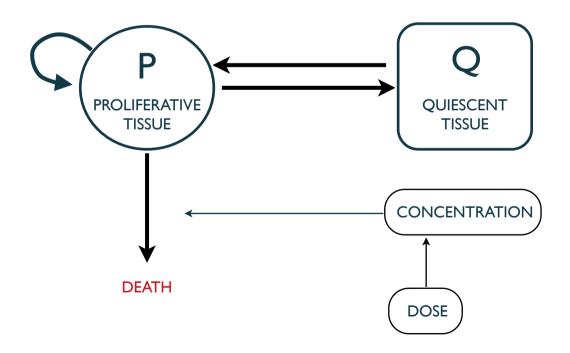


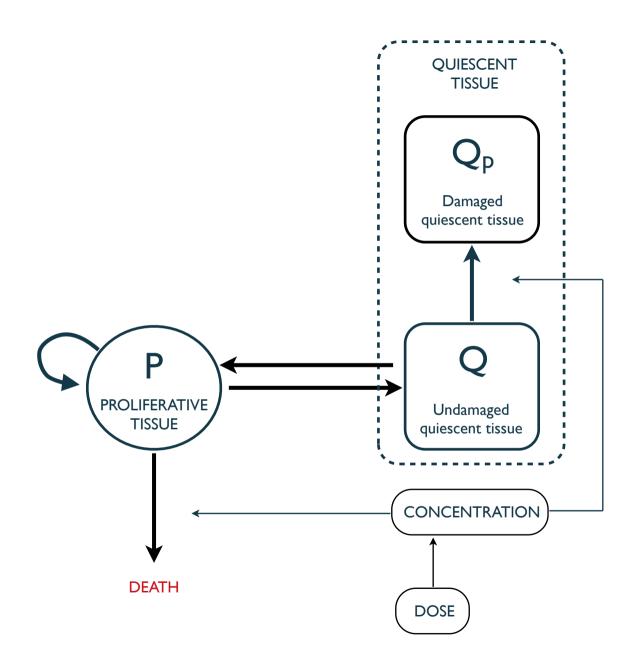


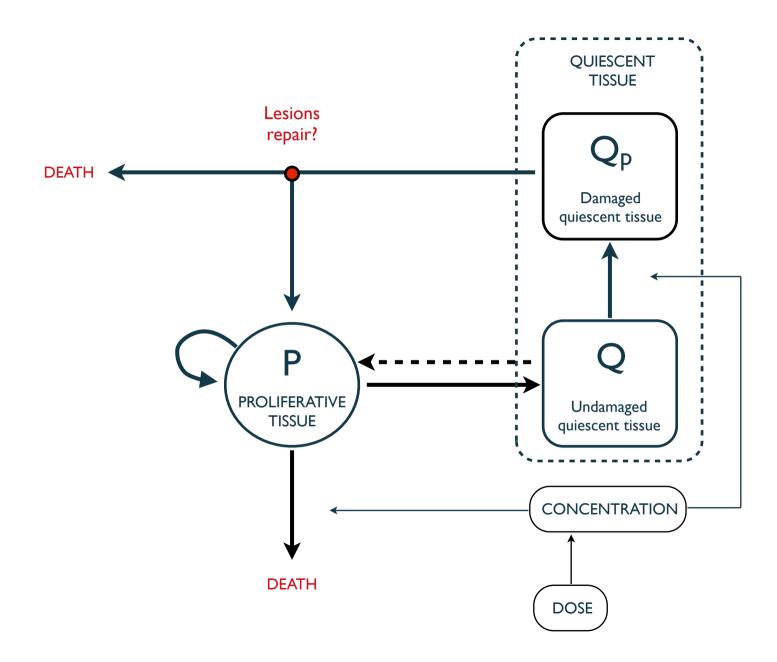
What we know from biology

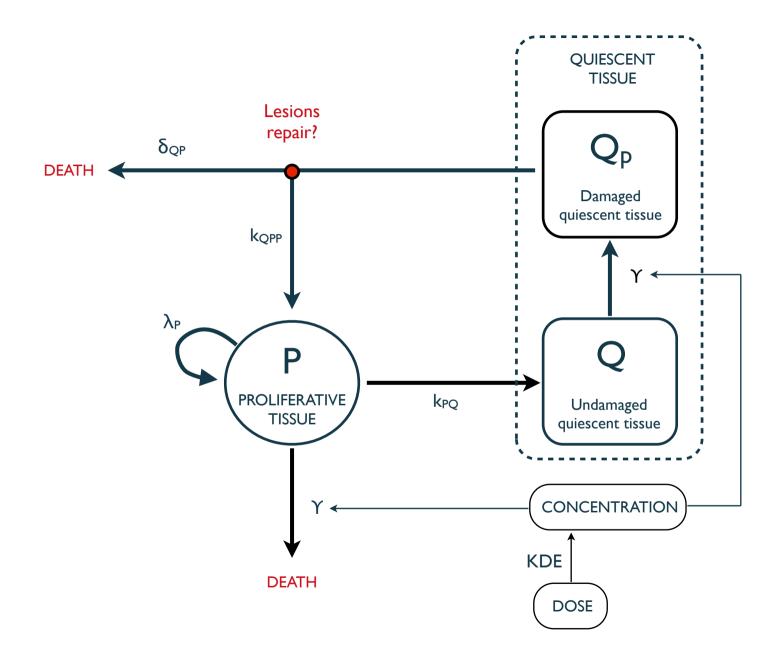
- The tumor is composed by two main types of tissues: proliferative and quiescent
- Drugs induce direct kill of proliferating cells
- Quiescent cells that have sustained DNA damages die when re-entering the cell cycle











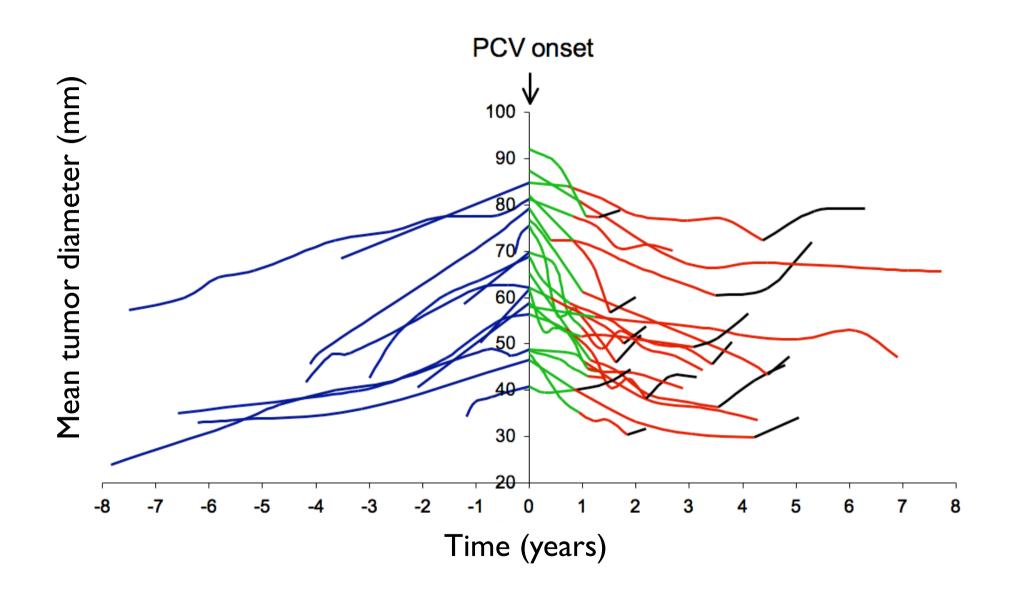
$$\frac{dP}{dt} = \lambda_P \times P\left(1 - \frac{P^*}{K}\right) - k_{PQ} \times P - \gamma_P \times C \times P + k_{Q_PP} \times Q_P$$

$$\frac{dQ}{dt} = k_{PQ} \times P - \gamma_Q \times C \times Q$$

QUIESCENT TISSUE

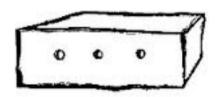
$$rac{dQ_P}{dt} = \gamma_Q imes C imes Q - k_{Q_PP} imes Q_P - \delta_{Q_P} imes Q_P$$
 Quiescent tissue

$$P^{\star} = P + Q + Q_P$$
 TUMOR SIZE

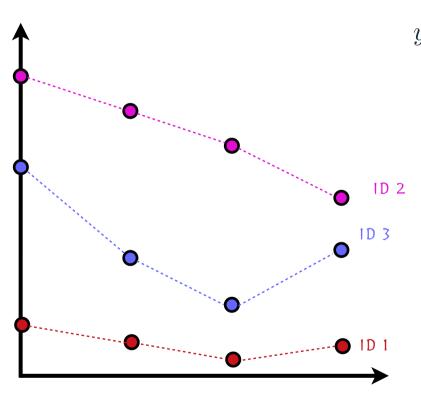


If you, please, draw me a sheep





Mixed-effect models



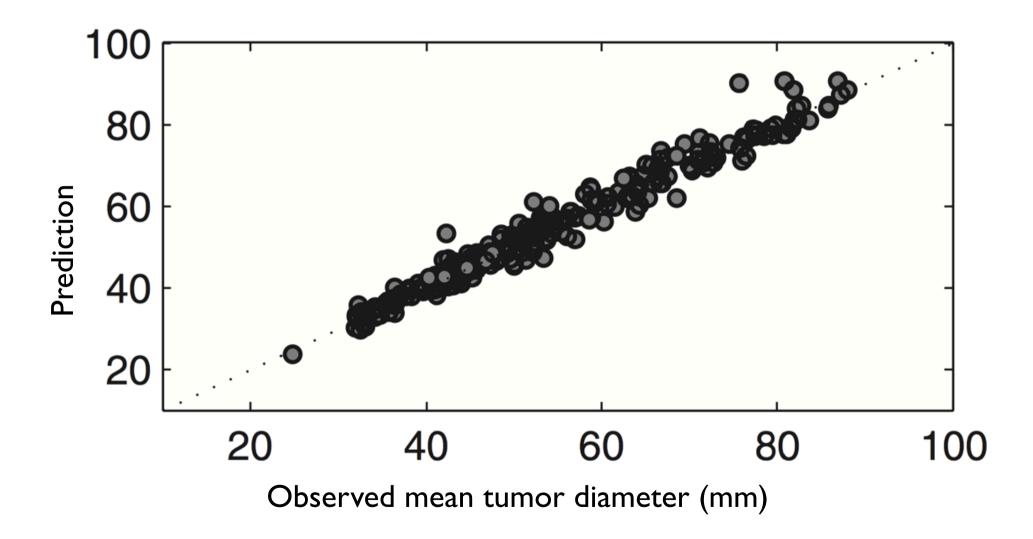
$$y_{ij} = f(t_{ij}, \boldsymbol{\psi_i}) + e_{ij} \quad 1 \le i \le N \quad 1 \le j \le n_i$$

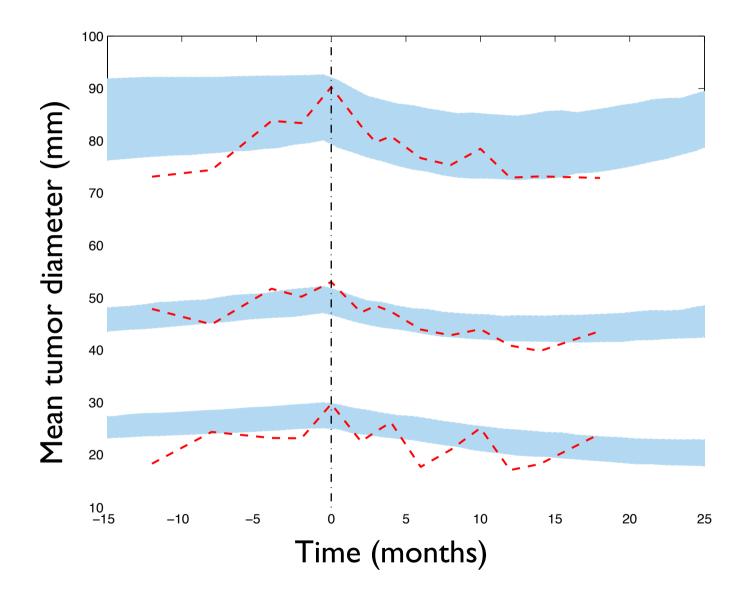
$$\boldsymbol{\psi_i} \sim \mathcal{N}(\theta_m, \theta_\sigma) \qquad e_{ij} \sim \mathcal{N}(0, a^2)$$

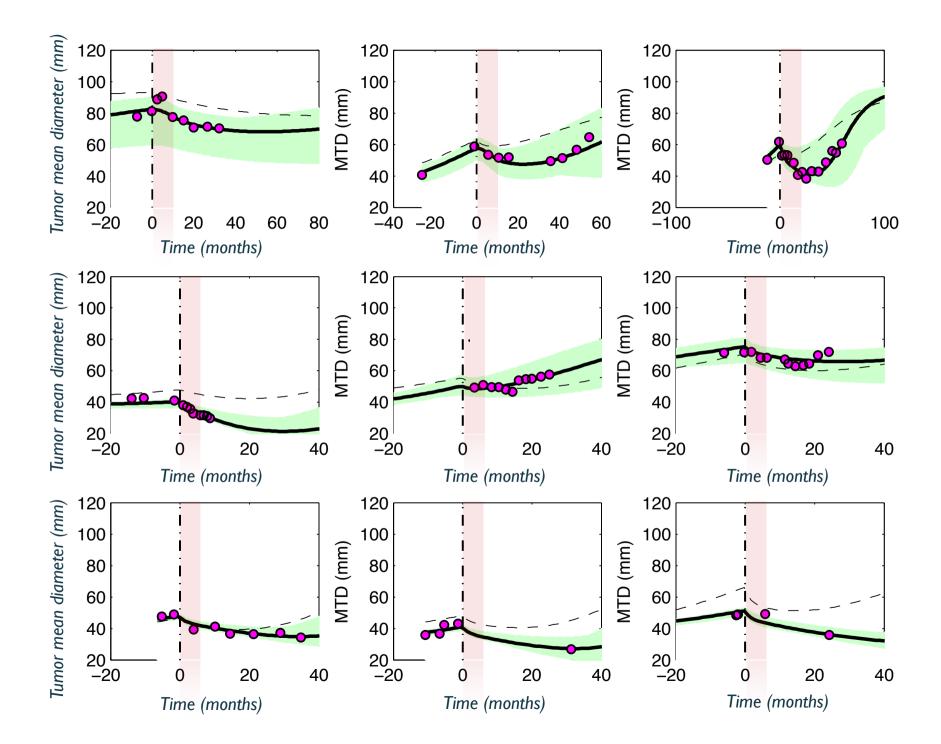
$$p(\psi_i) = \frac{1}{\sqrt{2\pi}\sqrt{\det(\theta_\sigma)}} e^{-\frac{1}{2}(\psi_i - \theta_m)'\theta_\sigma^{-1}(\psi_i - \theta_m)}$$

$$p_{y_i|\psi_i} = \prod_{j=1}^{n_i} p_{y_{ij}|\psi_i}$$

 $p(y_i, \psi_i) = p(y_i \mid \psi_i) \times p(\psi_i)$







Conclusions

- Importance of repeated measurements in oncology
- ODE model to correctly reproduce tumor size dynamic in brain tumor patients
- Mixed-effect regression techniques for population analysis
- Model simulation to suggest clinicians with new therapeutic hypothesis

Ribba et al. A tumor growth inhibition model for low-grade glioma treated with chemotherapy or radiotherapy. Clinical Cancer Research, 2012