### **Supplementary Figures**

Supplementary figure 1: plot of all tumor volumes for double and single (control) tumor groups Supplementary figure 2: single tumor model fit. Exponential model Supplementary figure 3: single tumor model fit. Gompertz model Supplementary figure 4: single tumor model fit. Power law model Supplementary figure 5: single tumor growth models: delay analysis Supplementary figure 6: double-tumors growth models fits. Competition model Supplementary figure 7: double-tumors growth models fits. Angiogenesis inhibition (SIA) Supplementary figure 8: double-tumors growth models fits.Proliferation inhibition. Log-kill effect Supplementary figure 9: double-tumors growth models fits. Proliferation inhibition.  $(P_i + Q_i)$  as source of IFs Supplementary figure 10: residuals analysis of the other models Supplementary figure 11: additional fits of 10 remaining animals of the control group under the "prolifer-

Supplementary figure 11: additional fits of 10 remaining animals of the control group under the "proliferation inhibition" model

Supplementary table 1: models equations

Supplementary table 2: goodness-of-fit metrics of these three single tumor growth models.

Supplementary table 3: parameter values and identifiability of the Gompertz, power law and "proliferation inhibition" model for the single tumor growth fits

### Supplementary figure 1: plot of all tumor volumes for double and single (control) tumor groups



Growth curves were plotted and distinguished between large and small tumors for the double (A) and single (control) (B) groups. In the second case, artificial and pairings were considered between tumors of the control group to determine large and small tumors that would have occurred by randomness only (i.e. under independent growth conditions). DT = double tumors group

#### Supplementary figure 2: single tumor model fit. Exponential model



### Supplementary figure 3: single tumor model fit. Gompertz model



### Supplementary figure 4: single tumor model fit. Power law model



#### Supplementary figure 5: single tumor growth models: delay analysis



Time  $t_D$  to reach a threshold volume  $V_D = 500 \text{ mm}^3$  for growth extrapolated from the fitted models. \* = p < 0.05, Student's t-test with unequal variance.



0 L 0

10 Days

15

20

#### Supplementary figure 6: double-tumors growth models fits. Competition model

### Supplementary figure 7: double-tumors growth models fits. Angiogenesis inhibition (SIA)



Dashed lines are simulations with no interactions between the two tumors, i.e. with the parameters inferred from the fits except for parameter e set to zero. Growth differences are only due to the difference in initial condition.

### Supplementary figure 8: double-tumors growth models fits.Proliferation inhibition. Log-kill effect



Dashed lines are simulations with no interactions between the two tumors. Growth differences are only due to the difference in initial condition.



Supplementary figure 9: double-tumors growth models fits. Proliferation inhibition.  $(P_i + Q_i)$  as source of IFs

Dashed lines are simulations with no interactions between the two tumors. Growth differences are only due to the difference in initial condition.

#### Supplementary figure 10: residuals analysis of the other models



Angiogenesis inhibition (SIA)



Proliferation inhibition. Log-kill



Proliferation inhibition.  $\left(P_i+Q_i\right)$  as source of IFs





# Supplementary figure 11: additional fits of 10 remaining animals of the control group under the "proliferation inhibition" model

### Supplementary table 1: models equations

Model nameEquationsCompetition
$$\begin{cases} \frac{dY_1}{dt} = aV_1 \ln \left(\frac{K}{V_1 + V_2}\right) \quad V_1(t = 0) = V_{0,1} \\ \frac{dY_2}{dt} = aV_2 \ln \left(\frac{K}{V_1} + V_2\right) \quad V_2(t = 0) = V_{0,2} \end{cases}$$
Angiogenesis inhibition
$$\begin{cases} \frac{dY_1}{dt} = aV_1 \ln \left(\frac{K_1}{V_1}\right) \quad V_1(t = 0) = V_{0,1} \\ \frac{dK_1}{dt} = bK_1 - dV_1^{2/3}K_1 - eV_2 \mathbbm{1}_{K_2 \times K_0} \quad K_1(t = 0) = K_0 \\ \frac{dY_2}{dt} = aV_2 \ln \left(\frac{K_2}{V_2}\right) \quad V_2(t = 0) = V_{0,2} \\ \frac{dY_3}{dt} = bK_2 - dV_2^{2/3}K_2 - eV_1 \mathbbm{1}_{K_2 \times K_0} \quad K_1(t = 0) = K_0 \end{cases}$$
Proliferation inhibition
$$\begin{cases} \frac{dP_1}{dt} = \alpha P_1 - (\beta P_1 + \gamma (P_1 + P_2)) \mathbbm{1}_{P_1 > 0} \quad P_1(t = 0) = V_{0,1} \\ \frac{dQ_2}{dt} = (\beta P_1 + \gamma (P_1 + P_2)) \mathbbm{1}_{P_1 > 0} \quad Q_1(t = 0) = 0 \\ V_1 = P_1 + Q_1 \\ \frac{dP_2}{dt} = (\beta P_2 + \gamma (P_1 + P_2)) \mathbbm{1}_{P_2 > 0} \quad Q_2(t = 0) = 0 \\ V_2 = P_2 + Q_2 \end{cases}$$
Proliferation inhibition (log-kill)
$$\begin{cases} \frac{dP_2}{dt} = \alpha P_1 - (\beta P_1 + \gamma (P_1 + P_2)) P_1 \quad P_1(t = 0) = V_{0,1} \\ \frac{dQ_2}{dt} = (\beta P_2 + \gamma (P_1 + P_2)) \mathbbm{1}_{P_2 > 0} \quad Q_2(t = 0) = 0 \\ V_2 = P_2 + Q_2 \end{cases}$$

Proliferation inhibition $(P+Q)$		$\begin{cases} \frac{dP_1}{dt} = \alpha P_1 - (\beta V_1 + \gamma (V_1 + V_2)) \mathbb{1}_{P_1 > 0} \\ \frac{dQ_1}{dt} = (\beta V_1 + \gamma (V_1 + V_2)) \mathbb{1}_{P_1 > 0} \\ V_1 = P_1 + Q_1 \end{cases}$	$P_1(t=0) = V_{0,1}$ $Q_1(t=0) = 0$
	<	$ \begin{aligned} \frac{dP_2}{dt} &= \alpha P_2 - (\beta V_2 + \gamma (V_1 + V_2))  \mathbb{1}_{P_2 > 0} \\ \frac{dQ_2}{dt} &= (\beta V_2 + \gamma (V_1 + V_2))  \mathbb{1}_{P_2 > 0} \\ V_2 &= P_2 + Q_2 \end{aligned} $	$P_2(t=0) = V_{0,2}$ $Q_2(t=0) = 0$

Supplementary table 2: goodness-of-fit metrics of these three single tumor growth models.

Model	SSE	AIC	RMSE	R2	р
Power law	0.117(0.0158 - 0.713)[1]	-12.3(-34.5 - 2.95)[1]	0.4(0.145 - 0.957)[2]	0.983(0.784 - 0.998)[2]	
Gompertz	0.121(0.019 - 0.67)[2]	-11.6(-32.4 - 2.39)[2]	0.394(0.159 - 0.928)[1]	0.984(0.815 - 0.997)[1]	
Proliferation inhibition	0.159(0.00741 - 0.883)[3]	-10.1(-33.2 - 4.88)[3]	0.45(0.0994 - 1.07)[3]	0.966(0.7 - 0.999)[3]	

SSE = Sum of Square Errors, AIC = Akaike Information Criterion, BIC = Bayesian Information Criterion, R2 = coefficient of determination. # = number of parameters.

# Supplementary table 3: parameter values and identifiability of the Gompertz, power law and "proliferation inhibition" model for the single tumor growth fits

Model	Par.	Unit	Median value (CV)	NSE (%) (CV)
Power law	$lpha \gamma$	$\mathrm{mm}^{3(1-\gamma)}\cdot\mathrm{day}^{-1}$ -	0.921 (41.9) 0.788 (9.35)	10.6 (55) 3.42 (62.4)
Gompertz	$lpha_0\ eta$	${\sf day}^{-1}$ ${\sf day}^{-1}$	1.84 (35.7) 0.0792 (43)	9.28 (65.3) 12 (74.4)
Proliferation inhibition	$\begin{array}{c} \alpha \\ \beta + \gamma \end{array}$	$day^{-1}$ $day^{-1}$	3.71 (54.4) 3.45 (59)	21.4 (61.7) 24.5 (74.5)

Par. = parameter. CV = coefficient of variation. NSE = normalized standard error.