Concomitant administration of bevacizumab and pemetrexed-cisplatin is a common treatment for advanced nonsquamous non-small cell lung cancer (NSCLC). Vascular normalization following bevacizumab administration may transiently enhance drug delivery, suggesting improved efficacy with sequential administration. To investigate optimal scheduling, we conducted a study in NSCLC-bearing mice. First, experiments demonstrated improved efficacy when using sequential vs. concomitant scheduling of bevacizumab and chemotherapy. Combining this data with a mathematical model of tumor growth under therapy accounting for the normalization effect, we predicted an optimal delay of 2.8 days between bevacizumab and chemotherapy. This prediction was confirmed experimentally, with reduced tumor growth of 38% as compared to concomitant scheduling, and prolonged survival (74 vs. 70 days). Alternate sequencing of 8 days failed in achieving a similar increase in efficacy, thus emphasizing the utility of modeling support to identify optimal scheduling. The model could also be a useful tool in the clinic to personally tailor regimen sequences.

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chemotherapy in patients with locally advanced rectal cancer led to better OS and progression-free survival compared to concomitant administration. Although promising, most attempts to revisit bevacizumab scheduling have been made on an empirical basis and a trial-and-error mode. Mathematical modeling could help assisting experimental and clinical studies to better understand the combined effects of anti-angiogenics and cytotoxics on tumor growth. Building on the Hahnfeldt et al. model for the effect of anti-angiogenic monotherapy, Wilson et al. developed a model to quantify the dynamics of interaction between tumor growth, vasculature generation, and anti-angiogenic treatment. They demonstrated a possible synergistic interaction between sunitinib and irinotecan. Recently, Hutchinson et al. developed a model of vascular tumor growth and normalization from breast cancer mice treated with bevacizumab alone. Still in breast cancer, our team tested multiple regimens for combining bevacizumab and paclitaxel. Experimental results combined with mathematical simulations suggested that scheduling bevacizumab 2 days before paclitaxel could improve antitumor efficacy and reduce metastatic spreading.

The aim of the present modeling and experimental study was to assess predictions of a semimechanistic mathematical model in terms of the optimal sequence of administration for the combination of bevacizumab and pemetrexed-cisplatin in non-small cell lung cancer (NSCLC).

MATERIALS AND METHODS

Mice experiments

Cell lines. Human NSCLC cells H460 stably transfected with luciferase (H460 Luc+) were purchased from Perkin Elmer, France. This BioWare light producing cell line was derived from the H460 human adenocarcinoma by stable transfection of the North American firefly gene expressed from the SV40 promoter. Cells were cultured per the manufacturer’s recommendation at 37°C in a humidified atmosphere with 5% CO2. H460 Luc+ cells were regularly authenticated based on viability, growth, morphology, and in vitro bioluminescence measuring.

For experiment two, tumor growth monitoring was carried out by fluorescence. Therefore, H460 Luc+ cells were transfected with tdTomato gene using pPGK-tdTomato lentiviral plasmid lentiviral plasmid as vector kindly provided by Dr. Valérie Le Morvan (Institut Bergonié, Bordeaux, France). Forty-eight hours after transfection, blasticidin (10 μg/mL) was added and selection was maintained for 2 weeks. Resulting H460 Luc+ tdTomato+ cells were regularly authenticated by microscopy based on viability, growth, morphology, and in vitro fluorescence monitoring.

Animal experiments

All experiments were approved by the local ethical committee of our institution and registered as #2015110616255292 (French Ministère de l’Education Nationale, de l’Enseignement Supérieur et de la Recherche) prior to starting the experiment. Guidelines for animal welfare in experimental oncology as recommended by European regulations were followed. Pathogen-free, immunocompromised 6-week-old female Swiss nude mice (Charles River Laboratories, France) were kept in a sterile environment for 2 weeks upon arrival. Mice were maintained in sterilized filter-topped cages and a sterile thermostatic cabinet throughout the study. Signs of distress, decreased physical activity, and any behavioral change were monitored daily. Bodyweights were monitored twice weekly as a surrogate marker for general toxicity. Water was supplemented with paracetamol (eq. 80 mg/kg/day) to prevent any disease-related pain. Animals showing signs of distress, pain, cachexia (i.e., loss of 10% of body weight), and tumor mass over 2 g (i.e., ~2 cm³) were euthanized.

Xenografting

H460 Luc+ (and tdTomato+ for the second experiment) cells were trypsinized, counted, centrifuged (5 minutes, 1,000 g) and washed twice with sterile phosphate-buffered saline. Cells were resuspended in Roswell Park Memorial Institute-1640 with 60% Matrigel (BD Sciences, France) and maintained in ice-cooled conditions until engraftment. A volume of 50 μL containing 80,000 cells (experiment one) and 120,000 cells (experiment two) was injected ectopically in the left flank of each mouse while under anesthesia. In total, 132 tumor-bearing mice were required to perform both experiments. However, overall, 139 mice were initially xenografted, to ensure that eventually at least 48 (experiment one) + 84 mice (experiment two) presenting positive and measurable tumors could be used, considering an estimated 5% of failure rate during the grafting procedure.

Bioluminescence imaging

In experiment one, monitoring of primary tumor growth started 1 week after engraftment. Acquisitions started 12 minutes after firefly D-Luciferin (Perkin Elmer, 300 mg/kg) i.p. injection to reach a plateau in bioluminescence signal. Acquisition and data processing were performed using the IVIS spectrum imager equipped with the Living Image 4.2 software (Perkin Elmer). Imaging was performed twice per week. All imaging was performed in anesthetized animals (sevoflurane).

Fluorescence imaging

In experiment two, imaging was performed twice in anesthetized animals (sevoflurane). Acquisition (excitation: 554 nm, emission: 581 nm), and data processing were performed using the IVIS spectrum imager equipped with the Living Image 4.2 software.

Experiment one treatments

In experiment one, 47 xenografted mice were randomized into 4 treatment arms (Supplementary Figure S1a): control (saline injection, n = 12), sequential treatment with bevacizumab administered 4 days before pemetrexed-cisplatin (“beva then chemo 4 days,” n = 12), sequential treatment arm with bevacizumab administered 4 days after pemetrexed-cisplatin (“chemo then beva 4 days,” n = 11), and a concomitant treatment arm (“beva + chemo,” n = 12). Treatment started 17 days after xenograft. Doses in each arm were 20 mg · kg⁻¹, 100 mg · kg⁻¹ and 3 mg · kg⁻¹ for bevacizumab, pemetrexed, and cisplatin, respectively. All treatments, including saline, were administered by i.p. route for three 14-day cycle. All animals were euthanized on day 76 post-xenografts.
Experiment two treatments

The second experiment was performed with model-based changes both in scheduling and sample sizes. In experiment two, 77 xenografted mice were randomized into 5 treatment arms (Supplementary Figure S1b): control (saline injection, n = 15), sequential treatment with bevacizumab administered 3 days (see next section below) before pemetrexed-cisplatin (“beva then chemo 3 days,” n = 16), sequential treatment with bevacizumab administered 8 days before pemetrexed-cisplatin (“beva then chemo 8 days,” n = 15), concomitant (“beva + chemo,” n = 15), and pemetrexed and cisplatin alone (chemo, n = 15). The administered doses in each arm were 20 mg · kg⁻¹, 100 mg · kg⁻¹ and 3 mg · kg⁻¹ for bevacizumab, pemetrexed, and cisplatin, respectively. All treatments were administered by i.p. route. As for experiment one, 3 cycles (14 days cycle) were administered, starting on day 14 after xenograft. All animals were euthanized at the end of the experiment on day 87 post-xenografts.

PHARMACOKINETIC/PHARMACODYNAMIC MODELING

Structural model

The tumor size at time t is denoted V(t). The function C(t) combines plasma concentration of pemetrexed and cisplatin \( C(t) = C_{pm}(t) + C_{cis}(t) \). The function A(t) represents the plasma concentration of bevacizumab. Pharmacokinetic time courses for bevacizumab, cisplatin, and pemetrexed were modeled each by a one-compartment model with absorption compartment and pharmacokinetic parameters from the literature.27–29 Supplementary Figure S2a depicts the concentration profiles for each treatment. A scheme describing the pharmacodynamic model is given in Figure 1. It is based on the following hypotheses:

- (H1) Without treatment the tumor size kinetics follow a Gompertzian growth governed by parameters α (proliferation rate of the tumor cells) and β (rate of exponential decrease of the tumor relative growth rate).30
- (H2) Cytotoxics act by targeting a fraction of the tumor size (log-kill effect).31 This effect is driven by the parameter γ.

![Figure 1 Scheme of the structural mathematical model.](image)

This scheme depicts the concentration profiles for each treatment. A structural model above depends on six parameters \( a, b, \beta, \gamma, \delta, \tau \). The dynamics of this improvement is controlled by a parameter \( \delta \). The above assumptions are translated into the following system of nonlinear ordinary differential equations:

\[
\begin{align*}
\frac{dV}{dt} &= (\alpha - \beta \ln \left( \frac{V}{V_0} \right)) V - \gamma QC\times V(t=0) = V_0 \\
Q(t) &= 1 + \delta A(t-\tau) \\
\frac{dZ_1}{dt} &= \gamma QC\times - kZ_1 \\
Z_1(t=0) &= 0 \\
\frac{dZ_2}{dt} &= k(Z_1 - Z_2) \\
Z_2(t=0) &= 0 \\
\frac{dZ_3}{dt} &= k(Z_2 - Z_3) \\
Z_3(t=0) &= 0 \\
N &= V + Z_1 + Z_2 + Z_3
\end{align*}
\]

The initial size \( V_0 \) was set to \( 7.04 \times 10^6 \) photons/second, considering that 80,000 cells were injected (experiment one) and a previously established conversion ratio of \( V_c = 1 \) cell ≈ 88 photons/second.34

Statistical model and parameters estimation

For a description of the inter-animal variability, we used the nonlinear mixed-effects statistical framework.36 It consists in assuming a distribution of the parameters within the animal population, taken here to be log-normal for each parameter. Importantly, these were the same for all treatment groups. The structural model above depends on six parameters \( \alpha, \beta, \gamma, \delta, \tau, k \). After an initial sensitivity analysis showing that not all of these parameters were identifiable from our dataset (Supplementary Methods), we reduced this to the four parameters \( \alpha, \beta, \gamma, \delta \). These parameters were then estimated by adapting the result of likelihood maximization performed with the Monolix software (Lixoft, version 2016R1) using visual assessment of the goodness-of-fit (visual predictive checks) and consideration of the root mean squared error. See Supplementary Methods and Supplementary Figures S3–S6 for details. Values of the resulting parameters are reported in Table 1. Model simulations were performed using Matlab software.

Statistical analysis

Statistical analyses were performed using R software 3.3.2 (R Core Team, 2016). Intergroup differences in tumor growth were tested by nonparametric Kruskal-Wallis tests, should the data not meet the assumptions for one-way analysis of variance. Further between-group comparisons...
were performed either by Dunn's multiple comparison tests when treatment groups were compared with the control group, or by Nemenyi post-hoc tests with Tukey approximation for pairwise multiple comparison between groups. Survival analysis was done using Kaplan-Meier analysis. Intergroup differences in survival were tested for significance by the log-rank tests. Owing to sample size, a $P \text{val} < 0.05$ was considered statistically significant.

RESULTS

Administering bevacizumab before pemetrexed + cisplatin improved efficacy and median of survival

Experiment one efficacy. Monitoring of tumor growth for experiment one is shown in Figure 2a. At the end of the treatment phase (day 53), mean tumor sizes (expressed in $10^9$ photons/second (p/s)) were 16.9 ± 3.7 (control), 19.2 ± 3.4 (“chemo then beva 4 days”), 15.4 ± 2.5 (“beva + chemo), and 6.9 ± 2.1 (“beva then chemo 4 days”). A statistical difference was found between the groups ($P = 0.012$). Further Dunn’s multiple comparison tests evidenced a significant difference between the sequential administration “beva then chemo 4 days” and control arms (59% tumor growth reduction, $P = 0.047$). Conversely, concomitant “beva + chemo” and reversed “chemo then beva 4 days” groups had modest effects on tumor growth inhibition (8% tumor reduction ($P = 0.98$) and 13% higher tumor size ($P = 1$) as compared with the control group, respectively). The sequential administration of “beva then chemo 4 days” also confirmed its superiority in efficacy when compared with other treatment groups (55% tumor reduction, almost reaching significance compared to “beva + chemo” ($P = 0.071$) and 64% tumor reduction compared to “chemo then beva 4 days” ($P = 0.0073$)). All mice in the control group had to be euthanized at day 53.

At study conclusion (day 67; i.e., 18 days after all treatments stopped), mean tumor growth were 19.4 ± 3.7 (“chemo then beva 4 days”), 15.7 ± 2.7 (“beva + chemo”), and 10.0 ± 2.0 (“beva then chemo 4 days”). The sequential administration in the “beva then chemo 4 days” arm had the lowest mean tumor size compared with the other remaining groups (36% lower than concomitant in the “beva + chemo” arm and 49% lower than reversed in the “chemo then beva 4 days” arm), although not statistically significant ($P = 0.42$ and 0.13, respectively).

Survival curves for experiment one are displayed in Figure 2b. The median survival times were 39 days (control), 49 days (“chemo then beva 4 days”), 55 days (“beva + chemo”), and 67 days (“beva then chemo 4 days”). A log-

| Parameters | Description | Units | Estimates | RSE* (%) | IAV% |
|------------|-------------|-------|-----------|----------|------|
| $\alpha$   | Proliferation rate | Day$^{-1}$ | 0.767 | 8 | 86.2 |
| $\beta$    | Exponential decay rate of the relative tumor growth rate (Gompertz model) | Day$^{-1}$ | 0.037 | 10 | 57.3 |
| $\gamma$   | Baseline effect of the chemotherapy | (mg/g)$^{-1}$·day$^{-1}$ | 1 | — | — |
| $\delta$   | Cytotoxics efficacy improvement following vascular normalization | (mg/mL)$^{-1}$ | 1,200 | 36 | 0 |
| $\tau$     | Delay parameter for dynamics of $Q$ | Day | 2 | 20 | 10 |
| $k$        | Delay of the tumor cell loss following chemotherapy | Day$^{-1}$ | 0.3 | — | — |
| $\Sigma$   | Exponential error parameter | — | 0.951 | 4 | — |

IAV, interanimal variabilities; RSE, relative standard error.

Values of the parameters corresponding to the adapted fit. See Supplementary Methods for details on the estimation procedure.

RSE is a measure of the precision of the parameter estimates, expressed as coefficient of variation (CV%).

The IAV is the standard deviation estimated using Monolix software.

Table 1 Population parameters and interanimal variabilities estimates for experiment one data

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Figure 2 Efficacy and Kaplan-Meier survival curves of experiment one. (a) Mean tumor growth curves for the four treatment arms of experiment one. Signs above curves indicate statistically significant difference with the control arm (Student’s $t$ test, $P < 0.05$). (b) Kaplan-Meier plot of the overall survival for the four treatment arms of experiment one.


**Figure 3** Visual predictive check for experiment one population analysis. (a–d) Visual predictive check plots. Circles: experimental data. Stars with broken lines: median data. Solid lines: tumor growth simulated curves using median parameter values, dashed lines: 95% intervals for interanimal variability, generated from the simulation of 1,000 virtual animals with parameters distributed according to the distribution estimated by the mixed-effects fit. Beva, bevacizumab; Chemo, chemotherapy.

The rank test showed a significant difference between all groups ($P < 0.0001$). Further log-rank tests showed that each treatment group was significantly different than the control arm ($P < 0.001$). Moreover, the sequential administration in the “beva then chemo 4 days” arm had greater survival median and was significantly different than concomitant in the “beva + chemo” ($P = 0.0485$) and reversed in the “chemo then beva 4 days” arms ($P = 0.0496$). Conversely, no significant difference was observed between the “beva + chemo” and the “chemo then beva 4 days” arms ($P = 0.631$).

**Mathematical modeling predicted an optimal time delay of 3 days between the administration of bevacizumab and pemetrexed + cisplatin**

The selected model is a modified version of the Gompertz model with a delay in the treatment effects and inclusion of a dynamic variable $Q$ accounting for the vasculature quality and, thus, the normalization effect. See the Materials and Methods section for a detailed description of the model equation, data fit, and parameters' estimation method. Population analysis yielded the median parameter and interanimal variability estimates reported in Table 1 with good relative standard errors. Goodness-of-fit was assessed by visual predictive check plots (Figure 3a–d), which demonstrated a good agreement between the model simulations and the experimental data (see residual analysis in Supplementary Figure S7). Individual simulations also demonstrated the ability of our model to reproduce tumor growth dynamics for each mouse (Supplementary Figure S8).

The model with parameters calibrated on the experimental data allowed us to perform simulations varying the time lag between the administrations of bevacizumab and the pemetrexed-cisplatin doublet. The criterion for quantification of efficacy was the area under the tumor growth curve. Delays ranging from 1–10 days were tested. Simulation results showed that a 2.8-days delay between bevacizumab and chemotherapy achieved greater reduction in tumor sizes, with a difference of 76.8% in tumor size as compared with concomitant scheduling (Figure 4a–c). Our quantification of the normalization dynamics also predicted that a delay of 8 days would perform substantially worse, with a difference of only 54.3% compared with concomitant administration (Figure 4c). Quantification of the interanimal variability of the model parameters using our population approach allowed to simulate the resulting interanimal variability of the optimal interdrug administration gap. The optimal gap ranged from 0–10 days with median of 2.8 days and standard deviation of 1.84 days (Figure 4d).
Validation of the optimal delay predicted by the model

To test the predictions from the mathematical model, we designed a new experiment that implemented the above-mentioned schedules (3 and 8 day lag for sequential administrations of “beva then chemo”).

Tumor growth for experiment two is shown in Figure 5a. At the end of the treatment phase (day 54), mean tumor size (expressed in p/s) were 12,567 ± 2,461 (control), 8,692 ± 543 (“beva then chemo 8 days”), 8,446 ± 1,253 (chemo), 7,486 ± 1,106 (“beva + chemo”), and 4,626 ± 868 (“beva then chemo 3 days”). A statistically significant difference between all arms was obtained (\( P < 0.0016 \)). Further, Dunn’s multiple comparison tests confirmed the superiority of the sequential administration “beva then chemo 3 days,” and showed a statistically significant difference in efficacy between the “beva then chemo 3 days” arm and the control arm (63% of tumor growth reduction (\( P = 0.002 \)). Other treatment sequences led to more modest effects on tumor growth as compared with the control arm. Furthermore, mean tumor size for the sequential administration in the “beva then chemo 3 days” arm was markedly lower and almost reached significant difference compared to all other treatment arms (i.e., 38% compared with concomitant “beva + chemo” (\( P = 0.072 \)), 47% tumor growth reduction as compared with “beva then chemo 8 days” (\( P < 0.001 \)), and 45% compared with the chemo group (\( P = 0.016 \)).

At the end of the experiment (day 67; i.e. 9 days after all treatment stopped), a statistically significant difference was found between the four remaining arms (\( P = 0.015 \)). Further, Dunn’s multiple comparison tests showed that “beva then chemo 8 days” was significantly different than the optimized “beva then chemo 3 days” group (48% tumor growth reduction, \( P = 0.007 \)). However, no more statistically significant difference was evidenced among the other treatment arms.

The median survival times were 48 days (control), 54 days (chemo), 67 days (“beva then chemo 8 days”), 70 days (“beva + chemo”), and 74 days (“beva then chemo 3 days”), as presented in Figure 5b. A log-rank test showed a statistically significant difference between each treatment arm and the control arm (\( P < 0.0001 \)). Consistently with our mathematical model predictions, further log-rank tests showed a statistically significant difference between the “beva then chemo 3 days” and “beva then chemo 8 days” arms (\( P = 0.0056 \)). Conversely, the difference did not reach statistical significance between the sequential “beva then chemo 3 days” and “beva then chemo 8 days” arms (\( P = 0.072 \)).

Figure 4 Data-informed modeling simulations of various gaps between bevacizumab and pemetrexed-cisplatin administrations. (a) Median tumor growth curves. (b) Simulations of the tumor growth using different time lags between the administration of bevacizumab and pemetrexed-cisplatin (“beva then chemo”). The red curve corresponds to a time lag of 3 days. (c) Area under the tumor growth curve (AUC) as a function of the time lag. (d) Interanimal variability on the optimal lag time between bevacizumab and chemotherapy (2.8 ± 1.84 days, median ± SD).
DISCUSSION

Bevacizumab induces a transient phase of vascular normalization. If correctly identified, this could increase drug delivery and improve treatment outcome. In this study, we proposed an integrative strategy that combined experiments on NSCLC tumor-bearing mice and mathematical modeling to explore and validate improved scheduling of the sequential administration of bevacizumab with cytotoxics. In our first experiment, our data confirmed the benefits, in terms of survival and efficacy, of administering bevacizumab before cytotoxics. To analyze these data, we developed a semimechanistic mathematical model with a critical component quantifying the dynamics of vascular normalization. Model parameters were estimated using a nonlinear mixed effects approach and simulations predicted that giving bevacizumab 3 days before cytotoxics would yield better efficacy. Subsequent experiments confirmed the superiority of this optimized sequence compared with other sequential and concomitant administrations. This proves that beyond a simple shift to sequential administration to achieve better efficacy, the precise timing of the administration of each drug does matter and that mathematical modeling may help to identify optimized alternate scheduling that would require too much resources to be explored empirically.

Human NSCLC H460 cells are a canonical model when performing experimental therapeutics studies in lung cancer. In this respect, following previously published experimental studies, we have chosen the H460 model as a paradigm for mimicking NSCLC tumors. 

Compared to our previous studies, the mathematical model presented here was simplified to focus on a minimal number of equations and parameters. We abandoned a more mechanistic description of the vasculature quality in terms of stable and unstable vessels to the benefit of a more phenomenological but more parsimonious and robust model that implements normalization in terms of a simple delay from the bevacizumab concentration.

Mean tumor growth curves obtained in experiment one highlight the high interanimal variability observed within each treatment group (Figure 2a). This variability can partially explain the high residual variability ($\Sigma = 0.951$). Moreover, comparing standard errors of tumor growth data between experiments one and two emphasizes the higher variability of bioluminescence measurements compared to fluorescence imaging. A likely explanation for this discrepancy is the interanimal variability of pharmacokinetics of luciferin used as a tracer for bioluminescence.

In experiment two, although differences in survival between the “beva + chemo” and “beva then chemo 3 days” groups did not reach statistical significance, median survival was still larger in the latter than in the former (74 days vs. 70 days). Moreover, final tumor size was 38% smaller in the “beva then chemo 3 days” as compared to the “beva + chemo” group. The $P$ value of a Kruskal-Wallis test for this difference was not below the arbitrarily level of 0.05, but was closest to it ($P = 0.072$), thus still supporting superiority of the “beva then chemo 3 days” group. We hypothesize that this lack of statistical significance in our results is due to a limited power of the study, itself linked to the restricted number of animals per group for ethical constraints.

Several studies have already explored alternate sequences for administering anti-angiogenics with other drugs, with sometimes contradictory results. Rocchetti et al. found that giving bevacizumab after targeted therapy in tumor-bearing mice yielded better efficacy. Recently, Hutchinson et al. inferred (using a mathematical model and experimental data from breast cancer models) a vessel normalization window beginning 15 days after the start of anti-angiogenics which is a much larger optimal delay compared to other studies. In addition, such lag-time could hardly meet the requirements of clinical testing, because bevacizumab is usually administered on a Q2W or a Q3W basis at bedside. The optimal lag we identified in our study is in line with other experimental studies that explored the normalization window after bevacizumab or

![Figure 5](image-url)
other anti-angiogenics administration, most of these reporting an optimal delay ranging between 2 and 5 days.\textsuperscript{19,33,41} In patients, Avalone \textit{et al.}\textsuperscript{19} showed that giving bevacizumab 4 days prior to chemotherapy yielded better efficacy compared to concomitant administration.

Intriguingly, two clinical studies in NSCLC and metastatic colorectal cancer showed reduced tumor drug delivery of cytotoxic agents when administered 1 and 4 days after bevacizumab,\textsuperscript{42} which contrasts with our preclinical findings and direct clinical observation of vessel normalization after bevacizumab administration by means of interstitial fluid pressure measurements and functional computed tomography.\textsuperscript{43} More precisely, a study conducted on 10 patients with NSCLC treated with a single dose of 15 mg/kg of bevacizumab followed by \textsuperscript{11}C labeled-docetaxel found a decreased tumor uptake of the chemotherapy after 4 days.\textsuperscript{44} These findings were obtained on a small group, with a different bevacizumab dose, no repeated cycle, and a different chemotherapeutic drug than here, which altogether might explain the discrepancy. Interestingly, in this same study a wide interpatient variability was observed in the reduction of perfusion and uptake of docetaxel. To this regard, our mathematical model could be of help by characterizing and quantifying the tumor response in a patient-specific fashion and integrate relevant biomarkers of vessel normalization into a predictive numerical tool for individually optimized scheduling.

Together, these findings highlight the importance of drug scheduling and advocate further studies to optimize scheduling of anti-angiogenic drugs. In this respect, our approach in conducting experimental studies assisted by semimechanistic mathematical modeling proved to be efficient and both time-effective and cost-effective. This proof-of-concept study suggests that simplified modeling could help to address the issue of finding the optimal dosing and scheduling of anticancer treatments to improve efficacy.

Critically, our model could be used in a biomarker-based strategy for improving anti-angiogenic therapy. As observed in Figure 4d, even in a homogenous controlled animal population, variability in the optimal gap was observed. Although no predictive biomarker has yet been clearly validated with anti-angiogenics,\textsuperscript{45} our model parameters (possibly included in a broader modeling of metastatic disease\textsuperscript{46}) could be quantitatively linked to imaging data and/or predictive circulating biomarkers acting as covariates.\textsuperscript{17,48} Consequently, this would provide personalized simulations of response to treatment allowing to individually adapt the dose and timing in order to maximize the efficacy and reduce the toxicity at the patient’s bedside.

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AUTHOR CONTRIBUTIONS. D.-C.I., R.E.C., A.B., J.C., and S.B. wrote the article. J.C., C.M., B.L., F.B., D.B., and S.B. designed the research. A.B. and J.C. performed the research. D.-C.I., R.E.C., D.B., and S.B. analyzed the data.

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