

OPEN ACCESS

EDITED BY Mirian A. F. Hayashi, Federal University of São Paulo, Brazil

REVIEWED BY lordanis Kesisoglou, Laboratory Corporation of America Holdings (LabCorp), United States Qimin Huang, College of Wooster, United States

*CORRESPONDENCE
Bruce R. Levin

☑ blevin@emory.edu

[†]These authors have contributed equally to this work

RECEIVED 21 May 2025 ACCEPTED 29 July 2025 PUBLISHED 01 September 2025

CITATION

Berryhill BA, Gil-Gil T and Levin BR (2025) The joint action of antibiotics, bacteriophage, and the innate immune response in the treatment of bacterial infections.

Front Microbiol. 16:1632267

Front. Microbiol. 16:1632267. doi: 10.3389/fmicb.2025.1632267

COPYRIGHT

© 2025 Berryhill, Gil-Gil and Levin. This is an open-access article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.

The joint action of antibiotics, bacteriophage, and the innate immune response in the treatment of bacterial infections

Brandon A. Berryhill[†], Teresa Gil-Gil[†] and Bruce R. Levin*[†]

Department of Biology, Emory University, Atlanta, GA, United States

Studies of antimicrobial therapeutics have traditionally neglected the contribution of the host in determining the course of treatment and its outcome. One critical host element, which shapes the dynamics of treatment is the innate immune system. Studies of chemotherapeutics and complementary therapies such as bacteriophage (phage), are commonly performed with mice that purposely have an ablated innate immune system. Here, we generate a mathematical and computer-simulation model of the joint action of antibiotics, phage, and phagocytes. Our analysis of this model highlights the need for future studies to consider the role of the host's innate immune system in determining treatment outcomes. Critically, our model predicts that the conditions under which resistance to the treatment agent(s) will emerge are much narrower than commonly anticipate. We also generate a second model to predict the dynamics of treatment when multiple phages are used. This model provides support for the application of cocktails to treat infections rather than individual phages. Overall, this study provides hypotheses that can readily be tested experimentally with both in vitro and in vivo experiments.

KEYWORDS

Innate immunity, infection dynamics, antibiotics, bacteriophages, antibiotic resistance, phage resistance, mathematical and computer-simulation modeling

Introduction

As a consequence of the increasing frequency of infections with antibiotic resistant bacteria there has been an increase in research on and the application of bacteriophage (phage) for the treatment of bacterial infections (Olawade et al., 2024; Salam et al., 2023). Phage therapy was employed before the advent of antibiotics but was ultimately replaced by these drugs; however, the recent resurrection of phage therapy sees these viruses used concomitantly with antibiotics almost exclusively (Summers, 2012; Berryhill et al., 2021; Li et al., 2021). There is currently a lack of understanding about the interactions between these bacterial viruses and these drugs, especially in regards to the conditions where they act either synergistically or antagonistically (Osman et al., 2023). The purpose of this this report is to use mathematical and computer simulation models to explore the population dynamic and evolutionary processes required for effective therapy with antibiotics and phage.

In exploring joint phage and antibiotic therapy, it is critical to consider the contribution of the host's innate immune system in the control of bacterial infections. The innate immune defenses play a prominent role to the course of antibiotic and phage therapy and need to be considered in studies evaluating the effect of these agents both independently

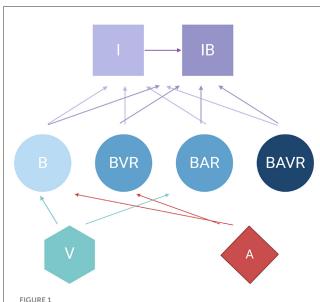


Diagram of the model of the joint action of antibiotics, phage, and the innate immune system in the dynamics of treatment of a bacterial infection. For the definitions of the variable in the above diagram, see Table 1. All parameters, their definitions, and values used in the simulations of this model are presented in Table 2 unless otherwise stated.

and when used together (Anonymous, 2013; Berti et al., 2020; Carroll-Portillo and Lin, 2019). In this report, we restrict our consideration of the innate immune system to phagocytes and phagocytosis. We give particular focus to the effect that dosing (e.g., antibiotic first, phage first, or co-administration) has to the outcomes of infection treatment and the role that the emergence of antibiotic and phage resistances have on treatment dynamics.

Phages, when given to treat an infection, are often administered as cocktails of multiple phages, with one of the primarily goals being preventing the ascent of phage resistance, which would decrease treatment efficacy (Abedon et al., 2021). The majority of our modeling results presented here assume only one phage and antibiotic are used; however, we do employ a second model to determine the contribution multiple phages would have to the dynamics of treated bacterial given the emergence of phage resistant mutants.

Mathematical models

A model of phage, antibiotics, and the innate immune system

The model developed here is an extension of that in Levin et al. (2017) which has been expanded to include phage. This model assumes continuous culture (chemostat) conditions (Chao et al., 1977). Briefly, this model combines the pharmacodynamics of antibiotic treatment developed in previous studies (Regoes et al., 2004; Wiuff et al., 2005; Levin and Udekwu, 2010) with the dynamics of phagocytosis consider in Ankomah and Levin (2014). Shown in Figure 1 is a diagram of the model employed in this report. Tables 1, 2 detail the variables of this model and the default parameters used in our simulations, respectively.

TABLE 1 Variables used in the model of the joint action of antibiotics, phage, and the innate immune system.

Variable	Definition	Color in figures
R	The limiting resource	Light blue
В	Antibiotic-sensitive, phage-sensitive bacteria	Dark blue
BVR	Antibiotic-sensitive, phage-resistant bacteria	Orange
BAR	Antibiotic-resistant, phage-sensitive bacteria	Green
BAVR	Antibiotic-resistant, phage-resistant bacteria	Red
V	Lytic phage	Teal
A	Antibiotic	Purple
I	Free phagocytes	Pink
IB	Phagocytes that have engulphed at least one bacterium	Light pink

Pharmacodynamics of antibiotic treatment

We assume that resource $(R, \mu g/mL)$ enters the environment at a constant rate and that the pharmacodynamics of the antibiotics and bacteria are modeled by a Hill function where: Π (A, R) is the net growth/death rate of the bacteria (Equation 1) (Regoes et al., 2004). Equation 1 is written generally with i in place of the bacterial states (e.g., B or BVR). For this model, we assume the resource is the unique agent limiting the growth and final density in the absence of antibiotics, phage, or the immune system—analogous to the carbon source in a minimal media (Stewart and Levin, 1973). To simulate the effect that the decreasing limiting resource concentration has on the physiological state of the bacteria, we include a term $\psi(R)$ defined in Equation 2 (Monod, 2012).

$$\Pi_{i}(A,R) = \left(vi - \frac{(vi - vmin) \cdot \left(\frac{A}{MIC}\right)^{\kappa}}{\left(\frac{A}{MIC}\right)^{\kappa} - \left(\frac{vmin}{vi}\right)}\right) \cdot \psi(R)$$
 (1)

$$\psi(R) = \frac{R}{R + k} \tag{2}$$

Mathematical model of phage, antibiotics, and the innate immune system

To simulate the treatment of populations of antibiotic-sensitive and phage-sensitive bacteria, antibiotic-resistant and phage-sensitive bacteria, antibiotic-sensitive and phage-resistant bacteria, and antibiotic-resistant and phage-resistant bacteria with antibiotics and phage, we construct a series of coupled, ordered differential equations (Equations 3 through 10). Free and bacteria-populated phagocytes, I and IB, engulf free bacteria at a rate proportional to the product of their densities, that of the free bacteria and a rate constant, γ , which is the same for free bacteria of all states and both I and IB phagocytes. With these definitions, assumptions, and the parameters defined and presented in Tables 1, 2, the rates of change in the densities of the different populations are

TABLE 2 Parameters and the values used in the simulation of the joint action of antibiotics, phage, and the innate immune system.

Parameter	Definition	Value	Units	Source
vb, vbar, vbvr, vbavr	Maximum growth rates	1.0, 0.9, 0.9, 0.8	per cell per hour	This report
С	Maximum resource concentration	1,000	μg/mL	This report
е	Resource conversion efficiency	5·10 ⁻⁷	μg/cell	Stewart and Levin, 1973
μ	Mutation rate	10 ⁻⁷	per hour	Luria and Delbrück, 1943
IMAX	Maximum phagocyte density	10 ⁵	per mL	This report
γ	Phagocyte gobbling constant	9.10 ⁻⁶	per cell per hour	This report
w	Flow rate	0.1	mL per hour	Chao et al., 1977
δ	Phage adsorption rate	10 ⁻⁷	per hour per mL	Berryhill et al., 2023a
β	Phage burst size	50	particles per cell	Berryhill et al., 2023a
da	Decay rate of the antibiotic	0	μg/mL per hour	Levin and Udekwu, 2010
vmin	Maximum kill rate of antibiotic	−4.0 or −0.001	per cell per hour	Berryhill et al., 2023b
К	Hill parameter	1.0		Regoes et al., 2004
MIC	Minimum inhibitory concentration	1.0	μg/mL	This report
k	Monod constant	1.0	μg	Monod, 2012
Dose	Dosing interval	8.0	Hours	This report

given by:

$$\frac{dR}{dt} = w \cdot (C - R) - e \cdot \psi (R) \cdot (vb \cdot B + vbar \cdot BAR + vbvr \cdot BVR + vbavr \cdot BAVR)$$
(3)

$$\frac{dB}{dt} = \pi (A, R) \cdot B - \gamma \cdot B (I + IB) - w \cdot B \tag{4}$$

$$\frac{dBAR}{dt} = vbar \cdot BAR \cdot \psi(R) - \gamma \cdot BAR \cdot (I + IB) - \delta \cdot V \cdot BAR \cdot \psi(R) - w \cdot BAR$$
 (5)

$$\frac{dBVR}{dt} = \pi (A, R) \cdot BVR \cdot \psi (R) - \gamma \cdot BVR \cdot (I + IB) - w \cdot BVR$$
(6)

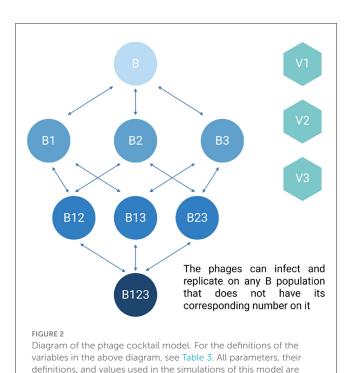
$$\frac{dI}{dt} = w \cdot (IMAX - I) - \gamma \cdot (B + BAR + BVR + BAVR) \cdot I - w \cdot I$$

$$\frac{dIB}{dt} = \gamma \cdot I \cdot (B + BAR + BVR + BAVR) - w \cdot IB \tag{8}$$

$$\frac{dIB}{dt} = \gamma \cdot I \cdot (B + BAR + BVR + BAVR) - w \cdot IB$$

$$\frac{dV}{dt} = \delta \cdot V \cdot (B + BAR) \cdot \beta \cdot \psi (R) - w \cdot V + VIN$$
(9)

$$\frac{dA}{dt} = -A \cdot (da + w) + AIN \tag{10}$$



A model of phage cocktails

Shown in Figure 2 is a diagram of the model employed for modeling phage cocktails in this report. Tables 3, 4 detail the variables of this model and the default parameters used in our simulations, respectively.

Mathematical model of phage cocktails

To simulate the treatment of populations of bacteria which are either phage-sensitive, resistant to one phage, resistant to two phages, or resistant to three phages and three different phages we construct a series of coupled, ordered differential equations (Equations 11 through 19). We assume resistance to each phage is independent. In this model, the transitions between the various phage-resistant states occurs stochastically (Stewart et al., 1990). We simulate these transitions with a Monte Carlo process (Metropolis and Ulam, 1949). A random number x ($0 \le x \le 1$) from a rectangular distribution is generated (Gentle, 2003). If x is less than the product of the number of cells in the generating state

presented in Table 4 unless otherwise stated.

(B, the density time the volume of the vessel, Vol), the transition rate (μ) and the step size (dt) of the Euler method employed for solving the differential equations (Hairer and Wanner, 1996), for example if $x < B^*\mu^*dt^*Vol$, then ADDBBR1 cells are added to the BR1 population and removed from the B population where ADDBBR1 = 1/(dt*Vol). With these definitions, assumptions, and the parameters defined and presented in Tables 3, 4, the rates of change in the densities of the different populations are given by:

$$\frac{dR}{dt} = w \cdot (C - R) - e \cdot \psi(R) \cdot (vb \cdot B + v1 \cdot BR1 + v2 \cdot BR2 + v3 \cdot BR3)$$

$$+ v12 \cdot BR12 + v13 \cdot BR13 + v23 \cdot BR23 + v123 \cdot BR123) \qquad (11)$$

$$\frac{dB}{dt} = vb \cdot B \cdot \psi(R) - \delta \cdot (V1 + V2 + V3) \cdot B \cdot \psi(R) - w \cdot B$$

$$+ ADDBR1B + ADDBR2B + ADDBR3B - ADDBBR1$$

$$- ADDBBR2 - ADDBBR3 \qquad (12)$$

$$\frac{dBR1}{dt} = vb1 \cdot BR1 \cdot \psi(R) - \delta \cdot BR1 \cdot (V2 + V3) \cdot \psi(R) - w \cdot BR1$$

$$+ ADDBR1B - ADDBR12BR1 + ADDBR13BR1 - ADDBR13BR1 - ADDBR1BR13 \qquad (13)$$

$$\frac{dBR2}{dt} = vb2 \cdot BR2 \cdot \psi(R) - \delta \cdot BR2 \cdot (V1 + V3) \cdot \psi(R) - w \cdot BR2$$

$$+ ADDBBR2 + ADDBR12BR2 + ADDBR23BR2$$

$$- ADDBR2B - ADDBR2BR12 - ADDBR2BR23 \qquad (14)$$

$$\frac{dBR3}{dt} = vb3 \cdot BR3 \cdot \psi(R) - \delta \cdot BR3 \cdot (V1 + V2) \cdot \psi(R) - w \cdot BR3$$

TABLE 3 Variables used in the model of bacteriophage cocktails.

+ ADDBBR3 + ADDBR13BR3 + ADDBR23BR3

R	The limiting resource
V1	Phage 1
V2	Phage 2
V3	Phage 3
В	Bacteria susceptible to all phages
BR1	Bacteria resistant to phage 1
BR2	Bacteria resistant to phage 2
BR3	Bacteria resistant to phage 3
BR12	Bacteria resistant to phage 1 and phage 2
BR13	Bacteria resistant to phage 1 and phage 3
BR23	Bacteria resistant to phage 2 and phage 3
BR123	Bacteria resistant to phage 1, phage 2, and phage 3

$$\frac{dBR12}{dt} = vb12 \cdot BR12 \cdot \psi(R) - \delta \cdot BR12 \cdot V3 \cdot \psi(R) - w \cdot BR12 \\ + ADDBR1BR12 + ADDBR2BR12 + ADDBR123BR12 \\ - ADDBR12BR1 - ADDBR12BR2 - ADDBR12BR123$$
 (16)
$$\frac{dBR13}{dt} = vb13 \cdot BR13 \cdot \psi(R) - \delta \cdot BR13 \cdot V2 \cdot \psi(R) - w \cdot BR13 \\ + ADDBR3BR13 + ADDBR1BR13 + ADDBR123BR13 \\ - ADDBR13BR3 - ADDBR13BR1 - ADDBR13BR123$$
 (17)
$$\frac{dBR23}{dt} = vb23 \cdot BR23 \cdot \psi(R) - \delta \cdot BR23 \cdot V1 \cdot \psi(R) - w \cdot BR23 \\ + ADDBR2BR23 + ADDBR3BR23 + ADDBR123BR23 \\ - ADDBR23BR2 - ADDBR23BR3 - ADDBR23BR123$$
 (18)
$$\frac{dBR123}{dt} = vb123 \cdot BR123 \cdot \psi(R) - w \cdot BR123 + ADDBR12BR123 \\ + ADDBR23BR123 - ADDBR123BR12 \\ - ADDBR123BR13 - ADDBR123BR12$$
 (19)

(15)

- ADDBR3B - ADDBR3BR13 - ADDBR3BR23

Results

Control by the immune system in the absence of treatment

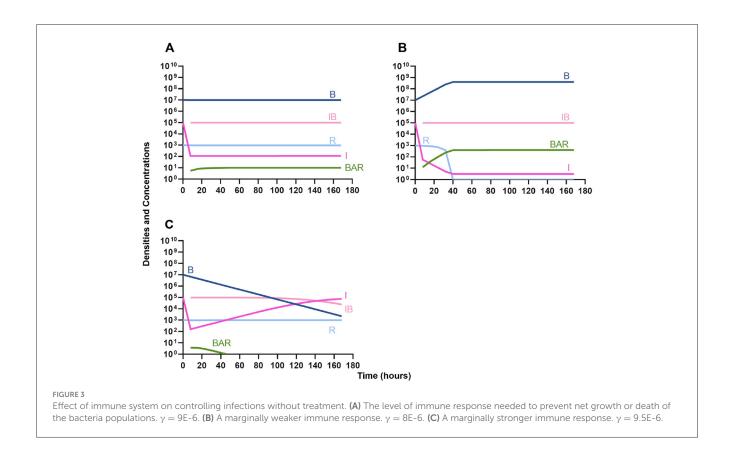
We begin our analysis of the predictions generated by the first model by considering the effect the primary, unmeasured parameter γ (the phagocyte gobbling rate) has on the dynamics of the infection. In Figure 3, we consider three differing values of γ , demonstrating that the model is highly sensitive to this parameter. Going forward, all simulations are performed with the value of γ in Figure 3A, where the immune system is capable of suppressing the growth of the bacteria but not capable of clearing the infection on its own.

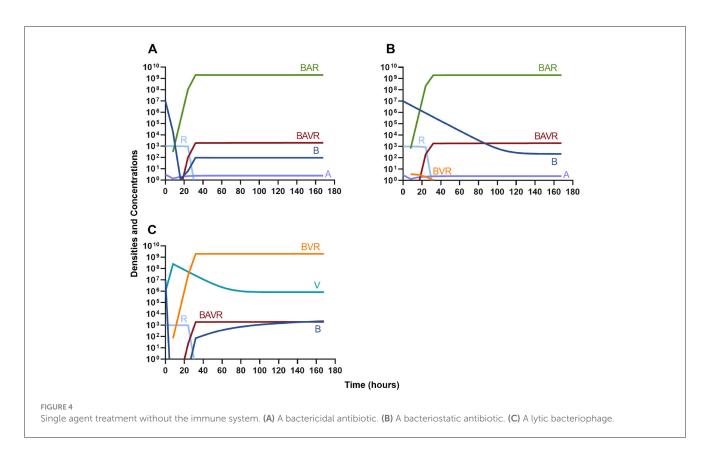
Treatment of infections in the absence of the immune system

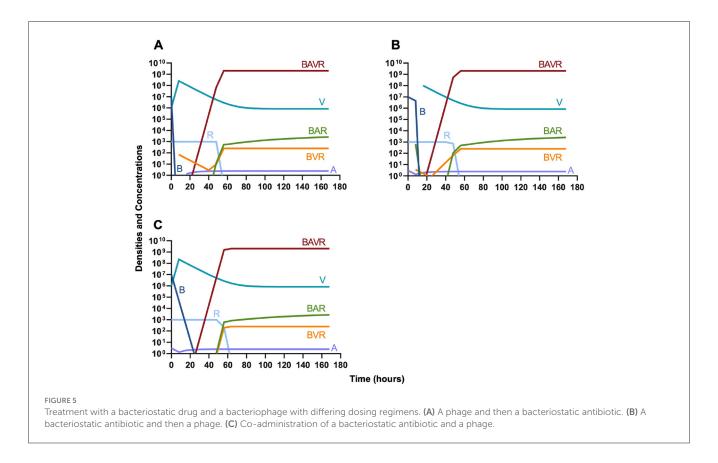
We then analyze the effects that treatment in the absence of the innate immune system has on the dynamics of infection.

TABLE 4 Parameters and the values used in the model of bacteriophage cocktails.

Parameter	Definition	Value	Units	Source
vb, vb1, vb2, vb3	Maximum growth rates (based on the number of resistant states)	1.0, 0.9, 0.8, 0.7	per cell per hour	This report
С	Maximum resource concentration	1,000	μg/mL	This report
e	Resource conversion efficiency	5.10-7	μg/cell	Stewart and Levin, 1973
μ	Mutation rate	10-7	per hour	Luria and Delbrück, 1943
w	Flow rate	0.1	per hour	Chao et al., 1977
δ	Phage adsorption rate	10 ⁻⁷	per hour per mL	Berryhill et al., 2023a
β	Phage burst size	50	particles per cell	Berryhill et al., 2023a
k	Monod constant	1.0	μg	Monod, 2012







Single agent treatment (controls)

In Figure 4A we consider treatment of an initially sensitive bacteria population with a highly bactericidal antibiotic; in Figure 4B the only treatment is a bacteriostatic drug; and, in Figure 4C the bacteria are treated with a lytic bacteriophage. Notably, all agents are capable of controlling the initial infection, however resistance to that agent does rapidly ascend.

Bacteriostatic antibiotics

In evaluating the joint action of phage and antibiotics, we first examine a highly lytic phage in combination with a bacteriostatic antibiotic. There are three distinct dosing regimens: phage first (Figure 5A), bacteriostatic drug first (Figure 5B), and co-administration of both the phage and antibiotic (Figure 5C). Our simulations predict that the phage first regimen clears the initial bacterial population the quickest, followed by antibiotic first, and then co-administration being the slowest to clear the initial population. However, in all cases resistance to both treating agents ascends in roughly the same amount of time, $\sim\!40$ h.

Bactericidal antibiotics

We continue our investigation of the dynamics of treatment without the immune system by studying the joint action of a bactericidal drug and a phage. Again, there are three distinct dosing regimens: phage first (Figure 6A), bactericidal drug first (Figure 6B), and co-administration of both the phage and antibiotic (Figure 6C). Our simulations provide the same predictions as those

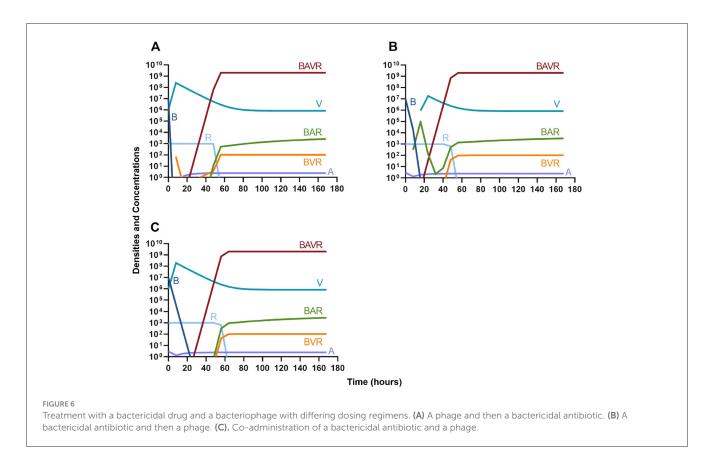
for the bacteriostatic drug. Indicating, that both bacteriostatic and bactericidal antibiotics can be equally as effective, however resistant will ultimately ascend.

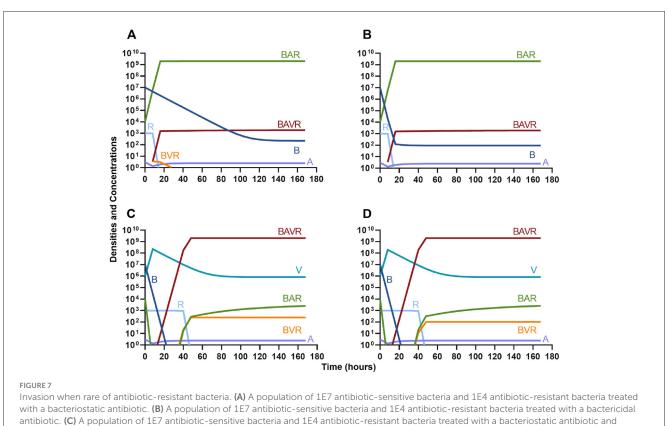
Phage suppression of antibiotic resistance

One argument for the use of phages in combination with antibiotics is that the virus is able to suppress the antibiotic-resistant population. To address this hypothesis, in Figure 7 we consider a scenario where the majority of the bacteria are susceptible to an antibiotic, but there is a minor population at a ratio of 1:1,000 which is resistant to the treating drug (either a bacteriostatic drug as in Figures 7A, C or a bactericidal drug Figures 7B, D). When the phage is not present (Figures 7A, B) the antibiotic-resistant minority population is able to ascend to dominance and treatment fails. While, when the phage is present, the antibiotic-resistant population is rapidly controlled, but a population which is resistant to both the phage and antibiotic ascends to dominance; however, the emergence of this double-resistant population takes twice as long to ascend to dominance.

Antibiotic suppression of phage resistance

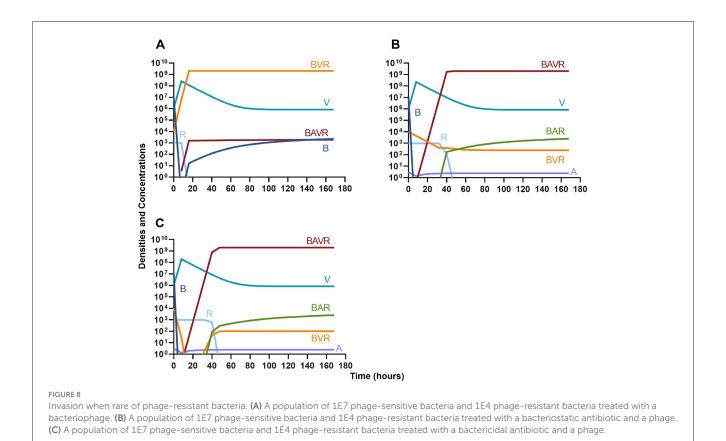
Finally, we address the same situation as in Figure 7 but instead consider that a phage-resistant population is the minor population present at the initiation of treatment. As expected, in Figure 8A, when treated with just the phage, the phage-resistant population ascends to a majority. As in Figure 7, when treated with either a bacteriostatic (Figure 8B) or bactericidal antibiotic

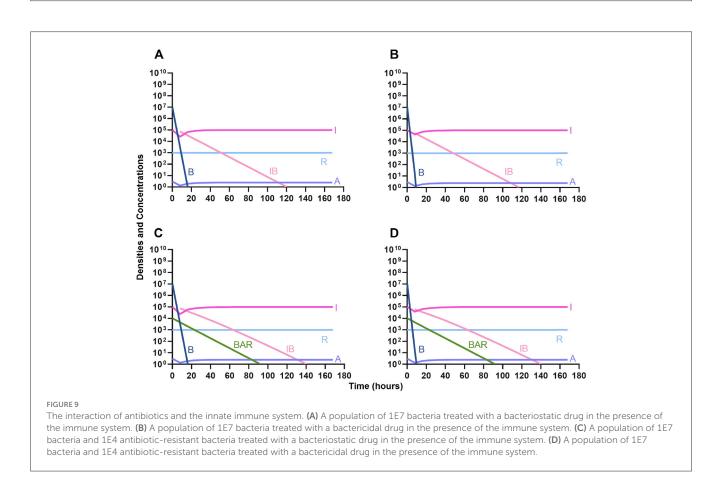


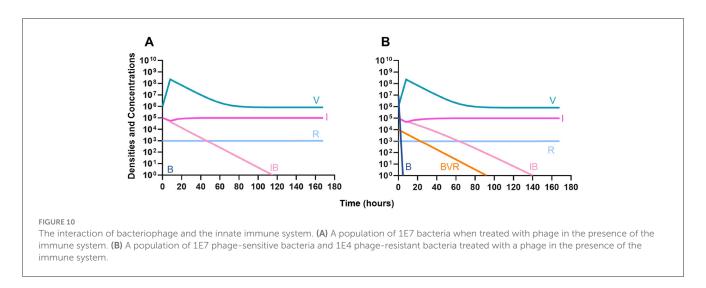


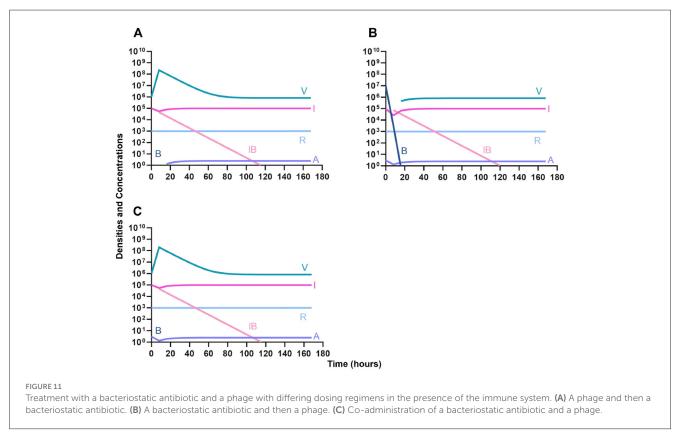
bacteriophage. (D) A population of 1E7 antibiotic-sensitive bacteria and 1E4 antibiotic-resistant bacteria treated with a bactericidal antibiotic and a

bacteriophage.









(Figure 8C) in conjunction with the phage, the initial phageresistant population is controlled, but a population resistant to both treating agents emerges. In this case, the double-resistant mutant takes approximately four times as long to dominate as the single-resistant population.

Single agent treatment and the innate immune system

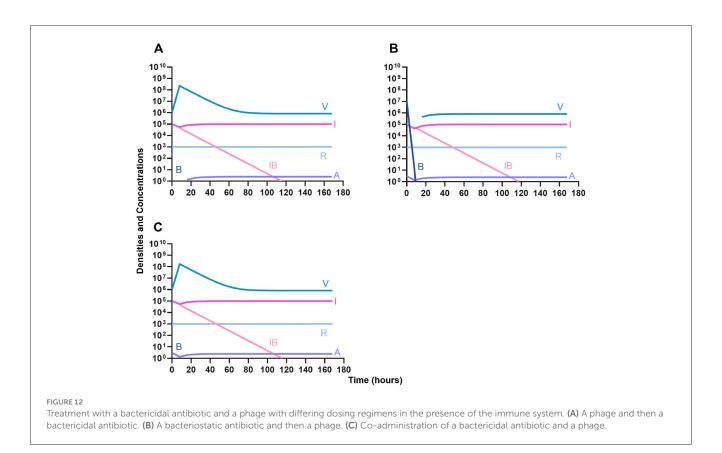
Given the above results where we do not consider the impact of the innate immune system, we continue our modeling by considering similar situations but with the immune response.

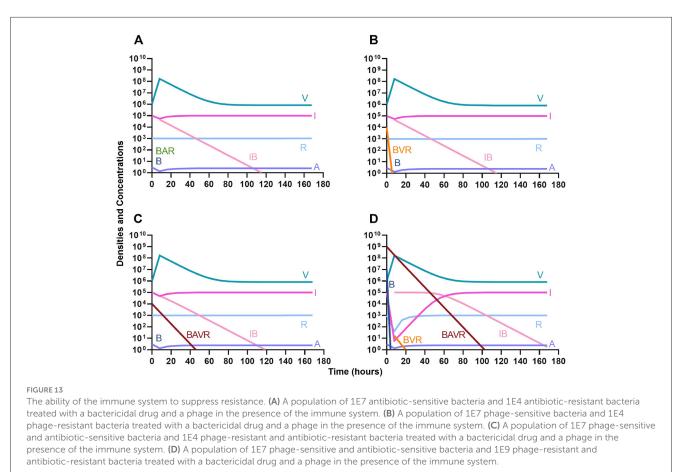
Antibiotics and the innate immune system

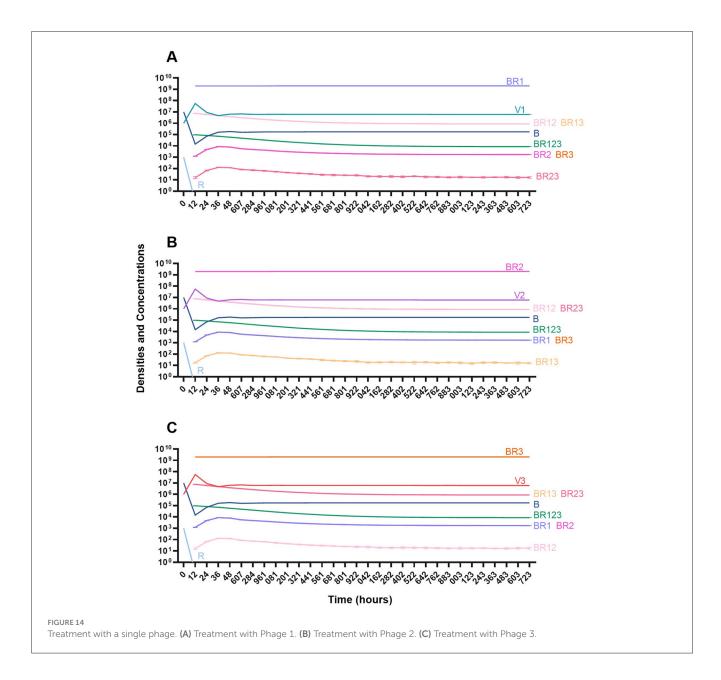
Given the consideration of the immune system alone in Figure 1, we begin this section, by considering the interaction of antibiotics and the immune system. With both bacteriostatic and bactericidal drugs (Figures 9A, B, respectively), the immune system and the antibiotics together rapidly clear the infection, and antibiotic-resistant populations do not appear. Moreover, if the antibiotic-resistant populations are present initially (Figures 9C, D), they are rapidly lost as well.

Phage and the innate immune system

In Figure 10, we consider the same situation as in Figure 9 but instead treat with a lytic phage. As in the previous section, the phage







and the immune system can rapidly clear the infection (Figure 10A) and minor phage-resistant populations do not ascend (Figure 10B).

Phage, antibiotics, and the innate immune system

We next expand our consideration of the joint action of treatment and the immune system to situations where phage and antibiotics are used in conjunction.

Bacteriostatic antibiotics

First, we evaluate the dynamics of infection when treated with a bacteriostatic drug and a lytic phage in the presence of the immune system. As above, we consider the effects that dosing order has on treatment outcome (Figure 11) and find that the effect of treatment dosing order is minimal, and all condition are capable of clearing the infection without the ascent of resistance. Although, the condition where the bacteriostatic drug is administered first does have the highest time to clearance (Figure 11B).

Bactericidal antibiotics

We determine the effect of dosing order for a bactericidal drug and phage with the innate immune system as in Figure 11. The results of these simulations in Figure 12 are parallel to those in Figure 11, once again demonstrating that there is no effect on treatment outcome with dosing order or using a bacteriostatic vs. a bactericidal drug. However, the time to clearance is once again longer when the antibiotic is applied first (Figure 12B).

Suppression of resistance

Once again, the logic of using multiple treating agents is predicated upon the suppression of resistance. We finally consider three situations where a minor population is resistant to either one (Figures 13A, B) or both treating agents (Figure 13C). Ultimately, our results indicate that resistant subpopulations, regardless of what they are resistant to, will not ascend under treatment when the immune system is present. Interestingly, when a population that is resistant to both the phage and a bactericidal drug is initially dominant and at a very high density, treatment can still control and eventually clear the infection (Figure 13D).

A model of phage cocktails

Motivating the use of cocktails of phage, rather than a single phage, for treatment is the suppression (or elimination) of phage-resistant mutants. Here, we consider a model where treatment can be with up to three phages and bacteria resistant to each phage and the various combinations of the three phages can emerge. This model does not have the innate immune system, nor does it have antibiotics.

Single phage treatment

In the absence of the immune system, when a single phage is used for therapy, resistance to the treating phage very rapidly ascends to dominate and treatment fails (Figure 14). However, the phage is maintained over time due to the transition from the resistant state to the sensitive state.

Two phage treatment

We then consider a situation where two phages are used in combination (Figure 15). Notably, the time before the mutant resistant to both treating phages ascends to dominance is longer than when one phage is used for treatment. However, since this model is stochastic, there is variability when the single-resistant mutants emerge and thereby variability in when the double-resistant mutants emerge.

Three phage treatment

Finally, under treatment with three phages (Figure 16), single phage-resistant mutants arise at various times and give way to double-phage resistant mutants, before the triple-phage resistant mutants ultimately arise and dominate such that treatment fails, but it takes longer for treatment with three phages to fail compared to treatment with two phage and substantially long to fail that treatment with one phage.

Discussion

Motivated by the well-warranted concern about the antibiotic-resistance crisis, there has been an increase in studies on

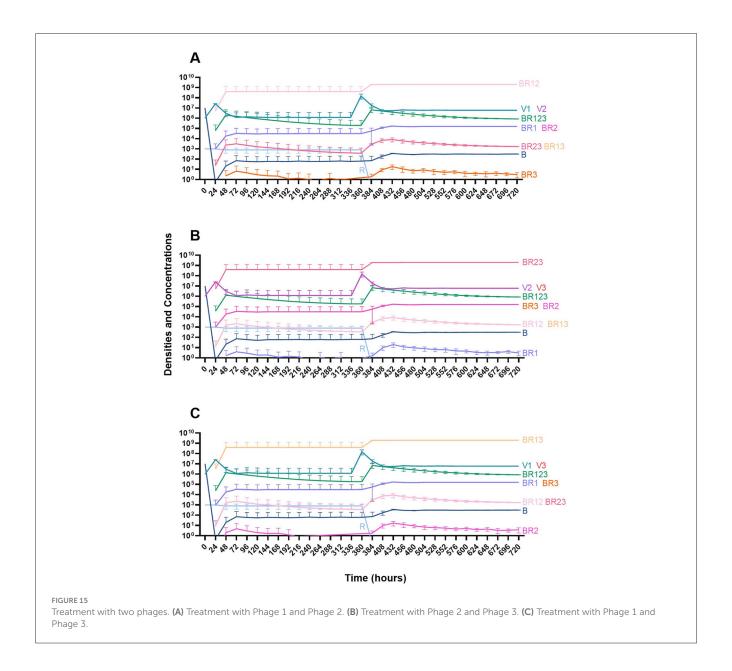
the treatment of bacterial infections (Ventola, 2015). There is no shortage of treatment options for infections given the numerous types and classes of antibiotics as well as burgeoning complementary therapies such as the use of bacteriophages (phages). However, many of the studies neglect the role of the host in the dynamics of infections, particularly the role of the innate immune system (Modlin, 2012). To lay the foundation for further experimental studies, in this report, we create two mathematical and computer-simulation models that generate testable hypotheses about the population and evolutionary dynamics of bacterial infections under treatment with antibiotics and phage in the presence of the host's innate immune system.

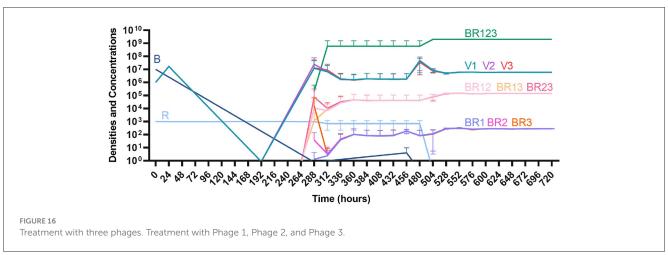
The results of the analysis of our models underscore the need to consider the role of the innate immune system in subsequent experimental studies. In the absence of the immune system, resistance to the treating agent invariable emerges independent of the treating agents or the regimens in which they are employed. On the other hand, when the immune system is present, resistance does not emerge; indeed, even when a high density of pan-resistant bacteria is present, the infection can still be controlled with treatment. As previously reported with numerous *in vitro* and *in vivo* studies, the difference in treatment outcome with bacteriostatic and bactericidal antibiotics is de minimis (Berryhill et al., 2023b; Wald-Dickler et al., 2018).

The predictions of this theory are also congruent with previous results that demonstrate that phage can be as effective as antibiotics in controlling infections (Lin et al., 2017). This model also provides support for the intuitive conclusion that more phages are better than fewer phages. While resistance to multiple treating phages does ultimately emerge, the time for resistance to dominate for one treating phage is measured in hours, while the time for resistance to dominate for three treating phages is measured in days.

As with all purely theoretical studies, we have had to make assumptions about the parameters which we could not find in previous reports. One key parameter to which the model is incredibly sensitive which has not been estimated is the rate of phagocyte gobbling. For this report, we have elected to use a phagocyte gobbling rate constant that keeps the density of the infecting bacteria steady without the presence of any treatment. This assumption allows for us to determine the potential impact that the treatments and their order are having on the dynamics of the infection. However, these immune parameters, and moreover, all the parameters used in this study can be readily estimated experimentally.

Taken together, the analysis of our mathematical and computer-simulation models makes highly testable predictions about the dynamics of treatment which could be supported or rejected by using a mix of *in vitro* and *in vivo* models. It is the intent of these authors to explore the validity of the hypotheses generated above with the *Galleria mellonella* infection model system (Berryhill et al., 2024). Be that as it may, these predictions are agnostic to the experimental system and the hypotheses could easily be tested in other systems such as cell culture or mice (Carryn et al., 2002; Anderson et al., 2019).





Materials and methods

Numerical solutions (simulations)

For our numerical analysis of the coupled, ordered differential equations presented (Equations 1–12), we used Berkeley Madonna with the parameters presented in Table 2 (Macey et al., 2000). Copies of the Berkeley Madonna programs used for these simulations are available at www.eclf.net.

Data availability statement

The original contributions presented in the study are included in the article, further inquiries can be directed to the corresponding author.

Author contributions

BAB: Conceptualization, Data curation, Formal analysis, Funding acquisition, Investigation, Methodology, Project administration, Resources, Software, Supervision, Validation, Visualization, Writing – original draft, Writing – review & editing. TG-G: Conceptualization, Data curation, Formal analysis, Funding acquisition, Investigation, Methodology, Project administration, Resources, Software, Supervision, Validation, Visualization, Writing – original draft, Writing – review & editing. BRL: Conceptualization, Data curation, Formal analysis, Funding acquisition, Investigation, Methodology, Project administration, Resources, Software, Supervision, Validation, Visualization, Writing – original draft, Writing – review & editing.

Funding

The author(s) declare that financial support was received for the research and/or publication of this article. BRL would like to thank the U.S. National Institute of General Medical Sciences for their funding support via R35GM136407 and the Emory University Antibiotic Resistance Center. The funding sources had no role in the design of this study and will not have any role during its execution, analysis, interpretation of the data, or drafting of this report.

Acknowledgments

We thank the other members of the Levin Lab for their comments on an earlier version of this manuscript.

Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

Generative AI statement

The author(s) declare that no Gen AI was used in the creation of this manuscript.

Any alternative text (alt text) provided alongside figures in this article has been generated by Frontiers with the support of artificial intelligence and reasonable efforts have been made to ensure accuracy, including review by the authors wherever possible. If you identify any issues, please contact us.

Publisher's note

All claims expressed in this article are solely those of the authors and do not necessarily represent those of their affiliated organizations, or those of the publisher, the editors and the reviewers. Any product that may be evaluated in this article, or claim that may be made by its manufacturer, is not guaranteed or endorsed by the publisher.

Author disclaimer

The content is solely the responsibility of the authors and does not necessarily represent the official views of the National Institutes of Health.

References

Abedon, S. T., Danis-Wlodarczyk, K. M., and Wozniak, D. J. (2021). Phage cocktail development for bacteriophage therapy: toward improving spectrum of activity breadth and depth. *Pharmaceuticals* 14:1019. doi: 10.3390/ph14101019

Anderson, L. S., Reynolds, M. B., Rivara, K. R., Miller, L. S., and Simon, S. I. (2019). A mouse model to assess innate immune response to Staphylococcus aureus infection. *J. Vis. Exp.* 144:e59015. doi: 10.3791/59015

Ankomah, P., and Levin, B. R. (2014). Exploring the collaboration between antibiotics and the immune response in the treatment of acute, self-limiting infections. *Proc. Nat. Acad. Sci. U. S. A.* 111, 8331–8338. doi: 10.1073/pnas.1400352111

Anonymous (2013). *Autoimmunity: From Bench to Bedside*, eds. J. M. Anaya, Y. Shoenfeld, A. Rojas-Villarraga, R. A. Levy, and R. Cervera. Bogota: El Rosario University Press.

Berryhill, B. A., Garcia, R., McCall, I. C., Manuel, J. A., Chaudhry, W., Petit, M. A., et al. (2023a). The book of Lambda does not tell us that naturally occurring lysogens of

Escherichia coli are likely to be resistant as well as immune. *Proc. Nat. Acad. Sci. U. S. A.* 120:e2212121120. doi: 10.1073/pnas.2212121120

Berryhill, B. A., Gil-Gil, T., Manuel, J. A., Smith, A. P., Margollis, E., Baquero, F., et al. (2023b). What's the matter with MICs: bacterial nutrition, limiting resources, antibiotic pharmacodynamics. *Microbiol. Spectr.* 11, e04091–e04022. doi: 10.1128/spectrum.04091-22

Berryhill, B. A., Gil-Gil, T., Witzany, C., Goldberg, D. A., Vega, N. M., Regoes, R. R., et al. (2024). The dynamics of *Staphylococcal* infection and their treatment with antibiotics and bacteriophage in the *Galleria mellonella* model system. *bioRxiv* [*Preprint*]. doi: 10.1101/2024.08.23.609294

Berryhill, B. A., Huseby, D. L., McCall, I. C., Hughes, D., and Levin, B. R. (2021). Evaluating the potential efficacy and limitations of a phage for joint antibiotic and phage therapy of *Staphylococcus aureus* infections. *Proc. Natl. Acad. Sci. U. S. A.* 118:e2008007118. doi: 10.1073/pnas.2008007118

- Berti, A., Rose, W., Nizet, V., and Sakoulas, G. (2020). Antibiotics and innate immunity: a cooperative effort toward the successful treatment of infections. *Open Forum Infect. Dis.* 7:ofaa302. doi: 10.1093/ofid/ofaa302
- Carroll-Portillo, A., and Lin, H. C. (2019). Bacteriophage and the innate immune system: access and signaling. *Microorganisms* 7:625. doi: 10.3390/microorganisms7120625
- Carryn, S., Van Bambeke, F., Mingeot-Leclercq, M. P., and Tulkens, P. M. (2002). Comparative intracellular (THP-1 macrophage) and extracellular activities of beta-lactams, azithromycin, gentamicin, and fluoroquinolones against Listeria monocytogenes at clinically relevant concentrations. *Antimicrob. Agents Chemother.* 46, 2095–2103. doi: 10.1128/AAC.46.7.2095-21
- Chao, L., Levin, B. R., and Stewart, F. M. A. (1977). Complex community in a simple habitat: an experimental study with bacteria and phage. *Ecology* 58, 369–378. doi: 10.2307/1935611
- Gentle, J. E. (2003). Random Number Generation and Monte Carlo Methods, Vol. 381. New York, NY: Springer. doi: 10.1007/b97336
- Hairer, E., and Wanner, G. (1996). Solving Ordinary Differential Equations II Stiff and Differential-Algebraic Problems, Vol. 14. Berlin: Springer-Verlag. doi: 10.1007/978-3-642-05221-7
- Levin, B. R., Baquero, F., Ankomah, P., and McCall, I. C. (2017). Phagocytes, antibiotics, and self-limiting bacterial infections. *Trends Microbiol.* 25, 878–892. doi: 10.1016/j.tim.2017.07.005
- Levin, B. R., and Udekwu, K. I. (2010). Population dynamics of antibiotic treatment: a mathematical model and hypotheses for time-kill and continuous-culture experiments. *Antimicrob. Agents Chemother.* 54, 3414–3426. doi: 10.1128/AAC.00381-10
- Li, X., He, Y., Wang, Z., Wei, J., Hu, T., Si, J., et al. (2021). A combination therapy of phages and antibiotics: two is better than one. *Int. J. Biol. Sci.* 17, 3573–3582. doi: 10.7150/ijbs.60551
- Lin, D. M., Koskella, B., and Lin, H. C. (2017). Phage therapy: an alternative to antibiotics in the age of multi-drug resistance. *World J. Gastrointest. Pharmacol. Ther.* 8, 162–173. doi: 10.4292/wjgpt.v8.i3.162
- Luria, S. E., and Delbrück, M. (1943). Mutations of bacteria from virus sensitivity to virus resistance. *Genetics* 28:491. doi: 10.1093/genetics/28.6.491
- Macey, R., Oster, G., and Zahnley, T. (2000). Berkeley Madonna User's Guide. Berkeley, CA: University of California, 67.

- Metropolis, N., and Ulam, S. (1949). The Monte Carlo method. *J. Am. Stat. Assoc.* 44, 335–341. doi: 10.1080/01621459.1949.10483310
- Modlin, R. L. (2012). Innate immunity: ignored for decades, but not forgotten. *J. Investig. Dermatol.* 132, 882–886. doi: 10.1038/jid.2011.373
- Monod, J. (2012). The growth of bacterial cultures. Mol. Biol. Jacques Monod 139:606. doi: 10.1016/B978-0-12-460482-7.50020-8
- Olawade, D. B., Fapohunda, O., Egbon, E., Ebiesuwa, O. A., Usman, S. O., Faronbi, A. O., et al. (2024). Phage therapy: a targeted approach to overcoming antibiotic resistance. *Microb. Pathog.* 197:107088. doi:10.1016/j.micpath.2024.107088
- Osman, A. H., Kotey, F. C. N., Odoom, A., Darkwah, S., Yeboah, R. K., Dayie, N. T. K. D., et al. (2023). The potential of bacteriophage-antibiotic combination therapy in treating infections with multidrug-resistant bacteria. *Antibiotics* 12:1329. doi: 10.3390/antibiotics12081329
- Regoes, R. R., Wiuff, C., Zappala, R. M., Garner, K. N., Baquero, F., and Levin, B. R. (2004). Pharmacodynamic functions: a multiparameter approach to the design of antibiotic treatment regimens. *Antimicrob. Agents Chemother.* 48, 3670–3676. doi: 10.1128/AAC.48.10.3670-3676.2004
- Salam, M. A., Al-Amin, M. Y., Salam, M. T., Pawar, J. S., Akhter, N., Rabaan, A. A., et al. (2023). Antimicrobial resistance: a growing serious threat for global public health. Healthcare 11:1946. doi: 10.3390/healthcare11131946
- Stewart, F. M., Gordon, D., and Levin, B. (1990). Fluctuation analysis: the probability distribution of the number of mutants under different conditions. *Genetics* 124, 175–185. doi: 10.1093/genetics/124.1.175
- Stewart, F. M., and Levin, B. R. (1973). Partitioning of resources and the outcome of interspecific competition: a model and some general considerations. *Am. Nat.* 107, 171–198. doi: 10.1086/282825
- Summers, W. C. (2012). The strange history of phage therapy. Bacteriophage 2, 130–133. doi: 10.4161/bact.20757
- Ventola, C. L. (2015). The antibiotic resistance crisis: part 1: causes and threats. *PT* 40, 277–283.
- Wald-Dickler, N., Holtom, P., and Spellberg, B. (2018). Busting the myth of "static vs cidal": a systemic literature review. *Clin. Infect. Dis.* 66, 1470–1474. doi: 10.1093/cid/cix1127
- Wiuff, C., Zappala, R. M., Regoes, R. R., Garner, K. N., Baquero, F., and Levin, B. R. (2005). Phenotypic tolerance: antibiotic enrichment of noninherited resistance in bacterial populations. *Antimicrob. Agents Chemother.* 49, 1483–1494. doi: 10.1128/AAC.49.4.1483-1494.2005