Parallel evolution of phage resistance - virulence trade - offs during *in vitro* and nasal *Pseudomonas aeruginosa* phage treatment

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Abstract

With rising antibiotic resistance, there has been increasing interest in treating pathogenic bacteria

with bacteriophages (phage therapy). One limitation of phage therapy is the ease at which bacteria

can evolve resistance. Negative effects of resistance may be mitigated when resistance results in

reduced bacterial growth and virulence, or when phage coevolve to overcome resistance.

Resistance evolution and its consequences are contingent on the bacteria-phage combination and

their environmental context, making therapeutic outcomes hard to predict. One solution might be

to conduct "*in vitro* evolutionary simulations" using bacteria-phage combinations from the

therapeutic context. Overall, our aim was to investigate parallels between *in vitro* experiments and

in vivo dynamics in a human participant. Evolutionary dynamics were similar, with high levels of

resistance evolving quickly with limited evidence of phage evolution. Resistant bacteria – evolved

in vitro and *in vivo* - had lower virulence. *In vivo*, this was linked to lower growth rates of resistant

isolates, whereas *in vitro* phage resistant isolates evolved greater biofilm production. Population

sequencing suggests resistance resulted from selection on *de novo* mutations rather than sorting of

existing variants. These results highlight the speed at which phage resistance can evolve *in vivo*,

and *in vitro* experiments may give useful insights for clinical evolutionary outcomes.

Introduction

By 2050, as many as 10 million people may die annually from antimicrobial resistant infections (WHO, 2019). To combat this threat, phage therapy is being increasingly researched as a complement to antibiotics (Abedon, 2019; Kwiatek et al., 2020). Phage therapy uses viruses (phage) that infect pathogenic bacteria to clear infections (Kwiatek et al., 2020). Phage therapy has several advantages to antibiotics by being species specific (thus not harming beneficial microbiota) (Koskella and Meaden, 2013) and self-amplifying at infection sites (Górski et al., 2015). A major consideration with the use of therapeutic phages is the ease and speed (sometimes within hours) at which bacteria can evolve resistance, which can greatly limit the therapeutic potential of phages (Jariah and Hakim, 2019; Torres-Barceló, 2018; Torres-Barceló and Hochberg, 2016). However, phages may contribute to reducing the severity of infections even after resistance evolves. First, phage can sometimes evolve to overcome this resistance (i.e. bacteria-phage coevolution) and so they may continue lowering bacterial densities (Gandon et al., 2008; Hampton et al., 2020). Second, resistance can trade-off against bacterial growth rates and virulence; as the alteration or loss of bacterial surface structures, typically associated with phage resistance, can affect other fitness-related phenotypes such as nutrient uptake, biofilm formation and binding of host receptors (Mangalea and Duerkop, 2020). Predicting the evolutionary outcome of phage therapy is therefore of crucial importance (Abedon, 2017).

Bacteria-phage dynamics and its consequences for phage therapy are highly context-specific (Gandon et al., 2008; Hampton et al., 2020). Both human clinical and animal models studies report a range of outcomes with differing phage therapy success, and differences in the extent of resistance evolution and associated costs and virulence (Abedon, 2019; Brix et al., 2020; Kwiatek et al., 2020; Leitner et al., 2021). One solution to better understand the outcome of *in vivo*

therapy is to conduct controlled *in vitro* experiments with the relevant organisms. However, while

most of our understanding of the ecology and evolution of bacteria-phage interactions comes from

highly controlled *in vitro* environments (Gandon et al., 2008; Hampton et al., 2020), we do not

know how well *in vitro* dynamics mirror *in vivo* dynamics quantitatively or even qualitatively.

There are good reasons to expect the outcomes may be very different. *In vivo* environments

are expected to be more stressful for bacteria owing to limiting resources, presence of immune

cells and competing microbiota (Everest, 2007). Consequently, these conditions may differ

significantly from studies of bacteria-phage coevolution in nutrient rich microcosms which could

drive differences in (co)evolutionary dynamics and associated trade-offs (Gandon et al., 2008;

Hampton et al., 2020). In order to evolve phage resistance, bacteria must have sufficient mutation

rates supplied either by high resources (facilitating rapid growth) or population mixing (genetic

exchange) (Gandon et al., 2008; Hampton et al., 2020; Morgan et al., 2010, 2005; Pal et al., 2007).

Smaller population sizes *in vivo* may therefore limit resistance. Moreover, resistance costs and

trade-offs may be greater as a consequence of the additional selection pressures facing bacteria,

which may limit selection for resistance, making it easier for phage to overcome resistance, or

reduce growth rates and virulence (Buckling et al., 2006; Scanlan et al., 2015).

Consequently, the main aim of this research was to determine how well *in vitro* bacteria –

phage coevolutionary dynamics mirror dynamics in a clinical context. As a starting point, we

followed bacteria - phage coevolutionary dynamics during a phage-mediated decolonisation trial

in parallel with the dynamics between the same bacteria and phage *in vitro*. The nasal cavity can

act as a source of infection, notably from highly drug-resistant *Pseudomonas aeruginosa* and

Staphylococcus aureus when patients undergo hospitalisation and are treated with broad spectrum

antibiotics (Mainz et al., 2009; Wertheim et al., 2004). The trial was conducted to try and

decolonise intensive care (ICU) patients with *P. aeruginosa* and *S. aureus*, using a phage cocktail

prior to antibiotic treatment in prevention of ICU infections with these pathogens. Mucirocin (trade

name Bactroban) and chlorhexidine are routinely used for nasal decolonisation in such contexts,

but extensive use of these agents has led to resistance and new nasal antiseptics need to be

considered (Edgeworth, 2011). We focused on a single patient, in which we could detect *P.*

aeruginosa at the beginning of the trial and both *P. aeruginosa* and phage during treatment.

Decolonisation of *P. aeruginosa* was ultimately successful in this patient. We used the pre-

treatment *P. aeruginosa* isolates and evolved these populations *in vitro* in the absence and presence

of treatment phages. We measured bacteria-phage (co)evolution and while we found extensive

resistance evolution both *in vivo* and *in vitro*, we only found evidence of limited phage infectivity

evolution *in vitro*. We determined the consequences of resistance, with respect to virulence, and

the possible mechanisms linking this resistance – virulence trade – off using phenotypic (growth

rate, biofilm assays) and metagenomic analyses for the *in vivo* and *in vitro* populations. As we

only had access to samples from a single patient, we did not have a control group to understand

the impact of phage application frequency *in vivo*. However, we were able to do this in our *in vitro*

experiments, where we included two phage treatments: one where phages were added repeatedly

(mimicking multiple phage doses during *in vivo* phage treatment) and one where phage were added

once (at the start of the experiment). By comparing these treatments and to the *in vivo* treatment,

we were able to gain further insight into how multiple phage doses may affect phage therapy.

Methods

Ethics statement

The decolonisation study protocol was approved by the Leading Ethics Committee of the

"Université Catholique de Louvain" (Avis N°: B-403201111110). The study was performed in

accordance with the ethical standards as laid down in the Declaration of Helsinki and as revised in

2013. The patients gave informed consent and their anonymity was preserved.

Patient information, bacteria and phage strains and phage-bacteria evolution *in vivo*

Bacterial isolates were sampled from a patient as part of a clinical trial in Brussels, Belgium. The

patient (85 years old, male) had been admitted to an intensive care unit in 2015 for treatment of

burns (10% body area). Independent of his burns, a nasal colonisation of *P. aeruginosa* was

observed. Here, the patient was treated every day with a three-phage cocktail (BFC-1) for five

days, which consisted of two phages (14-1 and PNM) that could lytically infect *P. aeruginosa*.

The phage cocktail included a third phage that specifically infects *S. aureus* bacterium but was not

detected in this patient. This phage cocktail has been characterised to ensure no prophage elements

are present or that the phage carry toxin-encoding genes that may increase bacterial resistance

and/or virulence (Merabishvili et al., 2009). Furthermore, a purification process including the

removal of endotoxins in phage production was used to ensure such waste products cannot harm

patients (Merabishvili et al., 2009). As 14-1 infects via a lipopolysaccharide (LPS) (Betts et al.,

2014; Kropinski et al., 1977) receptor and PNM uses type IV pili (Ceyssens et al., 2011), resistance

to one phage is unlikely to result in cross-resistance and neither phage should compete for receptor

sites. Swabs of both nasal cavities were taken on day 0 (commencement of phage therapy), 2, 4, 7

and 13 and frozen at -80 °C in lysogeny broth containing 20% glycerol. Individual *P. aeruginosa*

isolates were obtained by plating swabs onto cetrimide selection plates. Colonies were picked and

inoculated into Luria broth (LB) and grown overnight, shaking at 180 r.p.m., at 37 °C to achieve

high cell densities. Additionally, swabs were soaked in LB broth to amplify phage overnight in the

same conditions. Phage was isolated from overnight cultures via chloroform extraction: 100 µL of

chloroform was added 900 µL of culture, vortexed and then centrifuged (14000 r.p.m., 1 min); the

supernatant was isolated and stored at 4 °C. Culture samples were frozen in 80% glycerol (final

concentration 40%) at -80 °C. Spot assays on lawns of *P. aeruginosa* P573 (a phage susceptible

control strain) confirmed the presence of phage at all time-points. Twelve individual plaques were

picked at each time-point and amplified with an overnight culture of P573 to obtain clonal isolates.

Individual isolates of *P. aeruginosa* were isolated from days 0 (n = 24), 2 (n = 8) and 4 (n = 2) for

use in this study and swabs were negative on days 7 and 13. Individual isolates were treated as

biological replicates for *in vivo* study.

Experimental *in vitro* phage-bacteria evolution

To examine whether findings *in vitro* are consistent with *in vivo* bacteria-phage (co)evolution,

phage therapy was replicated in the laboratory. We used two treatments: (1) phage added "once",

at the start of the experiment, and (2) phage added at every second transfer ("repeated" treatment).

The repeated phage treatment aimed to emulate the conditions of the clinical trial while the "once"

condition was used to assessed whether repeated doses change the observed dynamics. The phages

were identical to the two lytic phages (14-1 and PNM) used in the clinical trial. A control group in

which bacteria evolved in the absence of phage was set up and there were 6 biological replicates

in the control and each phage treatment.

155   Prior to experimental set-up, *in vivo* ancestral isolates (n = 24, same isolates as day 0) were

individually amplified in LB media and cultured at room temperature for two days. Cell densities

were estimated via optical density (wavelength 600 nm; OD_{600}) measures. Cultures were

centrifuged (14000 r.p.m., 3 min), the supernatant removed, and resuspended in M9 buffer to

equalise cell densities. These were added to a master-mix in which isolates were equally

represented in equal ratio (total 10^8 CFUs / mL; ~4 x 10^6 CFUs / mL of each isolate). 100 µL of

master-mix (10^7 CFUs; ~4 x 10^5 CFUs per isolate) was inoculated into 5 mL LB media. 100 µL of

phage was added to treatment microcosms at a total multiplicity of infection (MOI) of 0.5.

Microcosms were cultured, shaken (180 r.p.m.), at 37 °C. 1% transfers (50 µL culture into 5 mL

LB) took place every two days for a total of 12 days. At each transfer, culture samples were frozen

in glycerol, and in the phage treatments, phage were extracted as described previously. Cultures

were diluted and plated from frozen stocks onto LB agar and incubated for two days at 37 °C.

From each biological treatment replicate and time-point (days 4, 8 and 12), 16 bacterial colonies

were picked (96 total colonies) for time-shift assays to measure coevolutionary changes. An

additional 24 isolates were isolated at day 12 from each treatment and control group replicate for

bacterial virulence assays. Individual isolates are treated throughout as biological replicates nested

within their treatment (biological) replicate. Picked isolates were individually amplified in LB

media overnight before being frozen as described previously.

Measuring resistance-infectivity (co)evolution

To assess whether bacteria and phage had coevolved *in vivo* and *in vitro*, resistance assays were

conducted where bacteria were exposed to phage populations from the same/contemporary, past,

and future time-points (time-shift assays) (Buckling and Rainey, 2002). If bacteria and phage

coevolved, both bacteria and phage would show significant changes in resistance and infectivity,

respectively, through time. Increases in resistance and infectivity would be indicative of arms race

dynamics (ARD), while non-directional changes would be indicative of fluctuating selection

dynamics (FSD) (Lopez Pascua et al., 2014). Infectivity/resistance was determined by spot assays

in which 100 µL of overnight bacterial cultures was added to 5-8 µL soft LB agar overlay and 5

µL of each ancestral phage, or evolved phage mixture, was spotted onto the overlay. For *in vivo*

isolates, ancestral isolates (isolated pre-phage therapy) and bacterial isolates from days 2 and 4

were tested against the phage clonal isolates from days 2, 4 and 7. A bacterial isolate was scored

as susceptible (1 = susceptible, 0 = resistant) if at least one of twelve clonal phage isolates could

infect. *In vitro* isolates from days 4, 8 and 12 were tested (16 colonies per time-point) against phage

populations isolated from each time-point. In addition, we assessed the infectivity of both ancestral

phage against the ancestral bacteria (n = 24), the *in vivo* isolates (n = 8), and the *in vitro* populations

from day 12 (n = 24 per treatment replicate).

Virulence assays

Changes in bacterial virulence were measured using a *Galleria mellonella* infection model (Tsai

et al., 2016). Virulence of all ancestral (n = 24) and *in vivo* isolates (n = 10) were measured,

however two *in vivo* isolates (from day 2 of phage therapy) were excluded from analyses as time

of death was not accurately determined. As few isolates remained phage susceptible (n = 3), these

were pooled with phage resistant isolates to determine a change in virulence from pre- (ancestral,

n = 24) to post-phage treatment (n = 8). For *in vitro* isolates, where possible, one resistant and one

susceptible (to ancestral phage) isolate were selected from each treatment replicate to give a total

of 12 resistant and 12 susceptible isolates, with 14 isolates from the "phage added once" and 10

isolates from the "phage added repeatedly". As some treatment replicates contained no susceptible

isolates, multiple resistant isolates were chosen (Table S1). Seven *in vitro* isolates evolved

resistance to one phage only (single resistance), therefore their virulence was measured to assess

whether changes in virulence were mediated by resistance to one or both phages. The virulence of

in vitro isolates was contrasted to six isolates from the control group isolated from 6 independent

replicates.

For each bacterial isolate (total of 56 biological replicates), 20 *G. mellonella* larvae (20

technical replicates) were infected with 10 µL (~100 CFUs) of overnight culture (cells resuspended

in M9 buffer) using a sterile syringe. Twenty *G. mellonella* larvae were injected with 10 µL of M9

buffer as a technical control. Survival of each larvae was assessed every two hours for up to 58

hours.

Growth rate assays

We measured bacterial growth rates as a potential mechanism underlying virulence variation and

cost of resistance for ancestral and all evolved isolates (*in vitro* and *in vivo*) tested in the virulence

assay. Growth curves were measured in a 96 well plate with 10 µL of culture (1.0×10^5 CFUs)

from each isolate inoculated into individual wells containing 140 µL LB broth. 26 wells remained

as blank controls. The plate was incubated in an OD reader (Biotek Instruments Ltd) at 37 °C in

shaken conditions and OD_{600} was measured every 20 mins for 24 hours to estimate changes in

cellular density. Growth curves were independently replicated 3 times (technical replicates).

Biofilm assay

Biofilm production of isolates isolated for virulence and growth rate assays was measured using a

resazurin assay, based upon a previously described protocol (Peeters et al., 2008). Overnight

monocultures were diluted to 0.1 OD_{600} (~10^9 CFU/mL) into M9 buffer. 100 µL of each culture

was inoculated into a sterile 96 well plate and incubated at 37 °C for 4 hrs for cellular adhesion.

Then, the supernatant was removed, and wells were washed gently with 100 µL of M9 buffer by

pipetting up and down 3 times. 100 µL of fresh LB media was added to each well including 25

controls (no bacteria) and plates were incubated for 16 hrs. Liquid was removed and wells were

washed with 100 µL M9 buffer. 100 µL of fresh M9 was added followed by 10 µL resazurin

solution (final resazurin concentration: 100 µM). Fluorescence (λex: 560 nm and λem: 590 nm)

was measured after 1 hr incubation. Biofilm assays were independently replicated 3 times

(technical replicates).

Sequencing

We sequenced the genomes of the bacterial isolates used for the *in vitro* experiment to establish to

what extent evolution was driven by *de novo* mutation or selection on existing genetic variation.

This is especially pertinent as the experiment did not start from a clonal population, rather from

24 isolates (which included an isolate that was resistant to one phage). At the end of the

experiment, we picked 24 isolates, extracted DNA from isolates individually and performed whole

genome sequencing (WGS) on pools of the bacterial isolates that were isolated from each replicate

(pool-seq). Total DNA extraction (19 pooled-isolate samples) was performed using the "Qiagen

QIAmp DNA Mini Kit" following the manufacturer's instructions. An Illumina HiSeq 2000

sequencer was used to generate 100 bp paired reads from a 500 bp insert library.

Reads were trimmed for the presence of Illumina adapter sequences and for quality scores

below 10 using *TrimGalore* (v0.6.4_dev, powered by *Cutadapt*). *TrimGalore* also removed any

paired end reads when one of the paired end reads was shorter than 20 bp. Trimmed reads were

mapped to the *P. aeruginosa* UCBPP-14 reference genome (Locus: NC_008463) with *bwa-mem*

(v0.7.17-r1188), and duplicates were marked using *samblaster* (v0.1.24). Subsequent bam files

were split into mapped and unmapped reads, sorted and indexed using *samtools* (v0.1.19-

44428cd). Across all files, ~93% of the reads mapped to the reference genome (minimum 92.5%),

giving on average ~12,000,000 reads per replicate (minimum 8,807,569). Variants were called

identified using *freebayes* (v1.3.2-dirty) with ploidy set to 24 (*-p 24*) and we assumed pooled

samples with a known number of genome copies (*--pooled-discrete*). Due to the high ploidy of the

pooled samples, we reduced the number of alleles that are considered at the cost of sensitivity to

lower frequency alleles at multiallelic loci by setting *--use-best-n-alleles 4*.

Next, we compiled the vcf files (standard output of SNP callers) into *R* using the package

vcfR and kept only the variants which were sequenced at a depth within two standard deviations

of the mean. Variants were further filtered to retain only those with quality scores >30. We used

the two replicate sequencing runs of the same samples to conservatively identify variants; variants

were only kept if they occurred in both sequencing runs and if there was less than 0.3 difference

in frequency between runs. We then only kept variants that had different frequency than the

ancestor (> 1/24 difference in frequency). To identify significant differences in genetic variants

(SNPs and indels) between treatments, we first filtered unique that were present in at least three

replicates of any treatment. This identified 960 potential variants. We then performed a Wilcoxon

test on each variant, with frequency as the response variable and treatment (control, phage added

once, and phage added repeatedly) as the predictor. Significant genetic variants were identified if

the adjusted p value (*fdr* method) was < 0.05. Of these significant changes, 160 were existing

genetic variants and 120 were novel mutations. We then checked where each variant was in the

genome and only retained variants that were occurred in a gene. After these steps, 50 genetic

variants were identified in genes of known function, and 236 in genes of unknown function.

For the *in vivo* isolates, DNA extraction and whole genome sequencing on each isolate

individually by MicrobesNG using in-house protocols ("MicrobesNG Sequencing Service

Methods v20210419," n.d.). Whole genome sequencing was carried out on each isolate

individually by MicrobesNG using Illumina sequencers (HiSeq/NovaSeq) to create 250 bp paired

end reads. In-house processing by MicobesNG trimmed the adapters and removed reads with a

quality < 15. We then processed the reads as for the pooled sequencing, except ploidy was set to

1 when using *freebayes*. On average, 81% of reads mapped back to the reference genome, but there

was more variation as compared to the *in vitro* experiment, with a minimum of 44% and a

maximum of 94%. As we only had isolates from a single patient, there was limited statistical

analysis we could do, so we primarily concentrated on whether the significant changes found *in*

vitro were also present in the *in vivo* isolates.

Statistical analyses

All data was analysed using R (v. 4.0.3) in RStudio (Team, 2013) and all plots were made using

the package "*ggplot2*" (Wickham, 2016). Model simplification was conducted using likelihood

ratio tests and Tukey's post-hoc multiple comparison tests were used to identify the most

parsimonious model using the R package "*emmeans*" (Lenth, 2018). First, we established whether

bacteria and phage had coevolved *in vivo* and in each *in vitro* treatment using general linear mixed

effects models. Models were fitted using the package "*lme4*" (Bates et al., 2014). *In vivo* isolates

from day 4 were excluded as their small sample size (n = 2) prevented model convergence. For *in*

vivo isolates: infection (1 = susceptible, 0 = resistant) was analysed against phage time-point (fixed

effect) with a binomial error structure. A random effect of "isolate" was included to account for

the same bacterial isolate being used in infectivity assays across each phage time-point. For *in*

vitro analyses, the proportion of isolates infected was analysed against fixed effects of phage and

bacterial time-points with separate models for each phage treatment. To see if there were any

differences in phage resistance to ancestral phage or *in vitro* phage isolates at the end of the

experiment, the proportion of isolates infected was analysed against fixed effects of phage

treatment and phage type (ancestral or end – point phage). All *in vitro* models included a random

effect of "replicate" (which treatment replicate isolates had been isolated from) with a binomial

error structure.

Virulence of bacterial isolates was analysed using survival curves from the survival of

Galleria infected with different isolated isolates. Survival curves were fitted using Bayesian

regression models using the package "*rstanarm*" (Goodrich, 2020) and parameters were estimated

using "*tidybayes*" (Kay and Mastny, 2020). For all models, we fit a proportional hazards model

with an M-splines baseline hazard. Models were ran for 3000 iterations and 3 chains were used

with uninformative priors. Model convergence was assessed using Rhat values (all values were 1)

and manually checking chain mixing. For all models, log hazards were estimated for each

treatment value and hazard ratios were calculated as the exponential of the difference between two

log hazards. A hazard ratio below one indicates a decrease in virulence compared to the baseline

treatment, with a value above one indicating an increase in virulence. Median hazard ratios with

95% credible intervals that do not cross 1 indicate a significant difference in virulence between

two factors. Proportion of *Galleria* that died and mean time to death of *Galleria* that died were

calculated as summary statistics for each treatment.

For virulence analyses of *in vivo* isolates, treatment (pre- or post-phage therapy) was the

only predictor of survival, with a random effect of "isolate" added to account for multiple *G.*

mellonella larvae being infected by the same isolate. For the *in vitro* isolates, survival was analysed

against interacting fixed effects of resistance (resistant, single resistant, susceptible) and treatment

(no phage added, phage added once, phage added repeatedly). Here, an additional random effect

of "replicate" was included to account for clonal non-independence.

Biofilm productivity of *in vitro* isolates was estimated using linear mixed effects models.

Biofilm production (fluorescence 560/590nm) was log transformed to account for the data being

abundance-based. Biofilm production was averaged over each technical replicate for each

biological replicate (individual bacterial isolates). Biofilm production was used as the response

variable, and explained with interacting fixed effects of treatment and phage resistance, while

treatment replicate was included as a random effect. One outlier was removed each from the

susceptible and control groups as these were likely contamination (full data cleaning process and

impact on parameter estimates explained in the SI and effects of outlier removal on estimates

shown in Tables S3 & S4). Due to the low number of isolates, biofilm productivity of *in vivo*

isolates from before and after phage treatment as well as phage susceptible and resistant isolates

were estimated using a non-parametric bootstrap using the R package "*boot*" (Canty and Ripley,

2021). Non-overlapping 95% CIs were interpreted as being significantly different.

Before growth rate analyses, data points from the first 2.5 hrs were removed as a sharp

increase in OD was observed before lag phase independent of exponential growth (Figure S1).

Growth rates were estimated using a rolling regression taking the steepest slope of the linear

regression between *ln* OD_{600} and time in hours in a shifting window of every 7 time points (every

2.3 hrs). Growth rate was averaged over each technical replicate for each biological replicate

(individual bacterial isolates). For *in vitro* isolates, growth rate was analysed against treatment and

resistance in a linear mixed effects model with a random effect of treatment "replicate". Growth

rates of isolates pre- and post-phage treatment, and resistant and susceptible *in vivo* isolates were

compared using 95% confidence intervals (CIs) that were estimated using a non-parametric

bootstrap.

To look at genetic differences *in vitro*, we calculated several commonly used metrics using

the frequency data of the 50 variants in known genes that differed significantly between our

treatments: (1) the genetic distance from the ancestral population, calculated as the sum of the

difference of the proportion of each SNP/indel in each population from the ancestral proportion;

(2) the number of unique *de novo* SNPs/indels in each population; (3) alpha diversity, calculated

using a modified version of the Hardy-Weinberg equilibrium, such that $\alpha = \sum(1 - p_i^2 - q_i^2)$,

where i is the position of each SNP/indel, p is the proportion of the SNP/indel and q is $1 - p$. This

is equivalent to expected heterozygosity. Differences between these metrics were analysed using

linear models. To test whether consistent genetic differences occur within treatments, we

performed non-metric multidimensional scaling on the Euclidean distance matrix of SNPs/indels

and their proportions in each population using "*metaMDS*" in the R package "*vegan*". Nonmetric

multidimensional scaling aims to collapse information from multiple dimensions (*i.e.*, from

different populations and multiple SNPs/indels per population) into just a few, allowing

differences between samples to be visualized and interpreted. Permutational ANOVA tests were

run using "*vegan::adonis*", with Euclidean distance as the response term and treatment as the

predictor variable with 9999 iterations. Changes in beta-diversity were examined using

"*vegan::betadisper*" using the same response and predictor variables as in the PERMANOVA.

For the *in vivo* sequencing, we looked for the presence of variants found to be significant

in vitro in the *in vivo* isolates. To see if any of the genetic changes could explain differences in

phenotypes of the isolates, we did non-metric multidimensional scaling as described previously.

Phenotypic traits of the isolates (virulence, growth rate, resistance profile, and biofilm production)

were then mapped onto the resulting NMDS plot to see whether observed genetic changes

correlated with any changes in phenotype (Figure S2).

Results

Resistance and infectivity (co)evolution

In vivo

Resistance to phage can evolve via *de novo* mutations or selection on pre-existing resistance and

we therefore tested the resistance of ancestral bacterial isolates to ancestral phage and pooled phage

isolates isolated *in vivo*. Of the 24 ancestral isolates isolated *in vivo* at day 0, 23 were susceptible

to both the ancestral phage (14-1 and PNM) while one was resistant to 14-1 but susceptible to

PNM. In subsequent tests of phage resistance to phage isolates isolated *in vivo*, all ancestral

bacteria were susceptible to phage (Figure 1a). We considered whether bacteria and phage

coevolved *in vivo* using time-shift assays to determine changes in resistance and infectivity through

time. After two days, we saw evidence of resistance evolution, with more than half of bacterial

isolates being resistant to ancestral phage. 1 of 2 isolates isolated at day 4 were phage resistant and

no *P. aeruginosa* was detected from nasal swabs after this time point. Although both bacterial

populations from days 2 and 7 showed observationally higher phage resistance (87.5% day 2,

100% day 4) against day 2 phage, this was non-significantly different to resistance shown to

ancestral phage and phage from days 4 and 7 (Tukey HSD: $p > 0.05$; Table S2; Figure 1a). This

non-significance is likely due to low statistical power from limited clonal numbers (only 8 isolates

from day 2) and the consistency of resistance levels in bacterial populations from day 2 apart from

to its contemporary phage; overall, results showed equal levels of phage resistance over time *in*

vivo. The absence of changes in phage infectivity through time did not warrant further investigation

of the population dynamics of the two inoculated phages.

Figure 1. Time-shift assays of bacterial evolution and phage during phage therapy. The proportion of phage susceptible bacterial isolates from (a) ancestral (pre-phage therapy, "Anc."), and days 2 (T1) and 4 (T2) of *in vivo* (clinical) phage therapy (n = 8) tested against ancestral phage ("anc.": 14-1 and PNM) and phage isolates from days 2 (T1), 4 (T2) and 7 (T3). The same ancestral bacterial isolates are evolved *in vitro* with phage added either (b) once, at the start of the experiment for comparison to (c) in which phage are added repeatedly, emulating clinical treatment. For (b) and (c) the proportion of bacterial isolates susceptible to phage at three time-points (T1 - T3: days 4, 8 and 12 of coevolution) is presented with small points as independent treatment replicates. Resistance of bacteria isolated at T3 to ancestral phage ("Anc.": 14-1 and PNM) is indicated within the shaded region, note that these isolates are distinct from those tested against phage from T1-T3. Filled points are the mean proportion of resistant isolates across replicates. Bars = ±SE.

In vitro

We used the same bacterial isolates and phages to determine how well the *in vivo* evolutionary

dynamics paralleled *in vitro* experiments in the lab. To assess the impact of phage application

schedule, we added phages once at the start of the experiment, repeatedly (as was the case *in vivo*)

or not at all. The dynamics in the phage treatments largely mirrored the *in vivo* dynamics.

Resistance to phage increased through time, reaching comparable levels to that observed *in vivo*

(>50%). While phage resistance increased in both phage treatments (ANOVA comparing models

with and without bacterial time: phage added: once $\chi_2^2 = 79.08$, p < 0.001; repeatedly $\chi_2^2 = 45.30$,

p < 0.001; Figure 1b, c), there was a significantly higher proportion of phage resistant bacteria in

the treatment where phages were added once (90.81%, ±SE = 0.0003) compared to when phages

were added repeatedly (66.7%, ±SE = 7.64; Tukey HSD comparing phage added once and

repeatedly treatments: estimate = -1.63, z-ratio = -3.75, p = 0.0002; Figure 1b, c). All isolates from

the control condition were susceptible to ancestral phages.

The lower resistance of bacteria when phages were added repeatedly was driven by

evolution of increased phage infectivity. When phages were added once, resistance to ancestral

phage (14-1: 88.64%, ±SE = 2.77; PNM: 90.15%, ±SE = 2.60) did not differ relative to phage

isolated at the end of experimental evolution (90.0%, ±SE = 4.27; Tukey HSD comparing end-

point phage resistance to contemporary phage to resistance to 14-1 (p = 1.0) and PNM (p = 0.938;

Figure 1b). However, when phages were added repeatedly, resistance to ancestral phage (14-1:

94.41%, ±SE = 1.93; PNM: 96.5%, ±SE = 1.54) was significantly higher than to phage isolated at

the end of experiment (61.5%, ±SE = 12.5), indicative of increased phage infectivity (Tukey HSD

comparing end-point phage resistance to contemporary phage to resistance to 14-1 (p < 0.001) and

PNM (p < 0.001); Figure 1c). We saw no further evidence of phage evolution in either treatments,

with phage resistance being independent to the time point from which phage were isolated during

experimental evolution (ANOVA comparing models with and without phage time: phage added:

once $\chi_2^2 = 0.41$; p = 0.814; repeatedly $\chi_2^2 = 0.68$; p = 0.713; Figure 1c). This suggests the increase

in phage infectivity evolved before the first sampling time point (the second transfer, or 96 hours).

The absence of any further changes in phages during experimental evolution did not warrant

further investigation of the population dynamics of the two inoculated phages.

Phage resistance-virulence trade-offs

We next assessed whether phage application altered the virulence of *P. aeruginosa* populations *in*

vivo and *in vitro* using a *G. mellonella* infection model. Here, virulence can be calculated as the

inverse of *G. mellonella* survival over time: the more virulent the bacterial isolate, the shorter the

time until death. *In vivo*, phage therapy reduced bacterial virulence by 95.5% on average (median

hazard ratio = 0.045, 95%CI = 0.012 – 0.163). When inoculated with bacteria isolated before phage

therapy, 479 (99.7%) of *G. mellonella* died within 46 hours, with death occurring at a mean time

of 15 hours (Figure 2a). In contrast, after phage therapy, bacteria killed only 70% of *G. mellonella*

within 50 hours, with mean time of death occurring at 20.5 hours (Figure 2a).

We also saw a reduction in virulence *in vitro* when populations had evolved with phage

compared to without phage. Bacteria in the phage added once and phage added repeatedly

treatments were 96.15% (median hazard ratio = 0.0385, 95%CI = 0.011 – 0.124) and 89.8%

(median hazard ratio = 0.102, 95%CI = 0.0294 – 0.355) less virulent on average compared to the

no-phage control populations (Figure 2b). In the control populations, 96 (80%) of the *G. mellonella*

died within 47 hours, with death occurring at a mean time of 19 hours. In contrast, only 32.5% and

41.2% of *G. mellonella* died in the phage added once and phage added repeatedly treatments

respectively, with a mean time to death of 36.7 and 35.4 hours (Figure 2b).

This difference was driven by two impacts of phage application. First, there was a huge

reduction in virulence of bacteria that were resistant to both phages. When resistant to both phage,

virulence reduced by 95.69% compared to susceptible bacteria (median hazard ratio = 0.0431,

95%CI = 0.0139-0.122) and was 93.33% lower than bacteria resistant to only a single phage

(median hazard ratio = 0.0663, 95%CI = 0.0179-0.247; Figure 2c). Meanwhile, there was no

reduction in virulence for bacteria resistant to just a single phage (median hazard ratio = 0.653,

95%CI = 0.185 -2.09) when compared to phage susceptible bacteria. Second, even susceptible

bacteria became less virulent after phage application, with a 91.14% median hazard ratio = 0.0886,

95%CI = 0.0246-0.344) reduction in the phage added once treatment and an 80.7% reduction when

phage was added repeatedly (median hazard ratio = 0.193, 95%CI = 0.04-1.06; Figure 2c).

Figure 2. Survival curves of *G. mellonella* inoculated with *P. aeruginosa* isolates isolated (a) *in vivo* and (b, c) *in vitro*. (a) Virulence of bacterial isolates isolated from before (black) and after (grey) phage therapy. (b) Virulence of bacterial in control populations where no phage is present (yellow) is much greater than populations which were exposed to either a single (orange) or repeated (red) phage applications. (c) Virulence is significantly reduced in bacterial isolates resistant to both phages while bacteria resistant to one phage are non-significantly different in virulence to phage susceptible isolates. Lines represent the average prediction and shaded bands represent the 95% credible intervals of those predictions. Proportion and mean time to death of *Galleria* that died are presented as summary statistics for each treatment.

Growth rate

Evolving resistance against phage can trade-off with growth rates (Buckling et al., 2006; Koskella

et al., 2011; Wright et al., 2018), and could be a plausible mechanism for virulence reduction. In

the case of the *in vivo* isolates, the growth rate of pre-phage treatment isolates ($\bar{x} = 0.36$, 95%CI =

0.342 - 0.369) was significantly higher than that of post-treatment isolates ($\bar{x} = 0.25$, 95%CI =

0.195 - 0.302). Post-treatment isolates which evolved resistance (to one or both phage) had a

significantly lower growth rate ($\bar{x} = 0.21$, 95%CI = 0.162 - 0.266) than the ancestral isolates ($\bar{x} =

0.36, 95%CI = 0.342 - 0.369). However, the 95% CIs of the susceptible isolates' growth rate ($\bar{x} =

0.33, 95%CI = 0.256 - 0.402) overlapped with the ancestral and resistant isolates, indicating non-

significance, which was likely driven by a lack of statistical power (n = 3 susceptible isolates)

(Figure 3a).

For *in vitro* isolates, growth rate was significantly affected by treatment ($\chi^2_2 = 20.14$, p <

0.001) with the control group having significantly higher growth rates ($\bar{x} = 0.43$, 95%CI = 0.373 -

0.477) compared to treatments in which phage were added once ($\bar{x} = 0.30$, 95%CI = 0.256 - 0.346;

p = 0.002) or repeatedly ($\bar{x} = 0.276$, 95%CI = 0.235 - 0.318; p < 0.001; Figure 3b). No significant

difference was observed in growth rates between each phage treatment (Tukey HSD: estimate =

0.025, t-ratio = 0.95, p = 0.628); and changes in growth were not associated with phage resistance

(interaction between treatment and resistance: $\chi^2_2 = 0.37$, p = 0.832; fixed effect of resistance: $\chi^2_2 =

2.98, p = 0.226; Figure 3b).

Figure 3. Phenotypic changes underpinning the resistance – virulence trade-off of bacterial

populations. Populations were evolved (a, c) *in vivo*, and (b, d) *in vitro*. (a, b) Changes in bacterial growth

were estimated for populations isolated (a) before and after phage treatment, and (b) between control and

the phage treatment groups. Single points represent individual bacterial isolates of differing phage

resistance levels: resistant to both phage (light blue), resistant to 14-1 or PNM (dark blue) or phage

susceptible (red). Tops and bottoms of the bars represent the 75th and 25th percentiles of the data, the

middle lines are the medians, and the whiskers extend from their respective hinge to the smallest or largest

value no further than 1.5* interquartile range.

Biofilm production

Biofilm production is often altered in phage resistant isolates (Harrison et al., 2015; Kim et al.,

2015; Scanlan and Buckling, 2012) and this can also affect virulence and growth rate (Li et al.,

2008; Wang et al., 2011). Therefore, we compared the biofilm productivity between isolates

evolved *in vitro* and *in vivo*. For *in vivo* isolates, there was no significant difference in biofilm

production between isolates isolated pre- ($\bar{x} = 5.53$, 95%CI = 5.45 - 5.6) or post-phage therapy

($\bar{x} = 5.39$, 95%CI = 5.27 – 5.5), and between phage resistant ($\bar{x} = 5.38$, 95%CI = 5.27 – 5.50) and

susceptible isolates ($\bar{x} = 5.39$, 95%CI = 5.1 – 5.69; Figure 3c).

For *in vitro* isolates, there was no significant difference in biofilm production between

treatments (treatment x resistance interaction: $\chi^2_2 = 0.785$, p = 0.675; treatment: $\chi^2_2 = 4.52$, p =

0.104). However, phage resistance was associated with increased biofilm production ($\chi^2_2 = 19.12$,

p < 0.001), with phage susceptible isolates ($\bar{x} = 5.35$, 95%CI = 5.16 – 5.54) having lower biofilm

production than isolates resistant to one ($\bar{x} = 5.85$, 95%CI = 5.56 – 6.13; p = 0.012) or both phages

($\bar{x} = 5.93$, 95%CI = 5.73 – 6.13; p = 0.004; Figure 3d).

Genetic basis of evolutionary changes

526   Given the parallel phenotypic differences we observed under *in vivo* and *in vitro* conditions, we

527   next explored whether the underlying genetic changes were the same for *in vitro* and *in vivo*

528   evolved populations. From the *in vitro* pooled sequencing, we found significant changes in 27

529   different genes with known function (Figure 4), 15 of which were changes in frequency compared

530   to the ancestral population while 12 were novel mutations.

Figure 4. Significant genetic changes identified *in vitro* and *in vivo*. The frequency of genes that differed

significantly between treatments *in vitro* are shown. Genetic changes that were already present in the

ancestral population (first column) are shown in blue, while *de novo* mutations are in orange. Only genes

whose function was identified are included. Multiple rows of the same gene indicate different genomic

variants of the same gene. Colour intensity shows how frequent the genetic change is in the population. For

in vitro treatments, each column is an independent replicate representing the results of the pooled

sequencing. For the *in vivo* isolates (final column), all isolates were pooled to visualise the frequency of

each genetic change *in vivo*.

In terms of genetic distance from the ancestral population, populations in the treatment

where phage was added repeatedly diverged more than when phage was added once (Tukey HSD:

t = -4.32, d.f. = 15, p = 0.0017), which diverged more from the ancestral populations than the

control populations (Tukey HSD: t = -8.81, d.f. = 15, p < 0.0001; Figure 5a, b). Consistent with

this, more novel SNPs/indels were found in the treatment populations where phages were added

repeatedly (mean = 19.2) compared to where phages were added once (mean = 11.8, Tukey HSD:

t = -4.80, d.f. = 15, p = 0.0006), which in turn had more novel SNPs/indels than control populations

(mean = 2.5, Tukey HSD: t = -6.105, d.f. = 15, p = 0.0001; Figure 5b). Within-population diversity

was highest in phage added repeatedly treatment (Tukey HSD phage added repeatedly *vs.* phage

added once: t = -5.96, d.f. = 15, p = 0.0001; Tukey HSD phage added repeatedly *vs.* control: t = -

4.33, d.f. = 15, p = 0.0016), but there was no difference between control populations and phage

added once treatment (Tukey HSD: t = -1.14, d.f. = 15, p = 0.2625; Figure 5c).

(a) NMDS of euclidean distances between genome changes of known function

Figure 5. Genomic changes for populations evolved *in vitro*. (a) The divergence of each treatment group as indicated by Euclidean distances of each gene from the ancestral isolates. Genetic changes were characterised as a change in frequency of an existing gene in the ancestor or selection on a mutation that was gained during experimental evolution. The presence of phage significantly affected (b) genotypic distance from the ancestor, (c) the number of new SNPs and (d) alpha diversity.

In vitro treatment significantly altered genetic composition (i.e. the centroids of phage

added once, phage added repeatedly and control populations were different, Figure 4,

PERMANOVA, $F_{3,15}$ = 16.05, R^2 = 0.76, p = 0.0001). There was no difference in beta-diversity

(calculated from distance-to-centroids between groups) between treatments (homogeneity of

multi-variation dispersion ANOVA: $F_{3,15}$ = 1.65, p = 0.22). Genetic changes associated with both

in vitro phage treatments, but not control, were flagellar hook-associated protein (*flgK*) and rod-

shape determining proteins (*mreC* and *mreD*) (Figure 4). Additionally, genetic changes in DNA

mismatch repair (*mutS*), elongation factor G (*fusA*), signal recognition particle (*ffs*), type 3

secretion system pilotin (*yscW*), and pilus assembly protein (*pilB*) were mainly found *in vitro* in

the phage added repeatedly treatment (Figure 4). We examined whether these same mutations

arose in the clonal *in vivo* isolates. Of the 50 genetic changes identified *in vitro*, the *in vivo* isolates

had a mean of 11.8 changes (maximum = 23, minimum = 1). Changes in frequencies or mutations

were found at 16/27 of the same genes identified *in vitro* in at least one isolate (Figure 4). Crucially,

some of the *de novo* mutations that were at high frequency in vitro were also at high frequency in

vivo (*flgL* = 30%, *mutS* = 45%, *mreC* = 50%, *mreD* = 30%, *pilB* = 40%). We next assessed whether

mutations arising in individual isolates *in vivo* could be linked to phenotypic changes (resistance,

virulence, biofilm productivity, growth rate). If strong genotypic – phenotype links can be made,

we would expect isolates to cluster according to their trait value. However, we found no obvious

genotype to phenotype links *in vivo* and due to the small number of isolates, we were unable to

conduct robust statistical analyses (Figure S2).

Discussion

Highly parallel evolutionary dynamics were found *in vivo* and *in vitro*, showing that laboratory

studies can be predictive of clinical phage therapy (phage-mediated decolonisation). Levels of

phage resistance were comparable *in vivo* and *in vitro* with limited phage infectivity evolution

occurring only *in vitro*. Additionally, phage resistance was associated with trade-offs with

virulence and bacterial growth and similar mutations or genetic variants could explain these

parallel evolutionary changes in both ecological contexts.

Parallel evolution is promising for future clinical work in which the trajectory of bacteria

and phage (co)evolution may be predicted for individual patients. Given the difference in

environments between that of a laboratory and human nasal cavity, it was surprising that evolution

was so similar. The nasal cavity would have been more nutrient limited and spatially heterogenous

and such conditions have been shown to change bacteria – phage interactions compared to

evolution in laboratory growth medium. For instance, bacteria – phage coevolution is altered in

soil compared to growth medium because of increased costs of resistance in the former (Gomez

and Buckling, 2011); and in plant leaves, bacteria and phage have failed to evolve while otherwise

coevolving in the laboratory due to lower bacteria – phage contact rates (Hernandez and Koskella,

2019).

Rapid phage resistance has been highlighted as a concern for phage therapy development

(Torres-Barceló, 2018) and was found in this case study. Without ongoing reciprocal phage

evolution, phage resistance is likely to lead to regrowth of bacteria during infections; this has been

found in a few case studies but has been combatted by selecting a different phage (single or

cocktail) (Aslam et al., 2020; Schooley et al., 2017; Wu et al., 2021; Zhvania et al., 2017). That

resistance emerged predominantly by *de novo* mutations and not selection on existing variation

further highlights the fact that stressful *in vivo* conditions need not prevent bacteria from rapidly

adapting to phage. However, phage resistance was found to have growth rate costs which may

further explain why some isolates remained sensitive to phage, and why decolonisation was

ultimately successful. The natural nasal microbiota may have further contributed to this success

by outcompeting slower growing phage resistant isolates (Wang et al., 2019). Such resistance –

virulence trade-offs have been found in previous *in vitro*, animal model and clinical studies and is

expected to improve treatment success in spite of phage resistance (Oechslin, 2018). Phage

resistance however is not always costly and likely depends on the genetic background of the

bacteria and phage, as well as the environmental context (Meaden et al., 2015; van Houte et al.,

2016). The absence of clear links between resistance phenotypes and growth rate within our *in*

vitro treatments, despite clear between-treatment effects, is presumably because the cultures were

initiated with genetically diverse clinical isolates, rather than a single ancestral clone. Ultimately,

phage resistance was beneficial in controlling the pathogen in this case study and such benefits

could have led to treatment success.

Population genomics of the isolates *in vivo* and *in vitro* revealed parallel genomic changes

and the effects of phage selection at a molecular level. In line with previous studies, phage selection

accelerated molecular evolution which resulted in populations being more diverged from their

ancestral pool than control isolates (*in vitro*) (Morgan et al., 2010; Pal et al., 2007). This suggests

that phage placed strong selection on bacteria and this further resulted in mutator strains (*mutS*

genotype), especially where phage were added repeatedly *in vitro* and *in vivo*, that likely

accelerated the rate of evolution further (Morgan et al., 2010). Between both *in vivo* and *in vitro*

results, mutations selected by phage gave insight into potential mechanisms of resistance. In

particular, mutations to *mreD* and *mreC* encoded for cell shape (Cabeen and Jacobs-Wagner, 2005)

while missense mutations at *flgK* and *flgL* probably resulted in flagella loss (Macnab, 2003). The

result of these mutations was likely reduced cell surface area which reduces the probability of

phage absorption (Dennehy and Abedon, 2021). *pilB* mutations were also selected which is likely

a surface mutation to prevent PNM (a pilus binding phage) from attaching (Ceyssens et al., 2011).

Strong genotype – phenotype links were difficult to distinguish as a high number of the

same mutations or existing variants were selected upon both *in vivo* and *in vitro*, and no unique

candidate genes were obvious to explain phenotypic differences (i.e. growth rate, biofilm

production). Flagella and *mreC* mutations are associated with increased biofilm production

(Whiteley et al., 2001) and as they were found in the *in vitro* isolates, this may suggest a

generalistic phage defence strategy (Harper et al., 2014). However, as these same mutations were

found *in vivo*, without increases in biofilm production, strong genotype – phenotype links are

unclear. Surface mutation is a common form of phage resistance (van Houte et al., 2016) but here

it appears coupled with mechanisms that prevent absorption more generally; combined, these

strategies could have reduced bacterial growth and/or virulence by impairing swarming motility

(Haiko and Westerlund-Wikström, 2013). In contrast, other studies using the same phage but *P.*

aeruginosa PA01 instead reported mutations predominantly at pilus and lipopolysaccharide

receptors, showing that greater selection is placed on specific phage resistance strategies at phage

binding sites (Wright et al., 2019, 2018). Our results highlight the importance of genetic

background, and hence the use of clinical isolates over laboratory clones, in determining

evolutionary trajectories (Betts et al., 2014; Friman et al., 2016; León and Bastías, 2015; Rossitto

et al., 2018; van Houte et al., 2016).

How phages were applied *in vitro* had an impact on phage resistance. Although we

observed a greater number of mutations where phages were added repeatedly (mimicking multiple

phage doses), the proportion of resistant bacteria at the end of the experiment was significantly

lower (61.5%) compared to where phages were just added at the start of the experiment (90%).

That resistance of evolved bacteria to the ancestral phages (~90%) did not differ between

treatments suggests that differences in resistance to evolved phages was caused by phage

infectivity evolution in the repeated phage treatment. This infectivity evolution occurred before

the first time point, with no subsequent changes, and was presumably the result of greater mutation

supply rate in this repeated phage application treatment. Multiple (Manohar et al., 2018) and higher

(Barrow et al., 1998; Biswas et al., 2002; Debarbieux et al., 2010; Merril et al., 1996; Sunagar et

al., 2010; Vinodkumar et al., 2008; Wang et al., 2006; Wills et al., 2005) phage doses have been

shown previously to decrease morbidity and mortality in animal models. Our results build upon

this work to show that maintaining high phage titres may have a positive effect in reducing resistant

populations despite increasing the strength of selection for resistance.

Although our results suggest that results of laboratory studies may parallel phage resistance

– virulence trade-offs during *in vivo* phage therapy, there a number of important caveats. Firstly,

our *in vitro – in vivo* results were comparable with regards to the direction of change, but absolute

effect sizes were greater *in vitro*. For instance, we observed a higher proportion of phage resistant

isolates *in vitro* that were also far less virulent than isolates found *in vivo*. Therefore, it may be

expected that phenotypic changes will be exaggerated *in vitro*. Secondly, as this patient had a nasal

colonization, the role of the adaptive immune system would have been heavily reduced (Ooi et al.,

2008). The immune system has the potential to improve phage therapy success by synergistically

killing bacteria with phage which may diminish the potential for phage resistance by reducing

bacterial growth (Leung and Weitz, 2017; Roach et al., 2017). Alternatively, immune cells may

kill phage and therefore prevent phage acting on bacterial populations (Van Belleghem et al.,

2018). Additionally, the nasal cavity has a diverse microbiome that may have contributed to *P.*

aeruginosa decolonisation (Hu et al., 2018; Kumpitsch et al., 2019). The role of the microbiome

will differ in infection type, with gut and respiratory infections potentially having a greater

microbiome effect (Hu et al., 2018; Kumpitsch et al., 2019; Lloyd-Price et al., 2016) while septic

infections may have a greater immune system effect (Van Belleghem et al., 2018). Therefore, it is

important to compare laboratory vs *in vivo* phage therapy in a number of infection types. The

genomic background of the bacteria may further affect how sensitive its resistance strategies are

to the environment it is in as more costly resistance strategies are more likely to be selected *in vitro*

than *in vivo* (van Houte et al., 2016). This encompasses not only genetic variation within species

but also differences between species that may differ in infection sites, immune responses, virulence

and sensitivity to phage (Abedon, 2019; Doffkay et al., 2015; van Houte et al., 2016). Further work

using isolates from case studies and clinical trials will give an insight into the predictability of

bacteria and phage evolution from laboratory to *in vivo* studies.

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          <u>Supplementary material</u>

     Parallel resistance evolution to phages during clinical phage therapy and in vitro

Supplementary figures

Data cleaning of the growth rate assays. As shown in Figure S1, we observed a sharp increase in

OD$_{600}$ reads within the initial 2.5 hours which was separate from the exponential growth phase. To

prevent the rolling regression taking these initial measurements into account, these values were

removed from analyses.

(a) Replicate 1

(b) Replicate 2

970   (c) Replicate 3
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Figure S1. Growth curves of bacterial clones as measured by change in log-transformed ocular density

(OD_{600}) over time (hours). Measures from the first 2.5 hours were excluded (blue line) as a rise in OD was

observed independent of bacterial logistic growth which may skew the rolling regression process. Growth

curves were independently replicated a total of 3 times (a - c) owing to cellular clumping which may have

skewed OD measures. ("anc" = ancestral clones; "blank" = uninoculated wells; "control" = control clones;

mix add" = clones from repeated phage treatment; "mix" = clones from treatment where phage were added

once; "resist" = phage resistant clones; "susc" = susceptible clones; "sing_res" = in vitro clones resistant to

one phage; "TP1" = *in vivo* clones from day 2; "TP2" = *in vivo* clones from day 4).

Figure S2. Associations between genetic similarity and phenotypic traits for *in vivo* clones. Individual

points represent individual clones isolated at day 2 (n = 8) and day 4 (n = 2) of phage treatment. Genes that

diverged in frequency *in vitro* from the ancestor are indicated in grey while *in vivo* clones are overlayed

according to their individual genetic make-up. Across quantitative (a – c) and qualitative (d) phenotypic

traits, no clustering in genotypes and clones of particular phenotypes is observed.

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989   **Supplementary Tables**
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991   Table S1. Susceptible colonies picked from *in vitro* phage treatments for phenotypic assays and

992   sequencing.

Phage resistance	Phage added	Treatment replicate	Clone number
Susceptible	Once	6	1
Susceptible	Once	6	2
Susceptible	Once	6	3
Susceptible	Once	3	4
Susceptible	Once	3	5
Susceptible	Once	3	6
Susceptible	Once	5	7
Susceptible	Once	5	8
Susceptible	Repeatedly	6	9
Susceptible	Repeatedly	6	10
Susceptible	Repeatedly	6	11
Susceptible	Repeatedly	1	12

1008   Table S2. Pairwise comparisons of resistance of *in vivo* clones from day 2 against phage from

1009   contemporary (day 2) and two future time-points (days 4 and 7). P-values adjusted using the Tukey

1010   method of comparing a family of three estimates. "SE" = standard error.

Contrast	Estimate	SE	z-ratio	p-value
Day 2 - Day 4	- 16.5	10	-1.65	0.2262
Day 2 - Day 7	- 16.5	10	-1.65	0.2264
Day 4 - Day 7	0.000001	5.32	0	1

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    Supplementary methods

    Data cleaning of biofilm dataset. In the *in vitro* dataset, one replicate in the control and one in

    the susceptible treatments were found to be outliers that significantly impacted data analysis. In

    the susceptible treatment, the outlier (711) had a value 2.9x greater than the group mean (246.9)

    and this was 2.7x greater than the next nearest value (266.1). Similarly, in the control group, the

    outlier (708.6) had a value 2.26x greater than the group mean (313.5) and this was 2.5x greater

    than the next nearest value (278). We subsequently determined the impact of these values

    individually and when together on model output by sequentially removing these from the dataset

    (Table S3, S4). Resistance was significant regardless of outlier removal (Table S3) as were

    comparisons between resistant and susceptible groups (Table S4). Although treatment became

    significant when the susceptible outlier was removed, all Tukey-HSD comparisons were non-

    significant. The only change detected in Tukey HSD comparisons was that between the single

    resistant and susceptible isolates which was found when the susceptible and both outliers were

    removed (Table S4).

1054   Table S3. The effect of outlier removal on the significance of model terms (treatment and phage

1055   resistance). Outlier removed states in which treatment each outlier was identified and subsequently

1056   removed. Significant p-values in bold.

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Model term	Outlier removed	Chi-sq value	P-value
Treatment * Resistance	None	1.32	0.5185
Treatment * Resistance	Susceptible	0.609	0.7824
Treatment * Resistance	Control	1.56	0.459
Treatment * Resistance	Both	0.785	0.675
Treatment	None	2.44	0.295
Treatment	Susceptible	6.207	**0.045**
Treatment	Control	1.501	0.472
Treatment	Both	4.53	0.104
Resistance	None	8.28	**0.016**
Resistance	Susceptible	15.44	**0.00044**
Resistance	Control	12.48	**0.0059**
Resistance	Both	19.13	**7.023e-05**

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1060   Table S4. The effect of outlier removal on Tukey HSD contrasts comparing biofilm productivity

1061   between phage resistance groups. Significant p-values highlighted in bold.

Outlier removed	Contrast	Estimate	SE	t-ratio	p-value
None	Resistant – Single resistant	0.0916	0.210	0.436	0.901
None	Resistant – Susceptible	0.439	0.16	2.698	**0.0289**
None	Single resistant - Susceptible	0.348	0.2	1.74	0.207
Susceptible	Resistant – Single resistant	0.069	0.184	0.375	0.926
Susceptible	Resistant – Susceptible	0.597	0.165	3.618	**0.0029**
Susceptible	Single resistant - Susceptible	0.528	0.194	2.714	**0.0282**
Control	Resistant – Single resistant	0.087	0.193	0.449	0.8953
Control	Resistant – Susceptible	0.508	0.151	3.36	**0.0057**
Control	Single resistant - Susceptible	0.421	0.185	2.28	0.073
Both	Resistant – Single resistant	0.085	0.168	0.506	0.869
Both	Resistant – Susceptible	0.581	0.133	4.36	**0.0004**
Both	Single Resistant - Susceptible	0.496	0.162	3.067	**0.0119**

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