Antibiotic Resistance Increases Evolvability and Maximizes Opportunities Across Fitness

Landscapes

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<u>Abstract</u>

Antibiotic resistance continues to grow as a public health problem. One of the reasons for this

continued growth is that resistance to antibiotics is strongly selected for in the presence of

antibiotics and weakly selected against after their removal. This is frequently thought to be due

to the effects of compensatory mutations. However, compensatory mutations are often not

found in clinically relevant strains of antibiotic resistant pathogens. Here, we conduct

experiments *in vitro* that highlight the role that fine scale differences in environment play in the

maintenance of populations after selection for resistance. We show that differences in the

mode of growth, dictated by environmental factors, are capable of reliably changing the force

and direction of selection. Our results show that antibiotic resistance can increase evolvability

in environments if conditions for selection exist, selecting differentially for newly arising variation

and moving populations to previously unavailable adaptive peaks.

Significance

Antibiotic resistant bacteria are a large and growing problem for public health. A major question

has been why antibiotic resistant strains do not disappear when they must compete with higher

fitness drug sensitive strains. Here we show that selection for antibiotic resistant strains is

particularly sensitive to differences in environmental conditions and that these differences help

to define the fitness landscapes upon which these populations adapt. The result is an increase

in evolvability, with many adaptive peaks that drug resistant populations can explore through

natural selection, making predictions of evolution difficult and selection against resistant strains

improbable.

Introduction

One of the most difficult aspects of the antibiotic resistance problem is the contrast

between the speed and strength of selection for antibiotic resistant strains and the slower

selection against antibiotic strains of bacteria. The evolution for antibiotic resistance occurs

either due to de novo mutations or due to horizontal gene transfer events. The concentration of

antibiotics in the environment can either be low enough to allow the sensitive strains to survive

but high enough to select for resistant strains or the concentration of antibiotics in the

environment can be high enough that the drug susceptible phenotypes die off, leaving only

resistant phenotypes to survive, grow, and reproduce. Different mutations are selected in these

two cases (Wistrand-Yuen et al. 2018). In this paper we are studying the latter case, where only

the resistant strains survive, because of high levels of antibiotics. Essentially, this is the

strongest way that selection can act: either you are susceptible and quickly die, or you are

resistant and survive. In laboratory experiments, the evolution of a population from susceptible

to resistant has been shown to occur on the scale of hours to days (Spagnolo *et al.*, 2016;

Toprak *et al.*, 2011; Miller *et al.*, 2013).

52   As frequencies of antibiotic resistance rose worldwide, coordinated attempts were

undertaken to reverse resistance by removing the environmental conditions that promoted it in

the first place (Ridley et al. 1970). In essence, the strategy was to select against resistance as

quickly and consistently as possible. This was done primarily by lowering rates of antibiotic

usage and relying on the phenomenon known as the "cost of resistance" (Levin et al. 1997;

Andersson and Levin 1999; Austin et al. 1999), whereby many resistant strains have lower

fitness that the sensitive strains in the absence of antibiotics. Typically, this occurs because a

vital phenotype, such as protein or membrane synthesis, is the target of the antibiotic, and

therefore of the resistance mechanism as well. As such, the phenotype of resistant mutants is

not like wild type, making them less fit in direct competition.

62   Results using this ecological approach have been mixed (Andersson and Hughes 2010),

63   with reversion away from resistant phenotypes and back to antibiotic sensitivity shown to occur

64   slowly (Seppala et al. 1997), quickly (van den Bogaard et al. 2000; Agersø and Aarestrup 2013),

65   or not at all (Smith 1975; Mölstad et al. 2008).

66   Resistant mutations are reliably selected for when antibiotics are used for medical or

agricultural purposes (Bush et al. 2011; The Review on Antimicrobial Resistance 2016), but do

not always decline in frequency when the antibiotics are removed for the environment. It has

been proposed that this is due to compensatory mutations (Maisnier-Patin and Andersson

2004). Compensatory mutations relieve the cost of resistance through partial restoration of the

impacted phenotype (Reynolds 2000; Szamecz et al. 2014; Hughes and Andersson 2017) and

have been consistently found in laboratory experiments (Schrag et al. 1997; Reynolds 2000;

Maisnier-Patin and Andersson 2004). But, compensatory mutations are often absent from

clinical strains of antibiotic resistant bacteria (Andersson and Levin 1999; MacLean and Vogwill

2015), suggesting that something more is at play.

In order better understand post-resistance adaptation in bacterial populations, we

conducted several investigations using streptomycin resistant strains of *Escherichia coli* with a

K42N mutation in ribosomal protein S12 under *in vitro* conditions with controlled differences

between them. The results indicate that the adaptive landscape available to a population is of

vital importance to how the population evolves, even when environmental differences seem to

be of extremely fine scale, suggesting that how populations adapt in antibiotic resistance is even

more intricate than previously feared.

Methods

85   **Strains**

All bacterial strains used are *E. coli*, K-12 MG1655 strains. The ancestral strain for all

subsequent experiments is denoted as DD1953 and is known to be free of any plasmids, to be

antibiotic sensitive and to be *rpoS*- due to a premature stop codon in the coding sequence of the

rpoS gene. The DD1953 strain served as the streptomycin sensitive ancestral strain for all

experiments. In order to assure genetic homogeneity at the start, DD1953 was grown from a

single *E. coli* colony from an agar plate, grown in liquid media and then frozen for long term use.

We used a standard mutant screen to generate streptomycin resistant mutants of

DD1953. The single nucleotide polymorphism (SNP) giving streptomycin resistance was

identified and confirmed by MiSeq as well as Sanger DNA sequencing to be a nonsynonymous

change in the *rpsL* gene at codon 42, changing the wild type leucine to asparagine (K42N). A

pure freezer stock was prepared and stored at -80°C. This K42N *rpsL* mutant is the direct

ancestor of all subsequent experimentally evolved populations. This resistant ancestor is

labeled as FS1, or in the case where the mutant is also resistant to T5 bacteriophage, as FS5.

Resistance to the bacteriophage T5 was used as a molecular marker in situations where

one was desirable, such as in direct competition experiments. Resistance to T5 is due to a SNP

in the *fhu* gene and has been shown to be selectively neutral in chemostat experiments (Moser

1958). The minimum inhibitory concentrations (MIC) for all strains were found experimentally

using methods identical to those described earlier (Spagnolo et al. 2016). We found that

DD1953 had a MIC for streptomycin of 2-4 µg/mL, with all resistant strains having MICs over

1000 µg/mL.

Chemostats and Serial Transfer Flasks

To inoculate the chemostats and flasks, we grew strains FS1 and FS5 overnight in flasks

inoculated from frozen cultures containing Davis salts and 0.1% glucose at 37°C shaken at 200

rpm. These cultures were diluted the next morning into fresh media and the growth followed by

measuring optical density. When the optical density reached 1.7, we added 1ml of each

inoculating strain to chemostats and flasks containing 30ml of experimental media, which was

Davis salts supplemented with 0.01% glucose (w/v) as the only available carbon source.

Streptomycin concentrations for all experimental evolution vessels were as indicated.

All continuous culture experiments were conducted in Kubitschek chemostats following

procedures previously described (Dykhuizen 1993). Here, the generation time was set to 3.69

hours per generation (6.5 gen/day, D = 0.1875), matching the mean number of generations per

day in the 1:100 dilution of flask cultures. Additional chemostat related methods are in the

Supplemental Information (**SI**). A total of 4 chemostats were run simultaneously. Chemostats 1

& 2 utilized the Davis minimal media with streptomycin at a concentration of 100 µg/mL (high

streptomycin condition). Chemostat 3 had the Davis minimal media and 16 µg/mL of

streptomycin (low streptomycin condition) as per Miller (Miller 1992). Chemostat 4 had Davis

minimal media but contained no streptomycin (no streptomycin condition). All streptomycin

came from a single prepared stock (10 mg/mL).

Two serial transfer flask experiments were run concurrently with the chemostat

experiments. Whereas the chemostats maintained the experimental populations in a constant

state of sub-maximal exponential growth, the serial transfer flask populations underwent full

cycles of lag phase, maximal exponential growth, and stationary phase each day. Two sterile

250 mL flasks were prepared with 30 mLs of fresh Davis minimal media, 0.01% glucose, and

100 µg/mL of streptomycin (high streptomycin condition). The flasks were placed in an

incubator, maintained at 37°C and constantly shaken at 200 rpm for 24 hours. After a 24 hour

period, 300 µL of the experimental sample was transferred into a fresh 250 mL flask with 30 mL

of fresh media. The 100 fold dilution regime and 24 hour time frame allows for the calculated

number of generations to be 6.5 generations, as per Lenski (Lenski et al. 1991). This

experimental design allows the *average* generation time of the *E. coli* in the flasks to match the

maintained generation time in the chemostats.

MiSeq

137   Miseq NG sequencing was performed on population samples. Libraries were prepared

138   using the Illumina TruSeq genomic DNA-nano and sequenced using Illumina MiSeq 2x300

139   protocols at the University of Wisconsin Biotechnology Center (Madison, WI). Overall mean

140   coverage of the genomic sequencing was 162X (SD: 28).

Bioinformatics

All bioinformatics was performed in the Galaxy environment. The raw sequence reads

were vetted for quality using FastQC and aligned to the ancestral genome (DD1953) using the

reference K-12 MG1655 genome as a scaffold (Genbank accession NC_000913.3). SNP

calling and indel identification were conducted via the VarCap workflow (Zojer et al. 2017),

which also provides SNP frequency data.

Protein Synthesis Assay

Streptomycin acts by binding near the A-site of the ribosome, interfering with the function

of protein S12, the product of the *rpsL* gene. The mechanism of action for streptomycin is

interfering with protein synthesis efficiency, both in rate of protein synthesis and fidelity (Kurland

1992). Structural changes to the S12 protein, particularly at amino acid position 42, are known

to result in high level resistance to streptomycin, but result in reduced protein synthesis efficacy.

In order to compare protein synthesis rates for streptomycin sensitive, resistant, and

evolved strains, we performed a protein synthesis assay. S30 cell lysate was extracted from an

exponentially growing population of *E. coli* that was to be assayed. The S30 lysate extraction

protocol was based upon that of Shrestha (Shrestha et al. 2012) with some modification (See

SI). Following S30 lysate extraction, a firefly luciferase assay was performed as per the *E. coli*

S30 Extract System for Circular DNA by Promega guidelines for control experiments using the

Promega pBEST*luc* plasmid for *E. coli* (Promega, Madison, WI). Following 60 minutes of

incubation at 37°C, relative light units (RLU) were measured in triplicate using a Turner

Biosystems (Sunnyvale, CA) Model 20/20n luminometer (See SI).

Fitness Assay

In order to compare relative fitness of resistant and evolved strains, direct competitions

between strains were conducted in chemostats in the manner of Dykhuizen (Dykhuizen and

Hartl 1980; Dykhuizen 1993). Using the T5 bacteriophage molecular marker, two strains were

inoculated in a chemostat in equal numbers. Changes in the frequencies of these strains were

tracked over 72-96 hours via plating as described.

By plotting the natural log of the ratio of mean CFUs per unit time, the coefficient of

selection was obtained from the slope of the plotted trendline. From these measures, the

relative fitness measure, w_R, was calculated by use of the equation:

$$w_R = e^s$$

171   All chemostat parameters were as described but the concentration of streptomycin was varied

172   to coincide with the context of the competition. We note that competition was under chemostat

173   conditions even for strains evolved in flasks.

Growth Curve Assay

We obtained growth curve data for strains by monitoring optical density (λ = 600 nm) in

96-well plates in a SpectraMax Model 384 Plus microplate reader using SoftMaxPro software

(Molecular Devices, Sunnyvale, CA) over a ten hour period. Briefly, we grew strains overnight

in 3 mL of LB (rich media). The next day, 300 μL of the overnight growth was added to 9 mL of

Davis minimal media (1:30 dilution). The media-strain mixture was vortexed to insure even

distribution. The inoculated media was then put into replicate wells on the 96 well plate, with

180 μLs in each well and at least 1 negative control well per set of technical replicates. A

minimum of 16 replicates were run for each sample.

Following inoculation, the 96 well plate was placed into the plate reader and the

populations were allowed to grow. The OD was measured every 60 seconds over a period of

10 hours, with mixing of the plate between reads. Typically, the populations reached stationary

phase well before the 10 hour mark. Data was analyzed using the grofit package (Kahm et al.

2010) and R statistical software (R Development Core Team 2012). The grofit package

determines the time point for transition into exponential growth as well as into stationary phase,

thereby allowing for analysis of length of time spent in lag and exponential growth phases, as

well as the maximum growth rate (μ_{max}), and the density at the transition into stationary phase.

The amount of time spent in exponential growth was calculated by finding the difference

between the time point where a well population entered stationary phase and the time point

where exponential phase was entered.

<u>Results</u>

Based upon the results of our fitness assays as well as the substantial body of previous

work regarding mechanisms of fitness increases, particularly with regard to streptomycin

resistance, we expected our experimental results would confirm previous work. This

presumption was only partially correct.

Loss of Fitness in Streptomycin Resistant Mutants

Amino acid changes at position 42 of the ribosomal protein S12 confer resistance to high

doses of streptomycin. This mutation also confers substantial fitness costs in *Escherichia coli*

(Kurland 1992), *Mycobacterium tuberculosis* (Sander et al. 2002) and *Salmonella enterica*

(Bjorkman et al. 1998). When two *E. coli* strains with spontaneous streptomycin resistant

mutations (both mutations were in codon 42 in the *rpsL* gene) were individually competed

against an isogenic susceptible strain in flask cultures of a modified Davis salts medium with

limited glucose (Schrag and Perrot 1996; Schrag et al. 1997), the selection found was 11 to

19% against the resistant strain.

In addition, Paulander *et al.*, using a streptomycin resistant mutation in *Salmonella*

typhimurium LT2 [K42N in the S12], measured maximal growth rate in pure culture and then

compared growth rates of the strep-resistant strain to the sensitive strain to determine relative

fitness (Paulander et al. 2009). The sensitive strain grew faster on a rich media (LB) and on M9

minimal media supplemented with glucose or with glycerol, but grew more slowly on poor

media—M9 MM supplemented with pyruvate or with succinate. They showed this was due to

the differential induction of the stress-inducible sigma factor, RpoS. On the poor media the

strep-sensitive cells induce *rpoS*, which retards growth. When both cells lack *rpoS*, the effect

disappears. This example shows how the effects of the environment and other physiological

systems in the cell can reverse expectations.

We competed streptomycin resistant mutants, FS1 and FS5, (Table 1) in glucose limited

chemostats. These mutations were in the codon 42 of the S12 ribosomal protein like the above

example. Strains FS1 and FS5 had lower relative fitness when compared to the streptomycin

sensitive ancestor, DD1953. The fitness of FS1 (T5 bacteriophage susceptible) and FS5 (T5

bacteriophage resistant) was measured to be 0.91 ±SD 0.02 relative to DD1953 (mean of three

samples) in our fitness assay (Table 3). This level of fitness loss is in line with that found in

previous experiments (Schrag and Perrot 1996; Björkman et al. 2000).

Fitness Increases via Compensatory Mutations in Batch Culture

As in previous experiments with streptomycin resistant bacteria in batch culture, the

streptomycin resistant mutants rapidly selected for compensatory mutations in the *rpsD* and

rpsE genes (S4 and S5 small ribosomal proteins, respectively) (Table 2) and increased in

fitness relative to the ancestor (2-4%). The compensatory mutations did not affect MICs of our

strep-resistant strains.

Ribosomal ambiguity mutations (ram) in *rpsD* and *rpsE* have been shown to be

compensatory (Maisnier-Patin et al. 2002) for hyperaccurate *rpsL* mutants, such as ours. This

is believed to be because of ram mutants' pleiotropic effect of increasing polypeptide synthesis

rates, thereby counterbalancing the increased synthesis time caused by slowed proofreading in

hyperaccurate mutants of *rpsL* (Bohman et al. 1984) (See Poehlsgaard and Douthwaite 2005;

Holberger and Hayes 2009 for reviews). These ram mutations do not increase polypeptide

synthesis when the genetic background includes a wild type *rpsL* gene, as measured *in vitro*

using radioactive isotope tagged amino acids (Andersson et al. 1982).

There can be two approaches to testing the overall fitness effects of these compensatory

mutations. The first was taken by Schrag and Perrot where they used P1 to transduce both the

streptomycin sensitive and the streptomycin resistant alleles into the evolved cultures carrying

the streptomycin resistant mutation (Schrag and Perrot 1996). Their results were quite variable

but in all cases the strain carrying the resistant allele had an advantage over the strain carrying

the sensitive allele in flask cultures. They found compensatory mutations in *rpsD* and *rpsE*

which were selected in conjunction with a change at position 42 in the S12 ribosomal protein,

but were detrimental when paired with the streptomycin sensitive allele. These compensatory

mutations led to a condition where reversion to sensitivity would be selected against; genetic

background matters in determining the fitness of the strain carrying a mutation giving resistence

to streptomycin.

We took the second approach. The evolved strains carrying the mutation giving

streptomycin resistance were competed against unevolved strains carrying the streptomycin

resistance allele. The evolved strains had all been evolved either in flask culture or in

chemostats (Table 1). Overall, evolved strains increased fitness relative to their streptomycin

resistant ancestors 2-4% above where they started in the experiments (Table 3).

No Previously Identified Compensatory Mutations in Chemostats

In our chemostat experiments, the growth medium was identical to that of the flask

experiments, although the growth conditions clearly were not. We also varied the concentration

of streptomycin in different chemostats to test for effect. Fitnesses also increased in these

evolved strains, but the genetic mechanism was different. In chemostat cultures, we found no

compensatory mutations in *rpsL, rpsE,* or *rpsD* (Table 2), all of which have previously been

associated with changes in *rpsL*. By performing whole-genome sequencing of evolved clones,

we identified changes in the glucose uptake systems of the chemostat evolved strains,

particularly in *galS*, which have previously been found in chemostat evolved strains of strep-

sensitive *E. coli* (Notley-McRobb and Ferenci 1999). Thus, the changes were not compensatory.

Unexpected Phenotypic Changes

We performed growth curve assays in order to better understand the nature of the

observed fitness increases. The expectation was that we would observe increases in maximum

growth rates in the flask-evolved strains as a result of the compensatory mutations increasing

protein synthesis efficiency. However, even prior to the experimental evolution experiments,

maximum growth rates did not differ between strep-sensitive DD1953 and the resistant clones

FS1 and FS5 in minimal glucose media. In the evolved strains, we were surprised to find no

discernible pattern of increase in maximal growth rates, regardless of whether strains evolved in

flasks or chemostats (Fig 1B). We note that the low maximum growth rate of strain 4C2-12R,

which could be because of some unknown mutational or regulatory change, seems to be

reflected in longer time in exponential growth. This could be a unique phenomenon for which

we have no explanation. Additionally, there was no observable pattern to changes in time spent

in lag phase or the population density at entry to stationary phase for evolved strains when

compared to the strep resistant ancestors (Fig 1A&C).

We did, however, find strong support for directional selection on the amount of time

spent in exponential growth for all evolved strains. This increase in the time spent in

exponential growth is consistent across all strains regardless of *in vitro* method, the

concentration of streptomycin present, or the type or number of mutations found in the evolved

populations. Our results indicate that regardless of experimental condition, selection

consistently acts upon the time in exponential growth. FS1 and FS5, the unevolved strains

containing new mutations giving rise to streptomycin resistance, have a normal growth rate but

a shortened time in exponential growth. In the evolved strains derived from these two strains the

time in exponential growth is like the original strain, DD1953, as if selection acts upon this

phenotype.

In order to better understand these results, a functional protein synthesis assay was

conducted. This assay also yielded unexpected results. Rather than indicating an increase in

the speed and/or accuracy in producing functional proteins (particularly after compensatory

mutations), the rate of synthesis of functional proteins for all evolved strains was *lower* than the

streptomycin resistant ancestors in the presence of streptomycin (Fig 3). This indicates that the

rate of protein synthesis, long thought to be the phenotype most effected by streptomycin

resistance, may not tell the entire story of streptomycin-resistance adaptation in bacteria. As a

control, we also performed the protein synthesis assay under strep free conditions. Under these

conditions, DD1953, the strep-sensitive ancestor, produced functional protein at a rate 2,372X

higher than when in the presence of streptomycin (Fig S1). For evolved strains, most showed

little or no significant difference to results with streptomycin present. Two of three flask evolved

strains had increases in functional protein synthesis when streptomycin was absent, but this is

an environmental condition in which they did not evolve and the rate of protein synthesis was

still approximately two orders of magnitude lower than in wild type. Additional data for the

protein synthesis assays is in the Supplemental Information.

Discussion

Based upon previously published work, many of our results were precisely as expected

for an experimental evolution study investigating adaptation post-resistance. We observed the

cost of resistance typical of streptomycin resistance in *E. coli* with subsequent increases in

relative fitness. In flasks, these fitness increases came via compensatory evolution at known

loci. In chemostat evolved strains, fitness increased through changes in glucose uptake

systems, which has been previously observed for *E. coli* evolved under glucose limited

conditions, although not in relation to streptomycin adaptation. However, when we sought to

investigate the mechanisms underlying these fitness increases, we found novel results

suggesting that the ways in which antibiotic resistant strains adapt post-resistance is more

complex than what is expected. We attempt to dissect our results and provide some potential

context here.

General changes in fitness were anticipated: fitness relative to the antibiotic resistant

ancestor (FS1, FS5) increased across all experimental populations. We found a strong signal

for directional selection for increased time spent in exponential growth in all evolved populations

even though the mechanism of this selection differs depending upon the environmental

conditions experienced during the evolution experiment. This phenotype is one we do not

believe has been highlighted before, but may play a large role in how bacterial populations

grow, particularly if they are also antibiotic resistant. The environmental differences between

conditions may seem minor, given that they were identical except for the *in vitro* method used

and the streptomycin concentration (for chemostats), but even at this extremely fine scale,

environment made all the difference.

In chemostats, the genotypic changes centered on glucose uptake involving the *galS*

system. Similar changes have been observed previously in *E.coli* growing under glucose limited

conditions in chemostats (Notley-McRobb and Ferenci 1999), however, in that case the strain of

E. coli used, BW2952 (Genbank Accession CP001396), was not resistant to streptomycin. This

tells us that such glucose uptake changes mean that this evolutionary path is accessible via

selection without regard to antibiotic resistance phenotypes and was a result of the glucose

limited environment.

In flasks, however, the mechanism was completely different. Genetic changes in serial

transfer populations centered on additional changes to the strep-resistant ribosome. Similar

compensatory mutations have been previously reported for streptomycin resistant strains of

both *E. coli* (Schrag and Perrot 1996; Schrag et al. 1997) and *Salmonella Typhimurium*

(Björkman et al. 1999, 2000). In the previous cases, these compensatory mutations have been

shown to increase rates of polypeptide chain synthesis in the ribosome (Andersson et al. 1982),

which is the phenotype most notably affected by the molecular mechanism of streptomycin

resistance in those ribosomes. The speed of protein synthesis is integral to bacterial growth,

and therefore fitness, in flasks and bacteria are expected to select for increases in maximum

growth rate (μ_{max}) under these conditions (Lenski 2017). Compensatory mutations in strep-

resistant strains is thought to aid in increasing μ_{max}.

However, when we directly measure μ_{max} in our flask evolved strains, the data do not

support this hypothesis. FS1 and FS5 do not have lower μ_{max} values than the streptomycin

sensitive ancestor, DD1953, even though they have a mutation in the *rpsL* gene known to

impact protein synthesis. Additionally, flask evolved strains do not show increases in μ_{max}

values, even though these populations have previously identified compensatory mutations

(Table 2). In fact, there is no pattern of increase in μ_{max} for any of our experimentally evolved

strains (Figure 1B). The lack of change in μ_{max} values is also supported by our functional

protein synthesis assay, which unexpectedly found *lower* rates of production of functional

protein in evolved strains than in the streptomycin resistant ancestor for all strains evolved in the

presence of streptomycin (Figures 2, S1). In fact, the only evolved strain that did not have a

significantly lower rate of protein synthesis was 4C4-10S, which was evolved without

streptomycin *in a chemostat*.

Across all of our evolved strains, the only pattern of selection was for increases in the

amount of time spent in exponential growth (Figure 1D). This signal was robust to all

experimental conditions and concentrations of streptomycin. At the endpoints of the

experiments, the time spent in exponential growth for evolved strains was similar to that for

DD1953, suggesting that this phenotype has a large impact on relative fitness, a conclusion

supported by our fitness assays.

In order to understand our results in context, we believe it is necessary to consider both

the evolvability (Wagner and Altenberg 1996) as well as the fitness landscapes (Payne and

Wagner 2019) of our strains in the environments tested. In chemostats, the genetic changes

(*galS*) were predictable based upon the experimental conditions, even if the phenotypic

changes (time in exponential growth) were not. Genotypic changes did not depend on the

antibiotic resistance state of the genetic background and reflected adaptation to the limiting

nutrient in a chemostat (Warsi and Dykhuizen 2017). In flasks, the selection was for genotypic

changes that were not accessible via natural selection prior to streptomycin resistance. These

mutations were, however, open to selection *after* the switch to a streptomycin laden

environment, suggesting that the antibiotic resistance mutation not only allowed for the strain to

survive in the new environment, but it also increased the evolvability of the strain in this new

environment by opening up adaptive paths that did not exist beforehand. In doing so, post-

resistance adaptation potential (i.e., the number of possible adaptive paths accessible by

selection) increased. This makes the success of any anti-selection regime, such as those often

relied upon in order to control the spread of resistance, highly improbable.

Considered in a different way, we understand that the addition of antibiotic changes the

environment substantially, which also changes the adaptive landscape from the perspective of

the bacteria. Strains that were high fitness in the strep-free environment cannot survive after

the addition of streptomycin. This may be most evident in a genotype to fitness map, where we

understand that the antibiotic sensitive ancestor, DD1953, was high fitness in the absence of

streptomycin. However, the DD1953 strain has an effective fitness of 0 once streptomycin is

added to the environment. In strep-laden environments, FS1 and FS5, which are lower fitness

in the antibiotic free environment, assume higher fitness. The environmental change further

reveals hidden genetic variation (Szamecz et al. 2014), offering FS1 and FS5 access to higher

fitness optima via selection. Our data show that these fitness peaks are related to the time

spent in exponential growth phenotype. Importantly, our results indicate that *how* strains access

these peaks differs based upon their specific *in vitro* environments, with chemostats having an

adaptive path similar to that available to strep-sensitive populations and batch cultures having

an adaptive path contingent upon the strep-resistance mechanism, which was not available to

strep-sensitive strains. In this way, phenotype and environment conspired to increase

evolvability in our streptomycin resistant strains.

The increase in the number of possible adaptive paths for antibiotic resistant strains may

provide a hypothesis to explain why selection against antibiotic resistance has proven so

difficult. Populations at global optima, such as antibiotic sensitive strains, are likely to have

inherent phenotypic compromises. For a large population with sizeable genetic diversity,

environmentally mediated release from a global optimum can be expected to allow for

directional selection for any variety of previously inaccessible phenotypes. Such a shift and

change in accessibility reflects an increase in variability (Payne and Wagner 2019). Antibiotic

resistance represents a specialized case of such an environmentally mediated release whereby

the effect upon selection is particularly strong, with high fitness genotypes suddenly becoming

lethal.

If evolvability does increase post-resistance, antibiotic rich environments may act as a

mechanism that releases populations from the generalized fitness peaks upon which they had

been stranded and may further provide additional paths with which they could then explore a

newly enriched fitness landscape. By changing allelic frequencies while also potentially

simplifying rough adaptive landscapes, the environmental shift increases variability and,

therefore, evolvability. The increased availability of high fitness peaks, particularly in new

environments, illustrates what we have come to think of as a rabbit hole effect, whereby

previously unavailable genotype/fitness combinations become available through a sudden

change in the environment. Much like Alice after tumbling to the bottom of the rabbit hole, there

are new worlds suddenly available for exploration. In this study, the higher fitness flask evolved

strains reached their fitness peaks by increasing the time spent in exponential growth relative to

their direct strep-resistant ancestors through additional changes to a mutated ribosome. These

changes were not genetically possible through selection without the sequence of resistance

mutation, streptomycin positive environment, and then adaptation in batch culture. When the

identical sequence is repeated in continuous culture, such genotypic changes are not observed,

suggesting such a path is less likely, if not altogether impossible.

In 1975, H.W. Smith warned against the idea of combating established antibiotic

resistance by expecting resistant strains to select for susceptibility simply by removing the

antibiotic from the environment (Smith 1975). This ecological fallacy still remains and we have

learned by some tough lessons that reversing resistance will not be quite as simple as hoped

(Andersson and Hughes 2010). Recently, however, new approaches have begun to show

promise in reversing resistance, such as in some phage treatments for drug resistant bacterial

populations (Chan et al. 2016). This study suggests that undoing what has been done may be

even more complicated yet: ephemeral switches in environments may increase the adaptive

potential of resistant populations and allow for increases in fitness that cannot be predicted *a*

priori.

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References

Agersø, Y., and F. M. Aarestrup. 2013. Voluntary ban on cephalosporin use in Danish pig

production has effectively reduced extended-spectrum cephalosporinase-producing Escherichia

coli in slaughter pigs. The Journal of Antimicrobial Chemotherapy 68:569–572.

Andersson, D. I., K. Bohman, L. A. Isaksson, and C. G. Kurland. 1982. Translation rates and

misreading characteristics of rpsD mutants in Escherichia coli. Molecular & General Genetics

187:467–472.

Andersson, D. I., and D. Hughes. 2010. Antibiotic resistance and its cost: is it possible to

reverse resistance? Nature reviews. Microbiology 8:260–71.

Andersson, D. I., and B. R. Levin. 1999. The biological cost of antibiotic resistance. Current

opinion in microbiology 2:489–93.

Austin, D. J., K. G. Kristinsson, and R. M. Anderson. 1999. The relationship between the volume

of antimicrobial consumption in human communities and the frequency of resistace.

Proceedings of the National Academy of Sciences of the United States of America 96:1152–

1156.

Bjorkman, J., D. Hughes, and D. I. Andersson. 1998. Virulence of antibiotic-resistant Salmonella

typhimurium. Proceedings of the National Academy of Sciences 95:3949–3953.

Björkman, J., I. Nagaev, O. G. Berg, D. Hughes, and D. I. Andersson. 2000. Effects of

Environment on Compensatory Mutations to Ameliorate Costs of Antibiotic Resistance. Science

287:1479–1482.

Björkman, J., P. Samuelsson, D. I. Andersson, and D. Hughes. 1999. Novel ribosomal

mutations affecting translational accuracy, antibiotic resistance and virulence of Salmonella

typhimurium. Molecular microbiology 31:53–8.

Bohman, K. T., T. Ruusala, P. C. Jelenc, and C. G. Kurland. 1984. Kinetic Impairment of

Restrictive Streptomycin-Resistant Ribosomes. Molecular & General Genetics 198:90–99.

Bush, K., P. Courvalin, G. Dantas, J. Davies, B. Eisenstein, P. Huovinen, G. A. Jacoby, et al.

2011. Tackling antibiotic resistance. Nature reviews. Microbiology 9:894–896.

Chan, B. K., M. Sistrom, J. E. Wertz, K. E. Kortright, D. Narayan, and P. E. Turner. 2016. Phage

selection restores antibiotic sensitivity in MDR Pseudomonas aeruginosa. Scientific Reports

6:1–9.

Dykhuizen, D. E. 1993. Chemostats Used for Studying Natural Selection and Adaptive

Evolution. Methods in Enzymology 224:613–631.

Dykhuizen, D., and D. L. Hartl. 1980. Selective neutrality of 6PGD allozymes in E. coli and the

effects of genetic background. Genetics 96:801.

Holberger, L. E., and C. S. Hayes. 2009. Ribosomal protein S12 and aminoglycoside antibiotics

modulate A-site mRNA cleavage and transfer-messenger RNA activity in Escherichia coli. The

Journal of biological chemistry 284:32188–200.

Hughes, D., and D. I. Andersson. 2017. Evolutionary Trajectories to Antibiotic Resistance.

Annual Review of Microbiology 71:579–596.

Kahm, M., G. Hasenbrink, H. Lichtenberg-Frate, J. Ludwig, and M. Kschischo. 2010. grofit□:

Fitting Biological Growth Curves with R. Journal of Statistical Software 33.

Kurland, C. G. 1992. Translational accuracy and the fitness of bacteria. Annual review of

genetics 26:29–50.

Lenski, R. E. 2017. Experimental evolution and the dynamics of adaptation and genome

evolution in microbial populations. The ISME Journal 11:2181–2194.

Lenski, R. E., M. R. Rose, S. C. Simpson, and S. C. Tadler. 1991. Long-term experimental

evolution in Escherichia coli. I. Adaptation and divergence during 2,000 generations. American

Naturalist 1315–1341.

Levin, B. R., M. Lipsitch, V. Perrot, S. Schrag, R. Antia, L. Simonsen, N. M. Walker, et al. 1997.

The population genetics of antibiotic resistance. Clinical Infectious Diseases 24:S9–S16.

MacLean, R. C., and T. Vogwill. 2015. Limits to compensatory adaptation and the persistence of

antibiotic resistance in pathogenic bacteria. Evolution, Medicine, and Public Health 2015:4–12.

Maisnier-Patin, S., and D. I. Andersson. 2004. Adaptation to the deleterious effects of

antimicrobial drug resistance mutations by compensatory evolution. Research in Microbiology

155:360–369.

Maisnier-Patin, S., O. G. Berg, L. Liljas, and D. I. Andersson. 2002. Compensatory adaptation to

the deleterious effect of antibiotic resistance in Salmonella typhimurium. Molecular microbiology

46:355–66.

Miller, C., J. Kong, T. T. Tran, C. A. Arias, G. Saxer, and Y. Shamoo. 2013. Adaptation of

Enterococcus faecalis to daptomycin reveals an ordered progression to resistance.

Antimicrobial agents and chemotherapy 57:5373–83.

Miller, J. H. 1992. A Short Course in Bacterial Genetics: A Laboratory Manual and Handbook for

Escherichia coli and Related Bacteria (1st ed.). Cold Spring Harbor Laboratory Press,

Plainview, NY.

Mölstad, S., M. Erntell, H. Hanberger, E. Melander, C. Norman, G. Skoog, C. S. Lundborg, et al.

2008. Sustained reduction of antibiotic use and low bacterial resistance: 10-year follow-up of the

Swedish Strama programme. Lancet Infectious Diseases 8:125–132.

Moser, H. 1958. The Dynamics of Bacterial Populations Maintained in the Chemostat. Carnegie

Institute publ. no. 614. Carnegie Institution for Science, Washington DC.

Notley-McRobb, L., and T. Ferenci. 1999. Adaptive mgl-regulatory mutations and genetic

diversity evolving in glucose-limited Escherichia coli populations. Environmental Microbiology

1:33–43.

Paulander, W., S. Maisnier-Patin, and D. I. Andersson. 2009. The fitness cost of streptomycin

resistance depends on rpsL mutation, carbon source and RpoS (sigmaS). Genetics 183:539–

46, 1SI-2SI.

Payne, J. L., and A. Wagner. 2019. The causes of evolvability and their evolution. Nature

Reviews Genetics 20:24–38.

Poehlsgaard, J., and S. Douthwaite. 2005. The bacterial ribosome as a target for antibiotics.

Nature reviews. Microbiology 3:870–81.

R Development Core Team. 2012. R: A Language and Environment for Statistical Computing. R

Foundation for Statistical Computing Vienna Austria. Foundation for Statistical Computing,

Vienna, Austria.

Reynolds, M. G. 2000. Compensatory evolution in rifampin-resistant Escherichia coli. Genetics

156:1471–1481.

Ridley, M., R. Lynn, D. Barrie, and K. C. Stead. 1970. Antibiotic-Resistant Staphylococcus

aureus and Hospital Antibiotic Policies. The Lancet 295:230–233.

Sander, P., B. Springer, T. Prammananan, A. Sturmfels, M. Kappler, M. Pletschette, and E. C.

Böttger. 2002. Fitness cost of chromosomal drug resistance-conferring mutations. Antimicrobial

Agents and Chemotherapy 46:1204–1211.

Schrag, S. J., and V. Perrot. 1996. Reducing antibiotic resistance. Nature 381:120–121.

Schrag, S. J., V. Perrot, and B. R. Levin. 1997. Adaptation to the fitness costs of antibiotic

resistance in Escherichia coli. Proceedings. Biological sciences / The Royal Society 264:1287–

1291.

Seppala, H., T. Klaukka, J. Vuopio-Varkila, A. Muotiala, H. Helenius, K. Lager, P. Huovinen, et

al. 1997. The Effect of Changes in the Consumption of Macrolide Antibiotics on Erythomycin

Resistance in Group A Streptococci in Finland. The New England Journal of Medicine 337:441–

446.

Shrestha, P., T. M. Holland, and B. C. Bundy. 2012. Streamlined extract preparation for

Escherichia coli-based cell-free protein synthesis by sonication or bead vortex mixing.

BioTechniques 53:163–174.

Smith, H. W. 1975. Persistence of Tetracycline Resistance in Pig E. coli. Nature 258:628–630.

Spagnolo, F., C. Rinaldi, D. R. Sajorda, and D. E. Dykhuizen. 2016. The Evolution of

Resistance to Continuously Increasing Streptomycin Concentrations in Populations of E. coli.

Antimicrobial Agents and Chemotherapy 60:1336–1342.

Szamecz, B., G. Boross, D. Kalapis, K. Kovács, G. Fekete, Z. Farkas, V. Lázár, et al. 2014. The

Genomic Landscape of Compensatory Evolution. PLoS biology 12:e1001935.

The Review on Antimicrobial Resistance. 2016. *Tackling Drug-Resistant Infections Globally:*

Final Report and Recommendations. London, United Kingdom.

Toprak, E., A. Veres, J.-B. Michel, R. Chait, D. L. Hartl, and R. Kishony. 2011. Evolutionary

paths to antibiotic resistance under dynamically sustained drug selection. Nature Genetics

44:101–105.

van den Bogaard, A. E., N. Bruinsma, and E. E. Stobberingh. 2000. The effect of banning

avoparcin on VRE carriage in The Netherlands. Journal of Antimicrobial Chemotherapy 46:146–

148.

Wagner, G. P., and L. Altenberg. 1996. Complex adaptations and the evolution of evolvability.

Evolution 50:967–976.

Warsi, O. M., and D. E. Dykhuizen. 2017. Evolutionary implications of Liebig's law of the

minimum: Selection under low concentrations of two nonsubstitutable nutrients. Ecology and

Evolution 7:5296–5309.

Wistrand-Yuen, E., M. Knopp, K. Hjort, S. Koskiniemi, O. G. Berg, and D. I. Andersson. 2018.

Evolution of high-level resistance during low-level antibiotic exposure. Nature Communications

9:1–12.

Zojer, M., L. N. Schuster, F. Schulz, A. Pfundner, M. Horn, and T. Rattei. 2017. Variant profiling

of evolving prokaryotic populations. PeerJ 5:e2997.

	Strain Name	Description	Growth Condition	Streptomycin Concentration
Antibiotic Sensitive Ancestor	DD1953	wild type ancestor: $2 \geq MIC \leq 4$		
Antibiotic Resistant Ancestors	FS1	resistant ancestor strain (*rpsL*: K42N)		
	FS5	resistant ancestor strain (*rpsL*: K42N) T5 bacteriophage resistant		
Evolved Antibiotic Resistant Strains	4C2-12R	Experimentally evolved	Chemostat	High
	4C3-10S	Experimentally evolved	Chemostat	Low
	4C4-10S	Experimentally evolved	Chemostat	No Streptomycin
	4S1-44R	Experimentally evolved	Serial Transfer	High
	4S1-44S	Experimentally evolved	Serial Transfer	High
	4S2-44S	Experimentally evolved	Serial Transfer	High

Table 1: Strain Names and Experimental Conditions

All strains were derived from a single antibiotic sensitive ancestor, DD1953. Two streptomycin
resistant mutants of DD1953 were generated, FS1 and FS5, differing only in resistance to T5
bacteriophage, which was used as a neutral molecular marker. Evolved strains experienced
different *in vitro* environments (4C designates chemostat evolved and 4S means evolved in
serial batch culture). Chemostat evolved strains also had a range of set streptomycin
concentrations for their time under experimental conditions (last column).

Evolved Strain	Gene and Change	Frequency	Description
4C2-12R	*galS* Deletion	0.14	1,022 bp deletion
	galS Deletion	0.26	1 bp deletion
	malT	0.41	Nonsynonymous change
	malK	0.43	Nonsynonymous change, ATP-binding subunit
4C3-10S	*galS*	1	Nonsynonymous change
4C4-10S	*galS*	1	Synonymous change
4S1-44R	*rpsD*	0.27	Structural constituent of ribosome; nonsynonymous change, L198P. Previously unreported compensatory mutation.
	rpsD	0.2	Structural constituent of ribosome; nonsynonymous change, I199S. Previously unreported compensatory mutation.
	rpsE	0.5	Structural constituent of ribosome; nonsynonymous change, G108V. Known compensatory mutation site.
	acs	0.6	Acetyl CoA synthase, acetate-scavenging enzyme, improves carbon starvation survival; promoter region
4S1-44S	*rpsE*	1	Structural constituent of ribosome; nonsynonymous change, G108A. Known compensatory mutation site.
	yjcO intergenic region	0.27	Sel1 family TPR-like repeat protein; function unknown
4S2-44S	*gstB* promoter region	1	glutathione-dependent bromoacetate dehalogenase
	rpsE	1	Structural constituent of ribosome; nonsynonymous change, G86C. Known compensatory mutation.
	yjcO intergenic region	0.26	Sel1 family TPR-like repeat protein; function unknown
	yjeJ proximal promoter region	1	Uncharacterized protein

578   Table 2: Genetic Changes in Evolved Strains Relative to Antibiotic Resistant Ancestor

579   Genetic changes identified through MiSeq relative to the relevant antibiotic resistant ancestor
580   (FS1 or FS5). SNP frequency, as quantified through the VarCap bioinformatics pipeline, as well
581   as function data (as per the Ecogene database, http://ecogene.org) are provided for each
582   mutated gene.

Pre-Experiment		Post-Experiment	
Strain	**Fitness**	**Strain**	**Fitness**
DD1953	1	FS1	1
FS1	0.919	FS5	0.998
FS5	0.917	4C2-12R	1.04
		4C3-10S	1.03
		4C4-10S	1.03
		4S1-44S	1.02
		4S1-44R	1.02
		4S2-44S	ND

Table 3: Strain Relative Fitnesses before and After Evolution Experiment

All relative fitnesses were quantified from direct competition in chemostats. Fitness was
measured as the slope of the best fit line for a time series of ratios of T5 sensitive to T5 resistant
colonies (in manner of Dykhuizen and Hartl, 1980). A. Fitnesses of streptomycin resistant
ancestors (FS1, FS5) relative to their streptomycin sensitive ancestor, DD1953. B. Relative
fitnesses of all evolved strains relative to their pre-experiment ancestor, as obtained by direct
competition. ND indicates not determined.

Figure 1: Boxplots of Growth Curve Assay for Ancestor and Evolved Strains

All strains were tracked in a 96-well plate assay for growth related phenotypes. A. Time Spent
in Lag Phase: No observable pattern for changes in time spent in lag phase was found for
strains. B. Maximum Growth Rate: Counter to expectation, maximal growth rate did not decline
for streptomycin resistant strains FS1 or FS5 relative to DD1953 and then did not increase after
evolution under serial passage conditions. In fact, μ_{max} declined in flask evolved strains relative
to the antibiotic resistant ancestors. C. Population Density at Entry to Stationary Phase: The
final density of experimental populations did not show a pattern after experimental evolution. D.
Time Spent in Exponential Phase: The time spent in exponential growth significantly declined
for streptomycin resistant mutants FS1 and FS5 relative to the DD1953 Ancestor. Following *in*
vitro evolution, all evolved strains showed significant increases in the time spent in exponential
growth, with values approaching those of DD1953. This phenotype has not been previously
described relating to antibiotic resistance.

Figure 3: Mean Functional Protein Synthesis with Streptomycin Present

A functional protein synthesis assay was developed that quantified the amount of functional
firefly luciferase protein produced in one hour. Quantities taken using Relative Light Units in a
luminometer. Contrary to expectation, evolved strains had lower rates of function protein
production than streptomycin resistant ancestors, FS1 or FS5. The evolved strain with the
highest mean functional protein production (and largest confidence interval) is 4C4-10S, which
evolved *without* streptomycin present in the chemostat. Error bars indicate 95% CI. See
Supplemental Information for assay conditions and protocol.