METABOLIC DISRUPTION IMPAIRS RIBOSOMAL PROTEIN LEVELS, RESULTING IN ENHANCED AMINOGLYCOSIDE TOLERANCE

Rauf Shiraliyev and Mehmet A. Orman

William A. Brookshire Department of Chemical and Biomolecular Engineering, University of Houston, Houston, TX, 77204
Corresponding author: Mehmet Orman, morman@central.uh.edu

ABSTRACT

Aminoglycosides, a class of antibiotics, have been in use for decades, displaying broad-spectrum activity against Gram-negative and Gram-positive bacteria. They target ribosomes and disrupt protein synthesis. Although their use declined due to newer antibiotics with lower toxicity, increasing drug resistance has renewed interest in aminoglycosides. Herein, we have demonstrated that energy metabolism plays a crucial role in aminoglycoside tolerance, as knockout strains with deleted genes associated with the tricarboxylic acid cycle (TCA) and the electron transport chain (ETC) exhibited increased tolerance to aminoglycosides in the mid-exponential growth phase of *Escherichia coli* cells. Our initial hypothesis posited that genetic perturbations would lead to a reduction in the proton motive force, subsequently affecting the uptake of aminoglycosides. This hypothesis is based on the prevailing notion that aminoglycoside uptake is dependent on the distinctive and energy-driven electrochemical potential across the cytoplasmic membrane. However, our results did not support this hypothesis. Despite genetic perturbations in mutant strains, we found no consistent metabolic changes, ATP levels, cytoplasmic pH variations, or membrane potential differences compared to wild-type strains. Additionally, intracellular concentrations of fluorophore-labeled gentamicin remained similar across all strains. To uncover the mechanism responsible for the observed tolerance in mutant strains, we employed untargeted mass spectrometry to quantify the proteins within these mutants and subsequently compared them to their wild-type counterparts. Our comprehensive analysis, which encompassed protein-protein association networks and functional enrichment, unveiled a noteworthy upregulation of proteins linked to the TCA cycle in the mutant strains, suggesting that these strains compensate for the perturbation in their energy metabolism by increasing TCA cycle activity to maintain their membrane potential and ATP levels. Furthermore, our pathway enrichment analysis shed light on local network clusters displaying downregulation across all mutant strains, which were associated

32 with both large and small ribosomal binding proteins, ribosome biogenesis, translation factor

33 activity, and the biosynthesis of ribonucleoside monophosphates. These findings offer a plausible

34 explanation for the observed tolerance of aminoglycosides in the mutant strains. Altogether, this

35 research has the potential to uncover mechanisms behind aminoglycoside tolerance, paving the

36 way for novel strategies to combat such cells.

38 INTRODUCTION

Global public health is currently confronted with a critical challenge in the form of antimicrobial

resistance, a threat that significantly jeopardizes human well-being (Binder et al., 1999).

Addressing this escalating issue necessitates a comprehensive understanding of the diverse

mechanisms employed by bacterial cells to survive antibiotic treatments. These mechanisms can

be broadly classified into two groups: reversible and irreversible. Reversible mechanisms, often

termed "tolerance" mechanisms, are not genetically inherited and encompass alterations in

bacterial growth and behavior. Within this category, antibiotic-tolerant cells, including persisters,

viable but nonculturable cells, and stationary-phase cells, exhibit the ability to transiently survive

high antibiotic concentrations. Their formation is linked to stochastic and/or deterministic

processes triggered by various stress factors such as the SOS response, stringent response, reactive

oxygen species, nutrient depletion, and overpopulation (Amato et al., 2013; Germain et al., 2015;

Molina-Quiroz et al., 2018; Theodore et al., 2013). The reversible nature of these cells allows them

to oscillate between antibiotic-tolerant and sensitive states. Conversely, irreversible mechanisms,

referred to as "resistance" mechanisms, are heritable and connected to mutagenic processes

(Balaban et al., 2019). Antibiotic-resistant mutants, falling into this category, can emerge due to

mutations in antibiotic target proteins or proteins that contribute to enhanced repair mechanisms,

cell dormancy, drug efflux systems, and alternative mechanisms that circumvent the antibiotic's

target (Lambert, 2005; Okusu et al., 1996).

Aminoglycosides, having received approval for human use, were among the earliest antibiotics to

be used in clinical practice (Becker and Cooper, 2013; Krause et al., 2016). Aminoglycosides are

a class of naturally occurring or semisynthetic amino-modified sugars known for their broad-

spectrum activity against both Gram-negative and Gram-positive bacteria (Aggen et al., 2010;

Endimiani et al., 2009; Landman et al., 2010; Ristuccia and Cunha, 1985). They have been a crucial

part of antibacterial therapy since the discovery and introduction of streptomycin for the treatment

of tuberculosis, which was isolated from the soil bacterium *Streptomyces griseus* (Kresge et al.,

2004; Schatz et al., 2005; Woodruff, 2014). This was followed by the emergence of other members

of the class, including gentamicin, kanamycin, tobramycin, amikacin, and neomycin (Becker and

Cooper, 2013; Krause et al., 2016). The use of aminoglycosides declined with the advent of newer

antibiotics, such as fluoroquinolones, which were thought to have lower toxicity (Becker and

Cooper, 2013; Krause et al., 2016). However, the rise of resistance to these drug classes has

sparked a revival of interest in the older aminoglycosides and the development of new ones with

improved dosing schemes (Becker and Cooper, 2013; Krause et al., 2016).

Studies on the mechanism of action of aminoglycosides against bacteria have shown that they

disrupt protein synthesis by targeting the ribosome. The primary action of aminoglycosides is to

bind to the 16S ribosomal RNA (rRNA) of bacteria, a component of the 30S ribosomal subunit

(Moazed and Noller, 1987; Recht et al., 1999; Woodcock et al., 1991). The initial entry of

antibiotics into cells causes misreading as the antibiotics interact with the ribosomes involved in

chain elongation. This is the first step in bactericidal action. Following this, the misread protein is

incorporated into the membrane, causing membrane damage (Davis et al., 1986). This increases

the amount of antibiotics entering the cells, leading to more misreading and the formation of

channels (Davis et al., 1986). Finally, the blockage of the initiating ribosomes occurs, causing the

inhibition of protein synthesis (Davis et al., 1986). Aminoglycosides can also bind to and stabilize

RNA helix 69 (H69) of 50S ribosomal subunit, potentially inhibiting ribosome recycling facilitated

by ribosome recycling factor (Borovinskaya et al., 2007). Moreover, the binding of

aminoglycosides to the 50S ribosomal subunit can lead to the inhibition of mRNA and tRNA

translocation (Borovinskaya et al., 2007).

It has been demonstrated that antibiotic-tolerant cells can become susceptible to aminoglycosides

by metabolizing certain carbon sources (Allison et al., 2011). This susceptibility arises from an

enhanced drug uptake as a result of an increase in the electron transport chain activity and

membrane potential, facilitated by the breakdown of these specific carbon sources (Allison et al.,

2011). The process of aminoglycoside uptake is considered to be a unique, energy-requiring

mechanism where the electrochemical potential across the cytoplasmic membrane and electron

flow through membrane-bound respiratory chains are believed to be significant factors (Taber

Harry W et al., 1987). However, the bactericidal effect of aminoglycosides may not result from

the downstream impacts of voltage-dependent drug uptake, but rather from an irregular membrane

potential (Bruni and Kralj, 2020). Bruni and Kralj suggested that hyperpolarization, stemming

from changes in ATP flux due to the reversal of F1Fo-ATPase activity, could potentially intensify

aminoglycoside-mediated cell death (Bruni and Kralj, 2020).

While a renewed interest in aminoglycoside antibiotics has unveiled interesting connections

between bacterial energy metabolism and aminoglycoside tolerance, the precise impact of energy

metabolism disruption on energy-dependent aminoglycoside uptake remains unclear. In our study,

we found that knockout strains with genes related to the tricarboxylic acid cycle (TCA) and the

electron transport chain (ETC) deleted displayed increased tolerance to aminoglycosides.

Intriguingly, this increased tolerance was not attributed to reduced proton motive force, which

affects drug uptake, as evidenced by insignificant alterations in ATP levels or membrane potential

compared to the wild-type strain. We employed untargeted mass spectrometry to quantify proteins

in the mutant and wild-type strains, revealing a notable upregulation of proteins associated with

the TCA cycle in the mutants. This suggests that these strains compensate for the disruption in

their energy metabolism by enhancing TCA cycle activity to maintain their membrane potential

and ATP levels. Moreover, our pathway enrichment analysis emphasized local network clusters

that were consistently downregulated across all mutant strains. These clusters were related to both

large and small ribosomal binding proteins, ribosome biogenesis, translation factor activity, and

the biosynthesis of ribonucleoside monophosphates. These findings provide a credible rationale

for the observed tolerance to aminoglycosides in the mutant strains.

RESULTS

Deletions of the TCA cycle and ETC genes increased tolerance to aminoglycosides.

Given the crucial role of energy metabolism in aminoglycoside tolerance, our initial objective was

to assess various knockout strains that involved the deletion of genes associated with the TCA

cycle, such as *sucA, gltA, mdh, sdhC, icd, acnB,* and *fumA,* as well as the NADH dehydrogenase

enzyme of ETC, including *nuoM* and *nuoI* from *E. coli* MG1655. To conduct the experiments,

both wild-type and mutant strains were cultured overnight and then diluted 100-fold in fresh Luria-

Bertani (LB) medium. The cultures were grown until the mid-exponential phase (t=3.5 hours) in a

shaker at 37°C and 250 rpm. Subsequently, the cultures were exposed to various aminoglycosides

(50 μg/ml streptomycin, 50 μg/ml gentamicin, and 50 μg/ml amikacin) for a duration of 5 hours.

Samples were collected before and after the treatments, and these were plated on LB agar to

quantify the surviving cell fractions. Analysis of the results indicates that most of the knockout

strains (Δ*sucA*, Δ*gltA*, Δ*mdh*, Δ*sdhC*, Δ*nuoM*, and Δ*nuoI*) exhibited increased tolerance to

streptomycin, gentamicin, and amikacin treatments when compared to the wild-type strain (**Figure**

1A-C). However, it is worth noting that mutant strains such as Δ*icd*, Δ*acnB*, and Δ*fumA* did not

always exhibit increased tolerance (**Figure 1A-C**), underscoring the complex interplay between

energy metabolism and specific antibiotic tolerance. Moreover, the surviving cells measured in

these assays may not necessarily represent antibiotic-resistant cells, as the antibiotic tolerance

assays were conducted at concentrations well above the Minimum Inhibitory Concentration (MIC)

levels. The MIC levels of the strains for the tested antibiotics displayed no drastic alterations

compared to those of the wild type, despite some minor variations among them (**Supplementary**

Table S1).

Figure 1. Deletions in the TCA cycle and ETC genes enhance tolerance to aminoglycosides. In the
mid-exponential phase (t=3.5 h), *E. coli* MG1655 wild-type and knockout strains underwent (A)
streptomycin, (B) gentamicin, and (C) amikacin treatments at a concentration of 50 μg/ml for a duration of
5 hours. Following the treatments, cells were washed to eliminate the antibiotics and then plated on LB
agar plates to quantify the surviving cell fractions. CFU: Colony forming units. WT: Wild type. For pairwise
comparisons, one-way ANOVA with Dunnett's post hoc test was used where *, $P < 0.05$, **, $P < 0.01$, ***,
$P < 0.001$, and ****, $P < 0.0001$. N=3. Data points represent mean and standard error.

To gain a thorough understanding of the observed tolerance, we chose four mutant strains that

exhibit either high tolerance (*sucA*, *gltA*, and *nuoI*) or low tolerance (*icd*) to gentamicin for the

subsequent assays. Our initial objective was to delineate the time-dependent profiles of antibiotic-

tolerant cells within cultures. To achieve this, cells from overnight cultures were diluted in fresh

media and cultured for 6 hours. Hourly samples were collected for cell quantification and antibiotic
tolerance assays. We opted for flow cytometry for precise cell quantification, a more reliable
method than optical density measurements (Mohiuddin et al., 2020) (**Figure 2A**). In the antibiotic
tolerance assays, samples were treated with gentamicin (50 µg/ml) for 5 hours and plated before
and after the treatments (**Figure 2B**). Notably, the observed tolerance in these mutant strains is not
linked to cell growth. For instance, the mutant strains associated with *sucA, gltA,* and *nuoI* genes
exhibited higher tolerance to aminoglycosides although they did not show significantly altered cell
growth compared to the wild type (**Figure 2A, B**). Notably, these mutant strains displayed a
substantial increase in the number of tolerant cells (more than 10^4-fold) between time points t=3
and t=4 hours during the mid-exponential growth phase when compared to the wild-type strain
(**Figure 2B**). Conversely, while the *icd* mutant exhibited a significant reduction in cell growth, it
was observed to be as sensitive to aminoglycosides as the wild type during the exponential growth
phase and the formation of tolerant cells in this mutant strain was only evident after the time point
t=5 hours (**Figure 2A, B**). We also emphasize that the observed tolerance in the mutant strains is
transient or reversible, as they all exhibit high sensitivity to aminoglycosides, similar to the wild
type, during the lag phase of growth (**Figure 2B**).

Figure 2. The observed tolerance in the mutant strains is not linked to cell growth. (A) Growth of *E.*
coli MG1655 wild type, Δ*sucA*, Δ*gltA*, Δ*nuoI*, and Δ*icd* strains was assessed by measuring the number of
cells per ml with flow cytometry. (B) The cells of both the *E. coli* MG1655 WT and mutant strains were
collected from the cultures at indicated time intervals and then subjected to gentamicin treatment. The figure
shows the colony-forming unit (CFU) levels of both the treated and untreated cultures, indicating the
surviving and total cell population, respectively. N=3. Data points represent mean and standard error.

Energy-dependent aminoglycoside uptake is not a contributing factor.

Aminoglycoside uptake is a unique and energy-requiring mechanism that depends on the

electrochemical potential across the cytoplasmic membrane (Taber Harry W et al., 1987). Initially,

we hypothesized that the genetic perturbations in these strains may have decreased PMF and

aminoglycoside uptake. The mutant strains exhibiting increased aminoglycoside tolerance

demonstrated no consistent pattern in metabolic activities during the mid-exponential growth

phase, which was assessed using redox sensor green (RSG) dye for the indicated time points (t=3,

4, 5 and 6 hours) of the cell growth (**Figure 3A**). RSG dye can be reduced by bacterial reductases

of PMF, allowing it to enter the bacteria and produce a green fluorescence signal. This fluorescence

signal should be suppressed by the presence of a metabolic inhibitor, such as carbonyl cyanide 3-

chlorophenylhydrazone (CCCP) (**Supplementary Figure S1**). Given that ATP is a pivotal product

of PMF, we also quantified intracellular ATP levels in both wildtype and mutant strains, employing

the BacTiter-Glo™ Microbial Cell Viability Assay (Catalog# G8230, Promega Corporation,

Madison WI). This assay utilizes a single reagent to lyse cells and produce luminescence through

the luciferase reaction, with the luminescent signal being directly proportional to the ATP content

(**Supplementary Figure S2**). We found no consistent pattern in ATP levels between the antibiotic-

sensitive and tolerant strains, particularly during the mid-exponential growth phase (at t=3 and 4

hours) (**Figure 3B**). While the $\Delta sucA$ mutant generally showed reduced ATP levels, this was not

the case for the other strains, which showed increased ATP levels around t=6 hours (**Figure 3B**).

Furthermore, since the proton gradient is a vital element of PMF and a reduced H^+ ion gradient

across the cell membrane is linked to reduced membrane potential, we conducted measurements

of intracellular pH using the ratiometric pHluorin, known as a pH-sensitive derivative of green

fluorescent protein (GFP) (Miesenböck et al., 1998). This GFP variant exhibits a bimodal

excitation spectrum characterized by peaks at 410 nm and 470 nm, along with an emission

maximum at 530 nm (Martinez II et al., 2012). When subjected to acidification, the excitation at

410 nm diminishes while the excitation at 470 nm concurrently rises, which allows us to construct

standard curves for the measurement of intracellular pH (**Supplementary Figure S3**). For these

assessments, we introduced pGFPR01 plasmids, where pHluorin is expressed under the arabinose-

induced promoter PBAD, into the mutant strains utilized. Our results revealed that there was no

anticipated acidification of the cytoplasm in the TCA and ETC mutants compared to the wild type,

both during the mid-exponential and early stationary phases (**Figure 3C**).

Figure 3. Comparable ATP levels, redox activities, and cytoplasmic pH levels were observed between the wild-type and mutant strains. (A) RSG staining was conducted by suspending wild-type and mutant strains in 0.85% sodium chloride solution during the mid-exponential and early stationary phases, as outlined in the Materials and Methods section. (B) The ATP levels were measured in wild type and mutant cells during the mid-exponential and early stationary growth phases. (C) Comparison of the cytoplasmic pH of the WT and mutant cells during the mid-exponential and stationary phases was performed using using the ratiometric pHluorin. For pairwise comparisons, one-way ANOVA with Dunnett's post hoc test was used where *, $P < 0.05$, **, $P < 0.01$, ***, and $P < 0.001$. $N \geq 3$. Data points represent mean and standard error.

Finally, we utilized fluorophore-labeled aminoglycoside (Gentamicin-Texas Red, or GTTR) to investigate the cellular uptake of the drug. Specifically, cells in the mid-exponential phase of both wild-type and mutant strains (at t=3.5 hours when a significant increase in gentamicin tolerance was observed in the mutant strains) were exposed to GTTR for an hour, followed by the analysis of cells using flow cytometry. A short-term treatment was preferred, in line with a previous study (Bruni and Kralj, 2020) as cells treated with aminoglycosides become permeabilized at later time

points (Bruni and Kralj, 2020), potentially introducing artificial impacts on drug uptake. Notably,

the deletion of genes related to the TCA cycle and the electron transport chain did not induce a

significant alteration in GTTR uptake in cells during the exponential phase when compared to the

wild type (**Figure 4A, B**). Collectively, these findings from multiple approaches strongly suggest

that energy-dependent drug uptake is not the primary determinant of the observed antibiotic

tolerance.

Figure 4. Deletion of genes associated with the TCA cycle and electron transport chain showed no significant alteration in drug uptake. (A) Exemplary quantification of gentamicin-Texas red uptake in cells during the exponential growth phase. (B) Gentamicin-Texas red staining was performed on cells in the mid-exponential phase for both the wild-type and mutant strains, followed by fluorescence measurement using flow cytometry after one hour. For pairwise comparisons, one-way ANOVA with Dunnett's post hoc test was used (no statistical significance was detected). N=3. Data points represent mean and standard error.

Membrane potential dysregulation is not associated with the observed aminoglycoside tolerance.

Fluorescent sensors for voltage and calcium have been utilized to monitor electrophysiology in

bacteria at the single-cell level (Bruni and Kralj, 2020), and the findings revealed that the

dysregulated membrane voltage was not essential for aminoglycoside uptake or inner membrane

pore formation in *E. coli*, but it proved crucial for bactericidal activity (Bruni and Kralj, 2020). To

assess the significance of dysregulated PMF, we employed a well-established assay based on 3,3′-

Dipropylthiadicarbocyanine [DiSC$_3$(5)] (Stokes et al., 2020; Wu et al., 1999), a fluorescent dye

commonly used for monitoring cell membrane potential. During cell hyperpolarization, DiSC$_3$(5)

infiltrates the cell membrane, undergoing self-quenching and resulting in reduced fluorescence intensity. Conversely, during depolarization, the fluorescence intensity of $DiSC_3(5)$ increases as it exits the cell membrane. Polymyxin B served as a control (**Supplementary Figure S4**), as the polycationic peptide ring of polymyxin B binds to a negatively charged site within the lipopolysaccharide layer of the cellular membrane (Domingues et al., 2012), leading to the dissipation of the electric potential gradient which enhances the fluorescence intensity of $DiSC_3(5)$ (**Supplementary Figure S4**). To assess the effect of gentamicin on membrane potential, metabolically active cells from exponential phase cultures at time points t=3, 4, 5, and 6 hours were transferred to a buffer solution containing $1\,\mu M$ $DiSC_3(5)$ (**Figure 5**). Upon reaching equilibrium, the cells were treated with gentamicin. At designated time points, samples were collected, and fluorescence levels were measured using a plate reader (**Figure 5**). Similar to Polymyxin B treatments, gentamicin disrupts the electric potential gradient of PMF, leading to an increase in the fluorescence intensity of $DiSC_3(5)$ (**Figure 5**). Despite significant and rapid alterations in membrane potential observed upon gentamicin treatment, there was no significant difference in membrane potential between aminoglycoside-sensitive and tolerant strains (**Figure 5**). Moreover, if any of the gene deletions had an impact on membrane potential, we would have expected to observe altered fluorescence intensity in the specific knockout strain compared to that of the wild type during the equilibrium stage (before the addition of the drug), as previously reported (Mohiuddin et al., 2022). However, no significant difference in $DiSC_3(5)$ fluorescence intensities was observed among the strains tested during the equilibrium phase (the initial 20 minutes, as illustrated in **Figure 5**).

The membrane dysregulation induced by aminoglycosides was previously attributed to the combined activity of NADH dehydrogenase and a reversed F1Fo-ATPase (Bruni and Kralj, 2020), and the knockout strains targeting the proton-conducting Fo domain (Δ*atpB,* Δ*atpE,* Δ*atpF*), as well as Δ*atpG*, showed increased colony-forming units in response to aminoglycoside treatments (Bruni and Kralj, 2020). However, when we tested ATP synthase knockout strains of *E. coli* BW25113 from the Keio collection (for consistency with the previous study (Bruni and Kralj, 2020)), we did not observe a comparable trend as reported earlier (Bruni and Kralj, 2020). In fact, the strains Δ*atpB,* Δ*atpE,* Δ*atpF,* and Δ*atpG* exhibited similar sensitivity to gentamicin compared to the wild type under the conditions studied here (**Supplementary Figure S5A**). Interestingly, we observed a drastic increase in gentamicin tolerance in the Δ*atpC* mutant strain (**Supplementary**

Figure S5A), which was not reported in the previous study (Bruni and Kralj, 2020). The *atpC* gene
encodes an F1 complex subunit of ATP synthase, promoting motor activity in the direction of ATP
production rather than the reversed direction (Bruni and Kralj, 2020; Guo et al., 2019). We
acknowledge potential variations between *E. coli* BW25113 and *E. coli* MG1655 strains, due to
differences in their genomic DNA. To address this, we deleted two Fo components (Δ*atpA* and
Δ*atpB*) and two F1 components (Δ*atpC* and Δ*atpD*) from *E. coli* MG1655 and assessed their
gentamicin survival profiles. Notably, the drastic increase in the number of tolerant cells reported
for the TCA and ETC mutant strains during the mid-exponential growth phase was not observed
in these ATP synthase mutant strains. However, a moderate increase in tolerance was noted for
the Δ*atpC* and Δ*atpD* mutants (F1 components) during the mid-exponential phase (t=3 and 4
hours) compared to the wild type (**Supplementary Figure S5B**), without a clear trend supporting
a reverse ATPase-mediated killing through the Fo domain. Overall, while we observe gentamicin-
induced dysregulation in membrane potential, it may not be the primary factor contributing to the
observed differences in antibiotic tolerance levels between the strains.

Figure 5. The dysregulation of membrane potential is not correlated with the observed
aminoglycoside tolerance. Deletion of TCA cycle and electron transport chain involved strains showed no
significant change in PMF. Mid-exponential and early stationary phase cells of wild type and mutant strains
were stained with DiSC$_3$(5), and at specified time intervals their fluorescence was measured using a plate
reader. Cells were treated with gentamicin after 20 minutes and fluorescence was measured again. N=4.
Data points represent mean and standard error.

Proteomic analysis reveals molecular responses in mutant strains, unveiling potential mechanisms underlying aminoglycoside tolerance.

In order to gain further insights beyond the conducted experiments and elucidate the mechanism responsible for the observed tolerance in TCA cycle and ETC mutants, we utilized untargeted mass spectrometry to quantify proteins within these mutants and subsequently compared them to their wild-type counterparts. Proteomics data analysis, specifically involving the determination of protein fold-change and calculation of P-value (using F-test and t-test), was carried out through a structured process involving data transformation, normalization, and statistical procedures, as detailed elsewhere (Aguilan et al., 2020). Proteins exhibiting a positive $\log_2$(fold-change) and a P-value less than 0.05 were categorized as significantly upregulated, while those displaying a negative $\log_2$(fold-change) and a P-value less than 0.05 were categorized as significantly downregulated. The STRING database was employed to predict both the physical and functional interactions between proteins by entering a group of upregulated and downregulated proteins (Szklarczyk et al., 2023). Additionally, STRING carried out an automated pathway-enrichment analysis, focusing on the proteins entered and identifying pathways that occurred more frequently than expected. The analysis was based on the statistical background of the entire genome and encompassed Gene Ontology annotations (all three domains), KEGG pathways, Uniprot keywords, and other functional pathway classification frameworks. This also included a hierarchical clustering of the STRING network itself. When discussing our findings, we primarily reference the latter method, as it offers the advantage of broader coverage, including potential novel modules that might not yet be classified as pathways (Szklarczyk et al., 2019).

In the context of the upregulated protein-protein association networks in the Δ*sucA* and Δ*gltA* mutant strains, the STRING analysis unveiled significant functional enrichments (**Supplementary Table S2 and S3**). These included the TCA cycle, carbon and pyruvate metabolism, formate c-acetyltransferase activity, and the fatty acid metabolic pathway (**Figure 6A, B, D, and E**). Specifically, key proteins associated with these enrichments encompassed FumA, FumC, AcnA, AcnB, SdhA, SdhB, Dld, TalA, FbaB, FadA, FadB, and Acs (**Figure 6A, B**). Additionally, the Δ*sucA* and Δ*gltA* mutants exhibited an upregulated cluster of membrane proteins, particularly from the ABC transporter family (MalK, RbsA, RbsB, DppA, OppA, UgpB, LuxS), with some proteins specialized in amino acid transport across the plasma membrane (ArtL, LivJ, GltI, GlnQ, ArgT, HisJ) (**Figure 6A, B**). In the case of the Δ*gltA* mutant, a smaller network of upregulated stress-

induced proteins was also observed, related to osmotic stress, oxidative stress, and starvation

(OsmC, OsmY, YbaY, YgaU, YjbJ, Dps, WrbA) (**Figure 6B**). Regarding the Δ*nuoI* mutant, the

STRING analysis identified an upregulated functional network associated with carbon and

pyruvate metabolism, featuring proteins SucD, FumA, SdhA, Crp, Eda, GalS, Crr, Mdh, PykF,

Lpd, GapA, Pgk, PoxB, and TalA, as well as upregulated flagellar proteins (FlgF, FlgK, FlgM,

FliD, FliH, FliJ, FliK, FliN, CheW) (**Figure 6C, F, Supplementary Table S4**). Interestingly, our

proteomics data on these three mutant strains, which display higher gentamicin tolerance, reveals

upregulation in proteins associated with energy metabolism (e.g., TCA cycle, pyruvate

metabolism) (**Figure 6**). This suggests that these mutant strains compensate for metabolic

perturbations by enhancing the TCA cycle to preserve their ATP levels, redox activities, and PMF.

Indeed, this proteomics data aligns well with measurements of ATP, RSG, and PMF highlighted

in previous sections (**Figures 3-5**).

In the context of downregulated protein networks, the substantial functional enrichments unveiled

through analysis using the STRING database were found to be intricately associated with both the

large and small ribosomal subunits, denoted by the specific ribosomal proteins Rplx, Rplm, Rplt,

Rply, Rplp, Rpls, Rpse, Rpsg, Rpsh, Rpsm, Rpst, Rpsi, Rpsp, Rpsk, Rpsc, Rpsd, Rpsl, Rpmg,

Rpmf, and Rpmb (**Supplementary Table S5, S6, and S7, Figure 7A-F**). These enrichments

extended to domains encompassing translation factor activities, translation processes, and protein

export mechanisms, featuring genes Rimm, Infa, Infc, Tsf, Yeip, Nusg, Efp, Prfc, Mfd, Glys, Cyss,

Asps, Args, and Tig (**Figure 7A-C**). Additionally, a noteworthy connection was established with

the ribonucleoside monophosphate biosynthetic process, as characterized by genes Cara, Carb,

Pyrd, Pyrf, and Pyrl (**Figure 7A-C**). The consistent downregulation of ribosomal binding proteins,

ribosome biogenesis, translation factor activity, and the biosynthesis of ribonucleoside

monophosphates is observed in the three mutant strains (Δ*sucA*, Δ*gltA*, and Δ*nuoI*), providing an

explanation for the observed aminoglycoside tolerance in these mutants (**Figure 7A-F**). However,

genome-level proteomics trends, specifically the upregulation in energy metabolism and

downregulation in ribosomal proteins, were not observed in the Δ*icd* mutant strain

(**Supplementary Table S8 and S9, Supplementary Figure S6**). This may be expected, as the

Δ*icd* mutant strain did not exhibit a significant difference in gentamicin tolerance compared to the

wild type.

Figure 6. Proteomics data on the mutant strains, exhibiting increased gentamicin tolerance, indicates upregulation in proteins linked to energy metabolism. Cells from both wild-type and mutant strains at the mid-exponential phase (t=3.5 h) were collected after which protein extraction and digestion were carried out for Liquid Chromatography–Mass Spectrometry (LC-MS) analysis. The STRING visual network depicts upregulated protein interactions of Δ*sucA* (A), Δ*gltA* (B) and Δ*nuoI* (C) mutants. The thickness of edges in the network reflects the strength of data support, derived from curated databases, experimental data, gene neighborhood, gene fusions, co-occurrence, co-expression, protein homology, and text mining. The protein clusters and their corresponding gene names are visually distinguished through color-coding on the networks. The panels D-F depict functional enrichments in the protein network for Δ*sucA* (D), Δ*gltA* (E), and Δ*nuoI* (F) mutants. This includes information on the number of proteins associated with the enrichment strength and the false discovery rate (FDR) indicating the significance of enrichment through P-value (calculated using the Benjamini-Hochberg method). N= 3.

Figure 7. Proteomics data on the mutant strains, exhibiting increased gentamicin tolerance, reveals downregulation in proteins associated with ribosomes. The STRING visual network displays upregulated protein interactions for ΔsucA (A), ΔgltA (B), and ΔnuoI (C) mutants. Panels D-F provide details on functional enrichments in the protein network for ΔsucA (D), ΔgltA (E), and ΔnuoI (F) mutants. N= 3.

DISCUSSION

The study aimed to explore the impact of gene knockouts associated with the TCA cycle and NADH dehydrogenase enzyme on aminoglycoside tolerance in *E. coli* MG1655. Various knockout strains showed increased tolerance to streptomycin, gentamicin, and amikacin compared to the wild-type strain. However, mutants like Δ*icd*, Δ*acnB*, and Δ*fumA* did not consistently exhibit enhanced tolerance, highlighting the complex relationship between energy metabolism and antibiotic tolerance. The investigation further focused on four selected mutant strains (*sucA, gltA, nuoI,* and *icd*), revealing time-dependent profiles of antibiotic-tolerant cells, with a substantial increase observed in the Δ*sucA, ΔgltA, and ΔnuoI* mutants during the mid-exponential growth phase. The observed tolerance was not linked to altered cell growth, and it appeared to be transient or reversible, as all mutants regained sensitivity to aminoglycosides during the lag phase of growth.

We further investigated the factors influencing aminoglycoside tolerance in *E. coli* MG1655 by examining the role of energy-dependent drug uptake, membrane potential, and genetic perturbations. Studies have shown that cells tolerant to antibiotics can transition to susceptibility to aminoglycosides through the metabolism of specific carbon sources (Allison et al., 2011). This shift in susceptibility is attributed to increased drug uptake, facilitated by increased activity in the electron transport chain and membrane potential, both triggered by the breakdown of these particular carbon sources (Allison et al., 2011). Aminoglycoside uptake involves a three-step process: initial ionic binding to cells, followed by two energy-dependent phases (EDP I and EDP II) (Taber Harry W et al., 1987). The first step, concentration-dependent ionic binding, is not affected by inhibitors of energized uptake (Taber Harry W et al., 1987). The EDP I phase is gradual, requiring aminoglycoside concentration and a membrane potential for substantial uptake (Taber Harry W et al., 1987). Elevated aminoglycoside concentrations can bypass EDP I. EDP II involves rapid transport across the cytoplasmic membrane, relying on energy from electron transport and potentially ATP hydrolysis. Once inside the cell, studies suggest bactericidal effects through mistranslation and misfolding of membrane proteins, triggering oxidative stress and hydroxyl radical formation, ultimately leading to cell death (Kohanski et al., 2008, 2007). Despite hypothesizing that reduced PMF and aminoglycoside uptake might be linked to genetic perturbations, the mutant strains with increased tolerance showed no consistent pattern in metabolic activities in our study. Analysis of ATP levels, intracellular pH, and fluorophore-labeled aminoglycoside uptake did not reveal a clear association with antibiotic tolerance. Certainly, our

proteomic analysis revealed significant enhancements in upregulated protein networks linked to the TCA cycle, carbon metabolism, pyruvate metabolism, and fatty acid pathways in the mutant strains exhibiting increased gentamicin tolerance. This sheds light on how these mutants manage and maintain their ATP levels, redox activities, proton motive force (PMF), and drug uptake, overcoming genetic disruptions in their metabolism.

Bruni and Kralj postulated that the bactericidal action triggered by membrane potential might involve mechanisms beyond drug uptake, potentially influenced by the combined effects of pore formation by mistranslated proteins and membrane hyperpolarization (Bruni and Kralj, 2020). Additionally, their study proposed that the sudden shift in energy demand resulting from ribosome dissociation during aminoglycoside treatment could boost cellular ATP flux, leading to hyperpolarization through the concerted activity of NADH dehydrogenase and a reversed F1F0-ATPase (Bruni and Kralj, 2020). Interestingly, while our study documented dysregulation in membrane potential after gentamicin treatment, no significant differences were noted between sensitive and tolerant strains. Despite prior research implicating increased survival of specific ATP synthase knockouts in response to aminoglycoside treatments (Bruni and Kralj, 2020), our study did not report a clear trend regarding the gentamicin tolerance of these knockout strains. We acknowledge that variations in experimental conditions, such as medium composition, bacterial strains, growth phase, treatment time, duration, and antibiotic compositions, may account for the observed differences. Additionally, we recognize that genetic perturbations can have pleiotropic effects, potentially altering cell survival and death mechanisms, which may differ from those observed in the wild-type strain.

We conducted untargeted mass spectrometry to quantify proteins in the TCA-cycle and ETC mutants. Utilizing the STRING database, we predicted functional protein interactions, revealing significant enrichments in downregulated protein networks associated with ribosomal subunits, translation factor activities, protein export mechanisms, and ribonucleoside monophosphate biosynthesis in the mutant strains that displayed higher gentamicin tolerance. The altered levels of ribosomal proteins observed in our mutant strains align well with a prior study that identified respiratory complex I in *E. coli* as a critical mutational target for promoting persister cell formation during the transition to a stationary phase (Van den Bergh et al., 2022). In that study, point mutations in respiratory complex I, responsible for proton translocation, were found to induce antibiotic tolerance (Van den Bergh et al., 2022). While they showed that mutations compromised

proton translocation, key components such as NADH oxidation, electron transfer, and drug uptake remained largely unchanged (Van den Bergh et al., 2022). They also demonstrated that the increased persistence correlated with an acidified cytoplasm during the stationary phase, impacting protein translation and contributing to increased antibiotic tolerance, yet no significant differences in pH and persistence were noted between mutant and wild-type strains during the exponential phase (Van den Bergh et al., 2022). Similarly, our investigation revealed that single gene deletions did not alter cellular PMF, redox activities, membrane potential, and drug uptake. However, the downregulation of ribosomal protein levels reported in our study may not be attributed to a reduction in cytoplasmic pH, as we did not observe alterations in pH levels in the knockout strains compared to the wild type during the mid-exponential phase. Certain genes linked to the deletion strains in our study, such as *icd, sucA, mdh,* and *fumA,* do not encode membrane-bound respiratory proteins. Strain-specific differences can significantly influence the outcomes and responses to various experimental conditions. Each strain of *E. coli* or any other microorganism can harbor unique genetic backgrounds, mutation profiles, and physiological characteristics, leading to distinct survival strategies in response to environmental stresses, such as antibiotic exposure.

In summary, our study revealed that deletions in TCA cycle and ETC genes, including *sucA, gltA, mdh, sdhC, icd, acnB, fumA, nuoM,* and *nuoI*, increased aminoglycoside tolerance. Analyses, including flow cytometry and proteomics, demonstrated enhanced tolerance without significant changes in energy-dependent drug uptake or membrane potential. Our pathway analysis identified a downregulation in large and small ribosomal binding proteins, ribosome biogenesis, translation factor activity, and ribonucleoside monophosphate biosynthesis in all mutant strains, offering a plausible explanation for the observed aminoglycoside tolerance in these mutants. Altogether, our findings underscore the complexity of energy metabolism's role in antibiotic tolerance, providing valuable insights into the molecular mechanisms of aminoglycoside tolerance in specific mutant strains.

MATERIALS AND METHODS

Bacterial strains, chemicals, media, and growth conditions

We employed the in-house *Escherichia coli* K-12 MG1655 strain for our experiments. The gene deletions in *E. coli* K-12 MG1655 were established in our prior studies(Ngo et al., n.d.) utilizing

the Datsenko and Wanner method (Datsenko and Wanner, 2000). *E. coli* K-12 BW25113 and its

corresponding knockout strains were procured from the Keio collection, purchased from Horizon

Discovery (Catalog #OEC4988, Lafayette, CO, USA). The validity of all deletions was confirmed

through the use of check primers (Ngo et al., n.d.). All chemicals utilized in the study were

purchased from Fisher Scientific (Atlanta, GA) or VWR International (Pittsburg, PA) unless stated

otherwise. The Gentamicin-Texas Red Conjugate (Catalog # 24300) was purchased from AAT

Bioquest, Inc (CA, USA). The ATP measurement kit (G8230) was acquired from Promega

Corporation (Madison, WI). Standard Luria-Bertani (LB) broth was prepared by dissolving 5 g

yeast extract, 10 g tryptone, and 10 g sodium chloride in 1 l of deionized (DI) water. LB agar was

made by dissolving 40 g of pre-mixed LB agar powder in 1 l DI water. Sterilization of LB broth

and LB agar was done via autoclaving at 121°C and 103.421 kPa. To determine tolerant cells in

cultures, gentamicin (50 μg/mL), streptomycin (50 μg/mL), and amikacin (50 μg/mL) were used,

and their concentrations were selected to be much higher than the minimum inhibitory

concentrations (MICs). MICs of antibiotics for *E. coli* were determined using MIC Test strips

(Fisher Scientific), as previously described (Andrews, 2001), and the results were presented in

Supplementary Table S1. Ampicillin (100 μg/mL) was utilized to maintain plasmids in the cells.

0.2% L-arabinose at 13.3 mM was added to the media to induce GFP expression. To decrease the

antibiotic concentration below the MIC, a sterile 1X phosphate buffer saline (PBS) solution was

utilized to wash the cells. Antibiotics were dissolved in DI water and sterilized using a 0.2 μm PES

syringe filter. Overnight cultures were prepared in 14 ml round-bottom tubes (Catalog # 14-959-

1B, Fisher Scientific) by incubating cells from a frozen 25% glycerol stock (-80°C) in 2 ml of LB

at 37°C and 250 rpm. After 24 hours, cells from the overnight cultures were diluted 1:100 fold in

2 ml of fresh medium and cultured further to achieve the desired growth phase for the assays.

Cell growth assay

To prepare overnight precultures, cells from frozen stocks were inoculated into 2 mL of LB

medium in 14-mL round-bottom tubes and cultured for 24 hours at 37°C with shaking at 250 rpm.

The overnight precultures were then diluted 100-fold in 14-mL round-bottom tubes containing 2

mL of LB medium and incubated in the shaker at 250 rpm and 37°C. The growth of the cultures

was determined by measuring the number of cells per mL using flow cytometry. To perform this

task, cells were transferred to PBS at specified time points and analyzed with a flow cytometer

(NovoCyte Flow Cytometer, NovoCyte 3000RYB, ACEA Biosciences Inc., San Diego, CA). The

cell populations were delineated on flow diagrams utilizing the forward and side scatter

parameters, with controls including PBS with cells and PBS without cells. The inclusion of a PBS

solution without cells aids in noise determination. The instrument is capable of quantifying both

the number of events and the volume of the solution under analysis. Roughly 30,000-50,000 events

were analyzed for each sample.

Clonogenic survival assay

Overnight cultures were prepared in 14 ml tubes by incubating cells from a frozen 25% glycerol

stock (-80°C) in 2 ml of LB at 37°C and 250 rpm. After 24 hours, cells from the overnight cultures

were diluted 1:100 fold in 2 ml of fresh medium and cultured in the incubator at 37°C with shaking.

At the designated growth phase or time points, cells were exposed to antibiotics at the specified

concentrations (50 μg/ml) for a duration of 5 hours. To determine the number of live cells before

antibiotic exposure, 10 μl of cell cultures were serially diluted in PBS and plated on an LB agar

plate, which was incubated for 16 hours at 37°C. During the antibiotic treatments, 1 ml cultures

were collected after 5 hours and washed twice with PBS through centrifugation at 13,300 rpm

(17,000 g) for 3 minutes to remove antibiotics. After the final centrifugation, 900 μl of supernatant

was removed using a pipette, and the cell pellets were resuspended in the remaining 100 μl of PBS.

Next, 10 μl of the cell suspensions were serially diluted in PBS, and 10 μl of the diluted cell

suspensions were spotted onto LB agar plates. The plates were then incubated at 37°C for at least

16 hours, and the colony-forming units (CFU) were counted to determine the number of live cells

present in the cultures. Survival fractions were calculated by dividing the number of surviving

cells (after treatment) by the initial number of cells (before treatment).

Redox sensor green dye staining

To measure bacterial reductase and ETC (electron transport chain) activities, we used the *Bac*Light

Redox Sensor Green Vitality kit (Catalog#B34954, Thermo Fisher) following the manufacturer's

instructions. Overnight cultures were prepared in 14 ml Falcon tubes by incubating cells from a

frozen 25% glycerol stock (-80°C) in 2 ml of LB at 37°C and 250 rpm. After 24 hours, cells from

the overnight cultures were diluted 1:100 fold in 2 ml of fresh medium and cultured in the incubator

at 37°C with shaking. For analyzing the cell populations during mid-exponential and early

stationary phases (t = 3h, 4h, 5h, 6h), we diluted the cells in 1 ml of 0.85% sodium chloride solution

in flow cytometry tubes (5 ml round bottom falcon tubes) by varying amounts (10-, 20-, 20-, 50-

fold, respectively). After that, Redox Sensor Green (RSG) dye was added to the cells at a

concentration of 1 μM, and the samples were incubated at 37°C for 10 minutes before flow

cytometry analysis. For the negative controls, cell suspensions were treated with 20 μM CCCP 5

minutes prior to RSG staining to disrupt membrane electron transport (**Supplementary Figure**

S2). Positive controls consisted of mid-exponential-phase cells. The cell populations were gated

on flow diagrams using the forward and side scatter parameters of unstained controls. Cells were

excited at 488 nm with a solid-state laser, and green fluorescence was collected with a 530/30

bandpass filter.

ATP measurement

The BacTiter-Glo Microbial Cell Viability assay kit (Catalog# G8230, Promega Corporation) was

used to measure the intracellular ATP levels of both *E. coli* MG1655 WT and mutant strains during

a specified growth phase, following the manufacturer's instructions. To generate a standard curve,

ATP solutions of known concentrations were used (**Supplementary Figure S2**). Background

luminescence was measured using LB broth.

pH measurement

For pH measurements in *E. coli*, a pGFPR01 plasmid was used, in which the GFP derivative

ratiometric pHluorin is expressed from the arabinose-induced promoter P_{BAD}. This plasmid was

kindly provided by Keith A. Martinez II (Department of Biology, Kenyon College, Gambier, Ohio,

USA). A comprehensive pH measurement protocol was obtained from a prior study (Van den

Bergh et al., 2022). Overnight cultures of WT and mutant strains were prepared in 14 ml tubes by

incubating cells carrying the pGFPR01 plasmid from a frozen 25% glycerol stock (-80°C) in 2 ml

of LB, 0.2% L-arabinose and 100 mg/ml ampicillin, at 37°C and 250 rpm. After 24 hours, cells

from the overnight cultures were diluted 1:100 fold in 2 ml of fresh medium, 0.2% L-arabinose

and 100 mg/ml ampicillin and cultured in the incubator at 37°C with shaking. Fluorescence was

measured at 410 nm and 470 nm using a plate reader at different time points, after measuring and

normalizing the optical density of cells in liquid culture. The 410/470 fluorescence ratios were

recorded to determine the cytoplasmic pH using the standard curve (**Supplementary Figure S3**).

To generate standard pH versus fluorescence ratio curves for *E. coli* MG1655 cells, transmembrane

pH was collapsed by adding 40 mM potassium benzoate and 40 mM methylamine hydrochloride

to the cells, equalizing the difference between the external and internal pH. The cultures were then

buffered to different pH levels ranging from 5 to 10 using a 50 mM concentration of 2-(*N*-
morpholino) ethanesulfonic acid (MES) or 3-(*N*-morpholino) propanesulfonic acid (MOPS), and
corresponding fluorescence values at 410 nm and 470 nm were obtained using a plate reader. The
Boltzmann equation was used to establish the standard curve for each bacterial strain based on the
provided data (Martinez II et al., 2012).

DiSC$_3$(5) assay

The fundamental mechanism of this assay and the assessment of PMF components using DiSC$_3$(5)
have been extensively detailed elsewhere (Farha et al., 2013; Panta and Doerrler, 2021; Stokes et
al., 2020). *E. coli* MG1655 wild-type and mutant cells in both exponential and early stationary
phases (at time points t = 3 h, 4h, 5h, 6h) were collected and subjected to two washes with an assay
buffer containing 5 mM HEPES and 20 mM glucose. The cell density was set to OD$_{600}$=0.1, and
the cells were stained with 1 μM DiSC$_3$(5). Fluorescence readings were taken at specified intervals
using a plate reader, with excitation and emission wavelengths set at 620 nm and 670 nm,
respectively. Gentamicin (50 μg/mL) was introduced 20 minutes after cells reached equilibrium.
At this point, the probe was released into the medium, leading to an upsurge in fluorescence. CCCP
(100 μM) and polymyxin B (32 μg/mL) were included as controls and administered 20 minutes
after equilibrium to dissipate the proton and electron gradients of PMF, respectively. The
concentration of polymyxin B at 32 μg/mL was determined based on a prior study (Stokes et al.,
2020).

Gentamicin uptake assay

Overnight cultures were prepared in 14 ml tubes by incubating cells from a frozen 25% glycerol
stock (-80°C) in 2 ml of LB at 37°C and 250 rpm. After 24 hours, cells from the overnight cultures
were diluted 1:100 fold in 2 ml of fresh medium and cultured in the incubator at 37°C with shaking.
At the mid-exponential phase (t=3.5 h), 100 μl of cell cultures for both wild-type and mutants were
exposed to gentamicin-Texas red (GTTR) at a final concentration of 25 μg/ml. Untreated mid-
exponential phase cells were used as negative controls. The samples containing gentamicin-Texas
red were then incubated for 1 hour at 37°C with shaking at 250 rpm. Subsequently, 20 μl from
each sample was washed with 500 μl of PBS (1X). After the final washing step, cell pellets were
resuspended in 500 μl of PBS. All samples were subjected to analysis using a flow cytometer

equipped with lasers emitting light at a wavelength of 561 nm, and the resulting red fluorescence

was detected using a 615/20 nm bandpass filter.

574   Proteomics sample preparation and analysis

Cells from both wild-type and mutant strains at the mid-exponential phase (t=3.5 h) were collected

after which protein extraction and digestion were carried out for Liquid Chromatography–Mass

Spectrometry (LC-MS) analysis by following the Sample Preparation by Easy Extraction and

Digestion (SPEED) procedure. The experiments were conducted at the University of Houston

Mass Spectrometry Laboratory under a service fee. Comprehensive details regarding the protein

isolation and digestion methods can be found elsewhere (Doellinger et al., 2020). Briefly, cell

pellets were added 20 uL TFA and incubated at room temperature for 5 min followed by addition

of 200 ul of 2M trisbase. After adding Tris(2-carboxyethyl) phosphine (10 mM) and 2-

Chloroacetamide (40 mM), the reaction mixture was heated at 95C for 5 min. Digestion was

performed by adding trypsin (1/40, w/w) and incubation at 37 ºC overnight. The digested peptides

were cleaned up using a C18 Ziptip and vacuum dried using a CentriVap (Labconco). Each dried

sample was resuspended in 2% ACN with 0.1% FA for LC-MS analysis. The method involving

LC-MS has been detailed in a separate publication (Qin et al., 2022). Specifically, a NanoElute

LC system connected to a timsTOF Pro (Bruker Daltonics, Germany) through a CaptiveSpray

source was utilized. Samples were loaded onto an in-house packed column (75 μm x 20 cm, 1.9

μm ReproSil-Pur C18 particle from Dr. Maisch GmbH, Germany) with a column temperature of

40 °C. Mobile phases included buffer A (0.1% FA in water) and buffer B (0.1% FA in ACN). The

short gradient was 0-17.8 minutes, from 2% B to 30% B, followed by 18.3 minutes to 95% B, and

20.7 minutes to 95% B. The parallel accumulation-serial fragmentation (PASEF) mode with 4

PASEF scans per cycle was employed. The electrospray voltage was set at 1.4 kV, and the ion

transfer tube temperature was maintained at 180 °C. Full MS scans were conducted across the

mass-to-charge (m/z) range of 150–1700. The target intensity value was 2.0×10^5 with a threshold

of 2500. A fixed cycle time of 0.53 s was established, and a dynamic exclusion duration of 0.4

minutes with a ± 0.015 amu tolerance was applied. Only peaks with a charge state of ≥ 2 were

chosen for fragmentation. The default settings of MSFragger (Yu et al., 2020) were applied to

analyze data obtained from the mentioned instrument. The UniProt-SwissProt *E. coli* K12 database

(Taxon ID 83333, downloaded on 6/19/2023, 4518 entries) served as the reference. Fixed

modification involved cysteine carbamidomethylation, while variable modifications included

methionine oxidation and acetylation. Peptide length was restricted to 7-50, allowing for 2 missed

cleavages. Both precursor and product ion masses were set as monoisotopic. The false discovery

rate (FDR) was controlled at <1% at the peptide spectrum match, peptide, and protein levels.

Proteomics data analysis

The processing of proteomics data and the calculations of fold change were essentially carried out

following the methods described in the paper by Aguilan *et al* (Aguilan et al., 2020). In summary,

we utilized Excel spreadsheets for key stages of data transformation, normalization, fold change,

and P-value calculation. Initially, proteins lacking quantitative values were excluded, and a

logarithm transformation was applied to achieve a normal distribution of data. Normalization,

using both average and slope methods, was then employed to minimize intragroup variation in

technical replicates and log fold change calculations compared to the transformed unnormalized

data. Subsequently, missing values were imputed by replacing them with approximated values

using the Probabilistic Minimum Imputation method. Following imputation, we determined the

relative ratio of each protein in mutant and wild-type strains, along with P-value calculation using

the parametric *t*-test. The selection of the *t*-test type involved an *F*-test to evaluate whether the

replicates for each protein exhibited homoscedastic (equal variances) or heteroscedastic (unequal

variances) characteristics. For the identification of significant networks among input proteins, we

utilized the STRING tool V 12.0. This entailed inputting proteins that were significantly

upregulated and downregulated based on specified thresholds, as detailed elsewhere (Szklarczyk

et al., 2023).

Statistical analysis

624   All assays were conducted using at least three independent biological replicates. The figures

625   display the mean value and standard error for each data point. Statistical analysis was performed

626   using GraphPad Prism software, with one-way analysis of variance (ANOVA) with Dunnett's post

627   hoc test to determine significance. The P value threshold was set at *, $P < 0.05$, **, $P < 0.01$, ***,

628   $P < 0.001$, ****, $P < 0.0001$, and ns nonsignificant.

REFERENCES

Aggen JB, Armstrong ES, Goldblum AA, Dozzo P, Linsell MS, Gliedt MJ, Hildebrandt DJ, Feeney LA, Kubo A, Matias RD, Lopez S, Gomez M, Wlasichuk KB, Diokno R, Miller GH, Moser HE. 2010. Synthesis and spectrum of the neoglycoside ACHN-490. *Antimicrob Agents Chemother* **54**:4636–4642. DOI: https://doi.org/10.1128%2FAAC.00572-10, PMID: 20805391

Aguilan JT, Kulej K, Sidoli S. 2020. Guide for protein fold change and: P-value calculation for non-experts in proteomics. *Mol Omics* **16**:573–582. DOI: https://doi.org/10.1039/D0MO00087F, PMID: 32968743

Allison KR, Brynildsen MP, Collins JJ. 2011. Metabolite-enabled eradication of bacterial persisters by aminoglycosides. *Nature* **473**:216–220. DOI: https://doi.org/10.1038/nature10069, PMID: 21562562

Amato SM, Orman MA, Brynildsen MP. 2013. Metabolic Control of Persister Formation in Escherichia coli. *Mol Cell* **50**:475–487. DOI: https://doi.org/10.1016/j.molcel.2013.04.002, PMID: 23665232

Andrews JM. 2001. Determination of minimum inhibitory concentrations. *Journal of Antimicrobial Chemotherapy* **48**:5–16. DOI: https://doi.org/10.1093/jac/48.suppl_1.5, PMID: 11420333

Balaban NQ, Helaine S, Lewis K, Ackermann M, Aldridge B, Andersson DI, Brynildsen MP, Bumann D, Camilli A, Collins JJ, Dehio C, Fortune S, Ghigo JM, Hardt WD, Harms A, Heinemann M, Hung DT, Jenal U, Levin BR, Michiels J, Storz G, Tan MW, Tenson T, Van Melderen L, Zinkernagel A. 2019. Definitions and guidelines for research on antibiotic persistence. *Nat Rev Microbiol* **17**:441–448. DOI: https://doi.org/10.1038/s41579-019-0196-3, PMID: 30980069

Becker B, Cooper MA. 2013. Aminoglycoside antibiotics in the 21st century. *ACS Chem Biol* **8**:105–115. DOI: https://doi.org/10.1021/cb3005116, PMID: 23110460

Binder S, Levitt AM, Sacks JJ, Hughes JM. 1999. Emerging Infectious Diseases: Public Health Issues for the 21st Century. *Science (1979)* **284**:1311–1313. DOI: https://doi.org/10.1126/science.284.5418.1311, PMID: 10334978

Borovinskaya MA, Pai RD, Zhang W, Schuwirth BS, Holton JM, Hirokawa G, Kaji H, Kaji A, Cate JHD. 2007. Structural basis for aminoglycoside inhibition of bacterial ribosome recycling. *Nat Struct Mol Biol* **14**:727–732. DOI: https://doi.org/10.1038/nsmb1271, PMID: 17660832

Bruni GN, Kralj JM. 2020. Membrane voltage dysregulation driven by metabolic dysfunction underlies bactericidal activity of aminoglycosides. *Elife* **9**:1–25. DOI: https://doi.org/10.7554/eLife.58706, PMID: 32748785

Datsenko KA, Wanner BL. 2000. One-step inactivation of chromosomal genes in Escherichia coli K-12 using PCR products. *PNAS* **97**:6640–6645. DOI: https://doi.org/10.1073/pnas.120163297, PMID: 10829079

Davis BD, Chen L, Tai PC. 1986. Misread protein creates membrane channels: An essential step in the bactericidal action of aminoglycosides. *PNAS* **83**:6164–6168. DOI: https://doi.org/10.1073/pnas.83.16.6164, PMID: 2426712

Doellinger J, Schneider A, Hoeller M, Lasch P. 2020. Sample preparation by easy extraction and digestion (SPEED) - A universal, rapid, and detergent-free protocol for proteomics based on

acid extraction. *Molecular and Cellular Proteomics* **19**:209–222. DOI:
https://doi.org/10.1074/mcp.tir119.001616, PMID: 31754045

Domingues MM, Inácio RG, Raimundo JM, Martins M, Castanho MARB, Santos NC. 2012.
Biophysical characterization of polymyxin B interaction with LPS aggregates and membrane
model systems. *Biopolymers* **98**:338–344. DOI: https://doi.org/10.1002/bip.22095, PMID:
23193598

Endimiani A, Hujer KM, Hujer AM, Armstrong ES, Choudhary Y, Aggen JB, Bonomo RA. 2009.
ACHN-490, a neoglycoside with potent in vitro activity against multidrug-resistant Klebsiella
pneumoniae isolates. *Antimicrob Agents Chemother* **53**:4504–4507. DOI:
https://doi.org/10.1128%2FAAC.00556-09, PMID: 19770287

Farha MA, Verschoor CP, Bowdish D, Brown ED. 2013. Collapsing the proton motive force to
identify synergistic combinations against Staphylococcus aureus. *Chem Biol* **20**:1168–1178.
DOI: https://doi.org/10.1016/j.chembiol.2013.07.006, PMID: 23972939

Germain E, Roghanian M, Gerdes K, Maisonneuve E. 2015. Stochastic induction of persister cells
by HipA through (p)ppGpp-mediated activation of mRNA endonucleases. *Proc Natl Acad
Sci U S A* **112**:5171–5176. DOI: https://doi.org/10.1073/pnas.1423536112, PMID: 25848049

Guo H, Suzuki T, Rubinstein JL. 2019. Structure of a bacterial ATP synthase. *Elife* **8**:e43128. DOI:
https://doi.org/10.7554/eLife.43128, PMID: 30724163

Kohanski MA, Dwyer DJ, Hayete B, Lawrence CA, Collins JJ. 2007. A Common Mechanism of
Cellular Death Induced by Bactericidal Antibiotics. *Cell* **130**:797–810. DOI:
https://doi.org/10.1016/j.cell.2007.06.049, PMID: 17803904

Kohanski MA, Dwyer DJ, Wierzbowski J, Cottarel G, Collins JJ. 2008. Mistranslation of
Membrane Proteins and Two-Component System Activation Trigger Antibiotic-Mediated
Cell Death. *Cell* **135**:679–690. DOI: https://doi.org/10.1016/j.cell.2008.09.038, PMID:
19013277

Krause KM, Serio AW, Kane TR, Connolly LE. 2016. Aminoglycosides: An overview. *Cold
Spring Harb Perspect Med* **6**. DOI: https://doi.org/10.1101/cshperspect.a027029, PMID:
27252397

Kresge N, Simoni RD, Hill RL. 2004. Selman Waksman: the Father of Antibiotics. *Journal of
Biological Chemistry* **279**:e7–e8. DOI: https://doi.org/10.1016/s0021-9258(20)67861-9

Lambert PA. 2005. Bacterial resistance to antibiotics: Modified target sites. *Adv Drug Deliv Rev
***57**:1471–1485. DOI: https://doi.org/10.1016/j.addr.2005.04.003, PMID: 15964098

Landman D, Babu E, Shah N, Kelly P, Bäcker M, Bratu S, Quale J. 2010. Activity of a novel
aminoglycoside, ACHN-490, against clinical isolates of Escherichia coli and Klebsiella
pneumoniae from New York City. *Journal of Antimicrobial Chemotherapy* **65**:2123–2127.
DOI: https://doi.org/10.1093/jac/dkq278, PMID: 20667885

Martinez II KA, Kitko RD, Mershon JP, Adcox HE, Malek KA, Berkmen MB, Slonczewski JL.
2012. Cytoplasmic pH Response to Acid Stress in Individual Cells of Escherichia coli and
Bacillus subtilis Observed by Fluorescence Ratio Imaging Microscopy. *Appl Environ
Microbiol* **78**:3706–3714. DOI: https://doi.org/10.1128/AEM.00354-12, PMID: 22427503

Miesenböck G, Angelis DA De, Rothman JE. 1998. Visualizing secretion and synaptic
transmission with pH-sensitive green fluorescent proteins. *Nature* **394**:192–195. DOI:
https://doi.org/10.1038/28190, PMID: 9671304

Moazed D, Noller HF. 1987. Interaction of antibiotics with functional sites in 16S ribosomal RNA.
Nature **327**:389–394. DOI: https://doi.org/10.1038/327389a0, PMID: 2953976

Mohiuddin SG, Kavousi P, Orman MA. 2020. Flow-cytometry analysis reveals persister resuscitation characteristics. *BMC Microbiol* **20**:1–13. DOI: https://doi.org/10.1186/s12866-020-01888-3, PMID: 32640993

Mohiuddin SG, Massahi A, Orman MA. 2022. High-Throughput Screening of a Promoter Library Reveals New Persister Mechanisms in Escherichia Coli. *Microbiol Spectr* **10**:e02253-21. DOI: https://doi.org/10.1128/spectrum.02253-21, PMID: 35196813

Molina-Quiroz RC, Silva-Valenzuela C, Brewster J, Castro-Nallar E, Levy SB, Camilli A. 2018. Cyclic AMP Regulates Bacterial Persistence through Repression of the Oxidative Stress Response and SOS-Dependent DNA Repair in Uropathogenic Escherichia coli. *mBio* **9**:e02144-17. DOI: https://doi.org/10.1128/mBio.02144-17, PMID: 29317513

Ngo HG, Mohiuddin SG, Ananda A, Orman MA. 2023. Unraveling Crp/cAMP-Mediated Metabolic Regulation in Antibiotic-Tolerant Cells of *Escherichia coli*. Under Review.

Okusu H, Ma D, Nikaido H. 1996. AcrAB Efflux Pump Plays a Major Role in the Antibiotic Resistance Phenotype of Escherichia coli Multiple-Antibiotic-Resistance (Mar) Mutants. *J Bacteriol* **178**:306–308. DOI: https://doi.org/10.1128/jb.178.1.306-308.1996, PMID: 8550435

Panta PR, Doerrler WT. 2021. A Burkholderia thailandensis DedA Family Membrane Protein Is Required for Proton Motive Force Dependent Lipid A Modification. *Front Microbiol* **11**:618389. DOI: https://doi.org/10.3389/fmicb.2020.618389, PMID: 33510730

Qin G, Zhang P, Sun M, Fu W, Cai C. 2022. Comprehensive spectral libraries for various rabbit eye tissue proteomes. *Sci Data* **9**:111. DOI: https://doi.org/10.1038/s41597-022-01241-5, PMID: 35351915

Recht MI, Douthwaite S, Dahlquist KD, Puglisi JD. 1999. Effect of Mutations in the A Site of 16 S rRNA on Aminoglycoside Antibiotic-Ribosome Interaction. *J Mol Biol* **286**:33–43. DOI: https://doi.org/10.1006/jmbi.1998.2446, PMID: 9931247

Ristuccia AM, Cunha BA. 1985. An overview of amikacin. *Ther Drug Monit* **7**:12–25. DOI: https://doi.org/10.1097/00007691-198503000-00003, PMID: 3887667

Schatz A, Bugie E, Waksman SA. 2005. Streptomycin, a substance exhibiting antibiotic activity against gram-positive and gram-negative bacteria. 1944. *Clin Orthop Relat Res* **437**:3–6. DOI: https://doi.org/10.1097/01.blo.0000175887.98112.fe, PMID: 16056018

Stokes JM, Yang K, Swanson K, Jin W, Cubillos-Ruiz A, Donghia NM, MacNair CR, French S, Carfrae LA, Bloom-Ackerman Z, Tran VM, Chiappino-Pepe A, Badran AH, Andrews IW, Chory EJ, Church GM, Brown ED, Jaakkola TS, Barzilay R, Collins JJ. 2020. A Deep Learning Approach to Antibiotic Discovery. *Cell* **180**:688-702.e13. DOI: https://doi.org/10.1016/j.cell.2020.01.021, PMID: 32084340

Szklarczyk D, Gable AL, Lyon D, Junge A, Wyder S, Huerta-Cepas J, Simonovic M, Doncheva NT, Morris JH, Bork P, Jensen LJ, Von Mering C. 2019. STRING v11: Protein-protein association networks with increased coverage, supporting functional discovery in genome-wide experimental datasets. *Nucleic Acids Res* **47**:D607–D613. DOI: https://doi.org/10.1093/nar/gky1131, PMID: 30476243

Szklarczyk D, Kirsch R, Koutrouli M, Nastou K, Mehryary F, Hachilif R, Gable AL, Fang T, Doncheva NT, Pyysalo S, Bork P, Jensen LJ, Von Mering C. 2023. The STRING database in 2023: protein-protein association networks and functional enrichment analyses for any sequenced genome of interest. *Nucleic Acids Res* **51**:D638–D646. DOI: https://doi.org/10.1093/nar/gkac1000, PMID: 36370105

Taber Harry W, Mueller JP, Miller PF, Arrow AS. 1987. Bacterial Uptake of Aminoglycoside Antibiotics. *Microbiol Rev* **51**:439–457. DOI: https://doi.org/10.1128/mr.51.4.439-457.1987, PMID: 3325794

Theodore A, Lewis K, Vulić M. 2013. Tolerance of Escherichia coli to fluoroquinolone antibiotics depends on specific components of the SOS response pathway. *Genetics* **195**:1265–1276. DOI: https://doi.org/10.1534/genetics.113.152306, PMID: 24077306

Van den Bergh B, Schramke H, Michiels JE, Kimkes TEP, Radzikowski JL, Schimpf J, Vedelaar SR, Burschel S, Dewachter L, Lončar N, Schmidt A, Meijer T, Fauvart M, Friedrich T, Michiels J, Heinemann M. 2022. Mutations in respiratory complex I promote antibiotic persistence through alterations in intracellular acidity and protein synthesis. *Nat Commun* **13**:546. DOI: https://doi.org/10.1038/s41467-022-28141-x, PMID: 35087069

Woodcock J, Moazed D, Cannon M, Davies J, Noller HF. 1991. Interaction of antibiotics with A- and P-site-specific bases in 16S ribosomal RNA. *EMBO J* **1**:3099–3103. DOI: https://doi.org/10.1002/j.1460-2075.1991.tb07863.x, PMID: 1915283

Woodruff HB. 2014. Selman A. Waksman, winner of the 1952 nobel prize for physiology or medicine. *Appl Environ Microbiol* **80**:2–8. DOI: https://doi.org/10.1128/AEM.01143-13, PMID: 24162573

Wu M, Maier E, Benz R, Hancock REW. 1999. Mechanism of interaction of different classes of cationic antimicrobial peptides with planar bilayers and with the cytoplasmic membrane of Escherichia coli. *Biochemistry* **38**:7235–7242. DOI: https://doi.org/10.1021/bi9826299, PMID: 10353835

Yu F, Haynes SE, Teo GC, Avtonomov DM, Polasky DA, Nesvizhskii AI. 2020. Fast Quantitative Analysis of timsTOF PASEF Data with MSFragger and IonQuant. *Molecular and Cellular Proteomics* **19**:1575–1585. DOI: https://doi.org/10.1074/mcp.TIR120.002048, PMID: 32616513

Data Availability

The proteomics data will be submitted to one of the NIH-designated repositories. All data generated or analyzed during this study have been incorporated into the manuscript and supplemental files.

Author Details

Mehmet Orman

William A. Brookshire Department of Chemical and Biomolecular Engineering, University of Houston

Contribution: Conceptualization, Data curation, Supervision, Funding acquisition, Investigation, Writing - original draft, Project administration, Writing – review and editing

For correspondence: morman@central.uh.edu

Competing interests: No competing interests declared

Rauf Shiraliyev

William A. Brookshire Department of Chemical and Biomolecular Engineering, University of
Houston

Contribution: Data curation, Investigation, Writing - original draft, review and editing

Competing interests: No competing interests declared

Acknowledgment

The authors would like to thank the members of Orman Lab for their help. This study was
supported by NSF CAREER 2044375 and NIH/NIAID R01-AI143643.

Proteomics experiments were conducted at the Mass Spectrometry Laboratory of Dr. Chengzhi
Cai at the University of Houston, with the associated service fees.