Metabolic mechanism of ceftazidime resistance in *Vibrio alginolyticus*

Shi-Rao Liu¹
Xuan-Xian Peng¹,²
Hui Li¹

¹Center for Proteomics and Metabolomics, State Key Laboratory of Biocontrol, Guangdong Province Key Laboratory for Pharmaceutical Functional Genes, School of Life Sciences, Sun Yat-sen University, University City, Guangzhou 510006, People’s Republic of China; ²Laboratory for Marine Fisheries Science and Food Production Processes, Qingdao National Laboratory for Marine Science and Technology, Qingdao 266071, People’s Republic of China

**Background:** Microbial metabolism confounds antibiotic efficacy. However, information regarding effect of metabolism on cephalosporin antibiotics-mediated killing and *Vibrio spp* is largely absent, although the drugs are widely used in clinic and the bacteria are pathogens to both human and aquaculture animals.

**Purpose:** This study explores the metabolome of cephalosporin antibiotic-resistant *Vibrio alginolyticus* and analyzes the role of bacterial metabolism in drug and multidrug-resistance.

**Results:** The metabolomes of isogenic ceftazidime-resistant *V. alginolyticus* (VA-RCAZ) and ceftazidime-sensitive *V. alginolyticus* (VA-S) were analyzed using gas chromatography -mass spectrometry. The metabolome of VA-RCAZ is characterized by inefficient respiration, an inefficient pyruvate cycle (P cycle), increased biosynthesis of fatty acids and decreased membrane proton motive force. This hypothesis was confirmed by the fact that furfural and malonate, inhibitors of pyruvate dehydrogenase and succinate dehydrogenase (P cycle enzymes), respectively, increased resistance of VA-RCAZ to antibiotics, while exposure to triclosan, to inhibit biosynthesis of fatty acids, decreased resistance.

**Conclusion:** These results contribute to our understanding of mechanisms of bacterial antibiotic-resistance and may lead to more effective approaches to treat, manage or prevent infections caused by antibiotic-resistant pathogens including those of the *Vibrio* species.

**Keywords:** antibiotic resistance, cephalosporin, central carbon metabolism, pyruvate cycle, biosynthesis of fatty acids, metabolomics

**Introduction**

Pathogenic *Vibrio* species cause intestinal disease in humans and systemic disease in fish and shrimp. The most common pathogenic *Vibrio* strains are *Vibrio alginolyticus*, *V. parahaemolyticus*, *V. vulnificus*, and *V. cholera*, which cause cholera.¹⁻⁴ Diseases caused by *Vibrio* pathogens have been treated or prevented with antibiotic drugs, with tremendous benefit to human health and animal breeding. However, widespread use, as well as misuse and/or inappropriate use of antibiotic therapy have promoted emergence and rapid spread of antibiotic-resistant species of *Vibrio* and other pathogens.⁵⁻⁸ Some *Vibrio* species develop multidrug resistance, and are no longer susceptible to ampicillin, cepotaxime, tetracycline, and chloramphenicol,⁹⁻¹⁰ posing a major challenge for health practitioners and a huge threat to human health. The process of developing new pharmaceutical agents to control antibiotic-resistant pathogens is slow and not a viable approach to manage the growing epidemic of multidrug-resistant infectious diseases,¹¹,¹² which are already a major global public health concern.¹³ Therefore, further understanding of antibiotic resistance mechanisms for control of these antibiotic-resistant pathogens is an important scientific issue.
Ongoing research demonstrates several mechanisms by which Vibrio and other bacterial species develop antibiotic resistance, including mutation, acquisition of resistance-conferring plasmids/episomes, modification or degradation of target sites, altered drug uptake/alteration of membrane permeability, and induction or upregulation of drug efflux. Recently, metabolic mechanisms have been revealed as a characteristic feature of antibiotic resistance in bacteria. For example, the abundance of glucose, glutamate, alanine, and fructose is greatly suppressed in kanamycin-resistant Edwardsiella tarda as determined by gas chromatography–mass spectrometry (GC-MS)-based metabolomics. Exposure to exogenous glucose, glutamate, alanine, or fructose restored susceptibility of multidrug-resistant bacteria including Edwardsiella tarda, Escherichia coli, V. parahaemolyticus, Klebsiella pneumoniae, Pseudomonas aeruginosa, and Staphylococcus aureus to kanamycin. The mechanism underlying this approach is that these metabolites promote the pyruvate cycle (P cycle), a previous unknown cycle that provides energy for bacteria, by substrate activation, which in turn increases production of nicotinamide adenine dinucleotide (NADH) and proton motive force (PMF) and stimulates uptake of antibiotic.

Here we analyzed the metabolome of ceftazidime (CAZ)-susceptible and -resistant V. alginolyticus (VA-S and VA-RCAZ, respectively) focusing on carbon, energy metabolism, and fatty acid biosynthesis, which appear to be altered in CAZ-resistant V. alginolyticus. Metabolic changes were validated using gene deletion mutants and measurement of enzyme activity. Finally, the effect of inhibitors targeting fatty acid biosynthesis or the P cycle on CAZ resistance was analyzed. These findings indicate that the inefficient P cycle and increased fatty acid biosynthesis promote resistance to CAZ in V. alginolyticus. These results increase our understanding of antibiotic resistance and the role of bacterial metabolism. They may also provide tools and/or knowledge for future new strategies to stop infections by multidrug-resistant human pathogens.

Materials and methods
Bacterial strains and culture conditions
The bacterial strain, V. alginolyticus V12G01, used in this study came from a collection of our laboratory. V. alginolyticus V12G01 was subcultured in Luria–Bertani (LB) medium with or without 1/2 minimum inhibitory concentration (MIC) of CAZ (0.1 µg/mL) for serial passages, which led to generation of 16 MIC CAZ-resistant V. alginolyticus (VA-RCAZ) and control (VA-S), respectively. A single colony was propagated in 3% NaCl LB (tryptone 10 g/L, yeast extract 5 g/L, NaCl 30 g/L) broth for 8 hours at 30°C. The cultures were diluted to 1:100 using fresh 3% NaCl LB medium and grown at 30°C. For growth curve, OD_{600} of the bacterial cultures was measured at the indicated time. For effect of carbon sources on bacterial growth, OD_{600} of the bacterial cultures was measured at 10 hours in medium with the indicated concentration of glucose, xylitol, fructose, sucrose, mannose, or maltose. Glucose, xylitol, fructose, and maltose were purchased from Sigma-Aldrich, and sucrose and mannose were provided by Sangon Biotech. At least three biologic replicates were performed.

Measurement of MIC
Measurement of MIC was performed as previously described. In brief, CAZ (Guangzhou Qiyun Biological Technology) was twofold serial diluted with LB broth ranging from 0.05 to 100 µg/mL CAZ. Ninety microliters of LB containing CAZ and 10 µL of logarithmic phase cells (10^7 colony-forming units [CFU]/mL) of VA-S or VA-RCAZ were incubated in microwell plate for 16 hours at 30°C. The lowest concentration showing no visible growth was recorded as the MIC. At least three biologic replicates were performed.

Metabolomics analysis
Bacterial sample preparation was carried out as previously described. In brief, equivalent cells were quenched with 1 mL of cold methanol and sonicated for 5 minutes at a 200 W power setting. Samples were centrifuged at 12,000 rpm for 10 minutes. Supernatant, containing 1 µg/mL ribitol (Sigma-Aldrich) as internal analytical standard, was transferred into a new Eppendorf tube and dried by vacuum centrifugation device (LABCONCO). The dried extracts were used for GC-MS analysis.

GC-MS analysis was performed with a variation on the two-stage technique. Briefly, samples were oximated with 80 µL methoxyamine hydrochloride (20 mg/mL, Sigma-Aldrich) in pyridine (Sigma-Aldrich) for 90 minutes at 37°C, and derivatization with identical volume of N-methyl-N-(trimethylsilyl)trifluoroacetamide (Sigma-Aldrich) for another 30 minutes. An aliquot (1 µL) of the derivative of the supernatant was injected into a 30 m × 250 µm internal diameter × 0.25 µm DB-5MS (Agilent) by splitless sampling and analyzed using GC-MS (Trace DSQ II system, Thermo Scientific). The separation conditions of GC-MS consisted of an initial temperature of 85°C (5 minutes) with a uniform increase to 270°C at a speed of 15°C min⁻¹ (5 minutes); 0.5 µL sample volume, splitless injection; injection temperature, 270°C; interface temperature, 270°C; ion source (EI) temperature, 30°C; ionization voltage, 70 eV; quadrupole temperature, 150°C; carrier gas, highly pure helium; velocity, 1.0 mL min⁻¹; and full scan way, 50–600 m/z. Each group
contains four independent biologic replicates with two technical replicates.

Spectral deconvolution and calibration were performed using AMDIS and internal standards as previously described. A retention time (RT) correction was performed for all the samples, and then the RT was used as reference against which the remaining spectra were queried and a file containing the abundance information for each metabolite in all the samples was assembled. Metabolites from the GC-MS spectra were identified by searching in National Institute of Standards and Technology (NIST11.L) Mass Spectral Library. Among the detected peaks of all chromatograms, 206 compound peaks were considered as endogenous metabolites and the same metabolite names were merged. The resulting data matrix was normalized by the concentrations of added internal standards and the total intensity. This file was then used for subsequent statistical analyses.

The abundance of a metabolite was scaled by total abundance of all metabolites in a sample as its relative abundance for further analysis. Z-score analysis scaled each metabolite according to a reference distribution, and calculated based on the mean and SD of reference. Hierarchical Clustering was completed in the R platform with the package gplots (https://cran.r-project.org/web/packages/gplots/) using the distance matrix. Multivariate statistical analysis included principal component analysis (SIMCA-P 12.0.1), which was used to discriminate sample patterns, to identify the metabolites and to minimize the interindividual variation’s influence. SPSS13.0 and Graphpad Prism 5 were used to draw the histogram as the scatter plot.

Measurement of enzyme activity
Measurement of enzyme activity was performed as previously described. In brief, 30 mL logarithmic phase (OD 600 1.0) of VA-S and VA-RCAZ was harvested and sonicated for 8 minutes at a 130 W power setting. Samples were centrifuged at 12,000 rpm for 10 minutes. Sample protein concentration was determined using a bicinchoninic acid assay (Beyotime Biotechnology, P0010). Enzyme activity was spectrophotometrically monitored reduction of MTT at 566 nm using the following reaction mixes: pyruvate dehydrogenase (PDH) and α-ketoglutaric dehydrogenase (KGDH) reaction mix (0.15 mM MTT, 0.5 mM PMS, 0.2 mM TPP, 2.5 mM MgCl2, 50 mM potassium phosphate buffer [pH 7]), 2 mM pyruvate, or 5 mM α-ketoglutaric acid potassium salt, isocitrate dehydrogenase (IDH) reaction mix (0.15 mM MTT, 1 mM PMS, 2.5 mM MgCl2, 100 mM Tris–HCl [pH 8.8], 3.5 mM isocitrate), succinate dehydrogenase (SDH), and malate dehydrogenase (MDH) reaction mix (0.15 mM MTT, 1 mM PMS, 50 mM potassium phosphate buffer [pH 7], 20 mM succinate, or 50 mM malate). Isocitrate lyase (ICL) activity was measured at OD 445 nm. The lysate was incubated with reaction mix (100 mM Tris–HCl [pH 7.7], 0.15 mM MgCl2, 60 µM l-cysteine hydrochloride monohydrate, 7 mM isocitrate) for 30 minutes, followed by addition of 0.2 mL 10% trichloroacetic acid to terminate the reaction and centrifugation for 5 minutes. Then, 0.1 mL supernatant was mixed with 20 µL 0.1% 2,4-dimethyl-1-nitrobenzene at 30°C for 30 minutes, and 0.2 mL 1.5 N NaOH was added to show colors. Citrate synthase (CS) activity was measured with the Citrate Synthase Assay Kit (Nanjing Jiancheng Biotechnology Institute). MTT, PMS, TPP, pyruvate, succinate, l-cysteine hydrochloride monohydrate, and 2,4-dimethyl-1-nitrobenzene were purchased from Sangon Biotech, and isocitrate and malate were purchased from Howei Pharm. At least three biologic replicates were performed.

Measurement of membrane potential
Measurement of membrane potential was performed as previously described. In brief, overnight VA-S and VA-RCAZ were diluted to 1:100 using fresh 3% NaCl LB medium and grown at 30°C for 3 hours (OD 600 1.0). The cultures were harvested at 8,000 rpm for 3 minutes, and washed three times with 30 mL sterile saline and resuspended in M9 minimal medium with NaAc (10 mM), MgSO4 (2 mM), and CaCl2 (0.1 mM) to 0.6 at OD 600. At 30°C, 10°C CFU/mL cells were stained with 30 µM DiOC6(3) (D14730, Thermo Fisher Scientific, Waltham, MA, USA) for 30 minutes. Samples were analyzed by flow cytometry (FACSCalibur, Becton Dickinson, San Jose, CA, USA). PMF was accounted according to the following computational formula: red fluorescence +green fluorescence + . Experiments were repeated in three independent biologic replicates.

Antibiotic bactericidal assay
Antibiotic bactericidal assay was performed as previously described. In brief, VA-S, VA-RCAZ, ATCC33787, ΔnqrA, and ΔnqrF were cultured for 12 hours at 30°C. Samples were harvested at 8,000 rpm for 3 minutes, and washed three times with 30 mL sterile saline and resuspended in M9 minimal medium with NaAc (10 mM), MgSO4 (2 mM), and CaCl2 (0.1 mM) to 0.6 at OD 600. Indicated concentrations of CAZ with or without furfual, malonate, NaN3, CCCP, or triclosan were added in a final volume of 5 mL. After incubation at 30°C for 6 hours, 100 µL aliquot samples were collected, serially diluted, and plated (10 µL aliquots) on LB medium containing 2.5% agar. The plates were cultured at 30°C for
8–10 hours. Only those dilutions generating 20–200 clones were counted to calculate CFU. Percent survival was determined by dividing the CFU obtained from a treated sample by the CFU obtained from control. For fatty acid-enabled killing of VA-S by CAZ, the bacteria were cultured in LB with 0.2 mM of palmitate or stearate for 12 hours. The collected cells were used for the antibiotic bactericidal assay described above. Furfural, malonate, triclosan palmitate, and stearate were purchased from Sangon Biotech, and Sigma-Aldrich provided NaN₃ and CCCP. Experiments were repeated in three independent biologic replicates.

Quantitative real-time PCR (qRT-PCR)
qRT-PCR was performed as previously described.²⁹ Briefly, total RNA was extracted using Trizol reagent following the manufacturer’s instruction (Thermo Fisher Scientific), and reverse transcription to cDNAs was carried out on 1 µg total RNA using PrimerScript™ RT Reagent Kit with gDNA Eraser (TaKaRa RR047A). Quantification of mRNA was measured by using real-time PCR system (BIO-RAD) and SYBRPremix Ex Taq™ II (TaKaRa RR820A). Gene-specific primers used for qRT-PCR are showed in Table 1 and reference 23. The \( \Delta C_t \) values were calculated by \( \Delta C_t = C_t(\text{target gene}) - C_t(\text{reference gene}) \), and the \( \Delta \Delta C_t \) values were calculated by \( \Delta \Delta C_t = \Delta C_t(\text{test sample}) - \Delta C_t(\text{calibrator sample}) \). Fold change of the target gene was calculated by \( 2^{-\Delta \Delta C_t} \).³⁰ Experiments were repeated in four independent biologic replicates.

**Results**

VA-\( R_{CAZ} \) are phenotypically different than VA-S

This study examines mechanisms that confer antibiotic resistance in two strains of *V. alginolyticus*, VA-S and VA-\( R_{CAZ} \), which are 16-fold differentially sensitive or differentially resistant to CAZ, respectively. More specifically, the minimal drug concentration to fully inhibit growth in liquid LB (MIC; see section “Materials and Methods” for exact growth conditions) was 0.2 µg/mL for VA-S and 3.2 µg/mL for VA-R (Figure 1A). Under these conditions, both percent survival depended on CAZ dose (Figure 1B) and slower growth rate was detected in VA-\( R_{CAZ} \) than VA-S (Figure 1C). In addition, VA-\( R_{CAZ} \) grew more slowly than VA-S in medium containing glucose, xylitol, mannose, or maltose (\( P<0.05 \)), while the growth rates of the two strains were similar in media containing fructose or sucrose (\( P>0.05 \)) (Figure 1D).

Metabolic profile of VA-\( R_{CAZ} \) is different from that of VA-S

The metabolomes of VA-\( R_{CAZ} \) and VA-S were analyzed by GC-MS as described (see section “Materials and Methods”). Two technical replicates were performed and a correlation coefficient equal to 0.997–1.0 was observed, indicating that the methodology is reliable (Figure 2A). Biologic roles of metabolites were defined using KEGG. The results showed that 36.23% (25), 31.88% (22), 14.49% (10), and 14.49% (10) of the metabolites were carbohydrates, amino acids, nucleotides, or lipids, respectively (Figure 2B). Four biologic replicates in each group and two technical repeats for each biologic replicate were performed, yielding a total of 16 data sets. The metabolite profiles of VA-\( R_{CAZ} \) and VA-S were compared and differences were displayed as a heat map (Figure 2C). A two-sided Wilcoxon rank-sum test coupled with a permutation test was used to compare metabolite abundance/distribution in VA-\( R_{CAZ} \) and VA-S. The results showed that 69 metabolites (\( P<0.05 \)) were differentially enriched or -depleted in VA-\( R_{CAZ} \) relative to VA-S, distributed as 31.88% carbohydrates, 31.88% amino acids, 13.04% nucleotides, and 14.49% lipids, respectively (Figure 2B). Four biologic replicates in each group and two technical repeats for each biologic replicate were performed, yielding a total of 16 data sets. The metabolite profiles of VA-\( R_{CAZ} \) and VA-S were compared and differences were displayed as a heat map (Figure 2C). A two-sided Wilcoxon rank-sum test coupled with a permutation test was used to compare metabolite abundance/distribution in VA-\( R_{CAZ} \) and VA-S. The results showed that 69 metabolites (\( P<0.05 \)) were differentially enriched or -depleted in VA-\( R_{CAZ} \) relative to VA-S, distributed as 31.88% carbohydrates, 31.88% amino acids, 13.04% nucleotides, and 14.49% lipids, respectively (Figure 2D, E). Calculated \( Z \)-values fell in the range −21.4 to 28.0 and the data are visualized on the \( Z \)-score plot shown in Figure 2F. These results indicate that the metabolic profiles of VA-S and VA-\( R_{CAZ} \) are substantially different from one another and that the difference is statistically significant.

**Table 1** Primers for QRT-PCR

| Gene | KEGG entry | primer | Sequence (5' → 3') |
|------|------------|--------|-------------------|
| tesA | N646_4268  | Forward| AGCATTTCTGGCCGATACAAC |
|      |            | Reverse| GAGGGGAAACCTTTTAGGC |
| tesB | N646_0021  | Forward| CTTTGATTCGTGCTAATGG |
|      |            | Reverse| ATAGAGTGTCAAAGTGGGC |
| fadD | N646_3023  | Forward| ATGGAGGGTGTAAGTGTT |
|      |            | Reverse| TATCCCAACTGTGCTAAGG |
| fadE | N646_1395  | Forward| CAACCGCAGACACACTCAAC |
|      |            | Reverse| GGAGGCGCTAAGACAGAAAAC |
| fadJ | N646_1310  | Forward| GCTGAAGTATGCAAACCCCAC |
|      |            | Reverse| AATGTTTGTAGCCGATATGCCCAC |
| fadB | N646_2210  | Forward| ACTGCCGAGTGTTTCTTG |
|      |            | Reverse| CCAGGCAAACCTTACGCTC |
| fadA | N646_1311  | Forward| CGAACAGTACGCCCATTGACT |
|      |            | Reverse| GCTAACAACGGTGATGACC |
| fadA | N646_2209  | Forward| TATAGCCGTCGAGCACAAG |
|      |            | Reverse| AACACCTTCTGTAATCCAG |
| atoB | N646_3909  | Forward| GATGTTATGATCGGATGATCACC |
|      |            | Reverse| ACGCCACGATAATCGCAC |
Metabolic pathways differentially up- or downregulated in Va-RcaZ

To better understand the metabolic differences between VA-S and VA-RcaZ, MetaboAnalyst (http://www.metaboanalyst.ca/) was used to identify pathways in the two strains that were differentially up- or downregulated. The analysis revealed that eleven pathways (P<0.05) are upregulated in VA-RcaZ, and these pathways involve metabolism of amino acids (alanine, aspartate, and glutamate; arginine and proline; glycine, serine, and threonine; cyano amino acids), glyoxylate and dicarboxylate, nitrogen, glutathione, butanoate, as well as biosynthesis of aminoacyl-tRNAs, and unsaturated fatty acids, and the tricarboxylic acid cycle (Figure 3A). However, all metabolites increased in the pathway affecting biosynthesis of unsaturated fatty acids, whereas some metabolites decreased and some increased in other metabolic pathways (Figure 3B). We conclude that biosynthesis of fatty acids is upregulated in VA-RcaZ and tentatively conclude that this pathway may play a direct or indirect role in CAZ resistance in V. alginolyticus.

Biomarkers of CAZ resistance

To identify biomarkers of CAZ resistance, the metabolic profiles of VA-RcaZ and VA-S were analyzed using orthogonal partial least-square discriminant analysis. Component (t[1]) distinguished VA-RcaZ from VA-S (Figure 4A). Discriminating variables were identified on an S-plot (Figure 4B). Cutoff values were ≥0.05 for absolute value of the covariance p and ≥0.5 for correlation p(corr). Larger weight and higher relevance metabolites are shown in red. Five putative biomarkers of CAZ resistance were identified, including stearic acid, palmitic acid, oxalic acid, and valine (all of which increase), and succinic acid, whose abundance decreases (Figure 4C). Palmitic and stearic acids play roles in biosynthesis of fatty acids, and succinic acid plays a role in the P cycles, respiration, and energy metabolism.

Decreased energy metabolism and inefficient P cycle in VA-RcaZ

Based on the results presented above and summarized in Figure 5A, we hypothesize that flux of acetyl-coA into and through the
Figure 2 Metabolic profiles of VA-S and VA-R<sub>caZ</sub>

Notes: (A) Abundance of metabolites quantified in samples over two technical replicates is shown. Pearson correlation coefficient between technical replicates varies between 0.997 and 1.0. (B) Percentage of metabolites in every category. (C) Heat map of differential abundance of metabolites (row). Yellow and blue indicate increase and decrease of the metabolites scaled to mean and sD of row metabolite level, respectively (see color scale). Four biologic replicates in each group and two technical repeats for each biologic replicate were performed, yielding a total of 16 data sets. (D) Percentage of differential abundant metabolites in every category. (E) Number of differential abundant metabolites increased and decreased in every category. (F) Z-score plots corresponding to the data in panel (A). The data from VA-R<sub>caZ</sub> are separately scaled to the mean and sD of VA-S. Each point represents one metabolite in one technical repeat and is colored by sample types (yellow: VA-S, blue: VA-R<sub>caZ</sub>).

Abbreviations: VA-R<sub>caZ</sub>, ceftazidime-resistant Vibrio alginolyticus; VA-S, ceftazidime-sensitive Vibrio alginolyticus.
Figure 3 Pathway analysis.

Notes: (A) Pathway enrichment of differential abundant metabolites. (B) Integrative analysis of differential abundant metabolites in significantly enriched pathways. Yellow and blue indicate increase and decrease of metabolites, respectively. Number shows the differential abundant metabolites.

Abbreviations: VA-RcaZ, ceftazidime-resistant Vibrio alginolyticus; VA-S, ceftazidime-sensitive Vibrio alginolyticus.
P-cycle and fatty acid biosynthesis is altered in VA-RCAZ (Figure 5B). This hypothesis is substantiated by the following observations: PDH transforms pyruvate into acetyl-CoA, CS catalyzes the condensation reaction of acetyl-CoA and oxaloacetate to form the citrate, IDH catalyzes the oxidative decarboxylation of isocitrate to produce α-ketoglutarate, KGDH converts α-ketoglutarate to succinyl-CoA, SDH catalyzes the oxidation of succinate to fumarate, and MDH reversibly catalyzes the oxidation of malate to oxaloacetate. Decreased PDH, CS, IDH, and OGDH activity supported the reduced citric acid and succinic acid, while increased SDH activity and decreased MDH activity are expected to lead to elevated fumaric acid and malic acid (Figure 5C). ICL catalyzes the cleavage of isocitrate to succinate and glyoxylate. To preclude the possibility that the glyoxylate cycle is activated in VA-RCAZ, we showed that ICL activity is lower in VA-RCAZ than in VA-S (Figure 5C). iPPath was also used to compare the metabolomes of VA-RCAZ and VA-S and the results are summarized in an overview map, on which red lines show upregulated pathways and blue lines show downregulated pathways, the latter including carbohydrate metabolism (Figure 5D). The importance of the P cycle in the CAZ resistance was demonstrated by the fact that cell survival decreased in a dose-dependent manner in the presence of increasing concentration of furfural and malonate, which inhibit PDH and SDH of the P cycle, respectively (Figure 5E, F).

Decreased NQR and PMF in VA-RCAZ

We hypothesized that antibiotic resistance may be associated with impaired energy metabolism leading to low PMF, and reduced capacity for active transport of drugs and other metabolites across the cell membrane. In fact, membrane potential was significantly lower in VA-RCAZ than in VA-S (Figure 6A). Consistent with this, VA-RCAZ is also resistant to killing by gentamicin, whose cytotoxicity also requires adequate PMF,19,22,31 and higher percent survival as found in Figure 6B. We also investigated the possible role of sodium-dependent NADH: ubiquinone oxidoreductase (Na⁺-NQR), the main ion transporter in Vibrio species, in the mechanism of antibiotic resistance in V. alginolyticus. Na⁺-NQR is a multisubunit (NqrA-F) respiratory complex that couples the free energy of electron transfer reactions to electrogenic pumping of sodium across the Vibrio cell membrane. Loss of nqrA or

Figure 4 Metabolites up- or downregulated in VA-RCAZ.
Notes: (A) Principal component analysis of VA-S and VA-RCAZ. Each dot represents the technologic replicate analysis of samples in the plot. (B) S-plot generates from orthogonal partial least-square discriminant analysis. Predictive component p[1] and correlation p(corr)[1] differentiate VA-S and VA-RCAZ. Each dot represents metabolites and candidate biomarkers are highlighted in red dot. (C) Candidate biomarkers by predictive component p[1], p(corr)[1], and enriched pathways. Results (C) are displayed as mean ± SEM, and significant differences are identified as determined by Student’s t-test. **P<0.01. At least three biologic repeats were carried out.
Abbreviations: VA-RCAZ, ceftazidime-resistant Vibrio alginolyticus; VA-S, ceftazidime-sensitive Vibrio alginolyticus.
Figure 5 P cycle and energy metabolism in VA-RcaZ.

Notes: (A) Abundance of five metabolites in the P cycle. (B) Flux of the P cycle and biosynthesis of fatty acids in VA-RcaZ compared with VA-s. (C) Activity of enzymes (PDh, CS, IDh, OGDh, SDh, MDh, and Icl) in the P cycle. Overnight VA-s and VA-RcaZ were diluted to 1:100 using fresh 3% NaCl LB medium and grown at 30°C for 3 hours. (D) Interactive Pathways Explorer (iPath) analysis. Metabolic network pathways in VA-RcaZ were analyzed with iPath 2.0 (http://pathways2.embl.de/iPath2.cgi#).

Analyses of the 42 differential metabolites (the other metabolites are not recognized in the metabolic network pathways). Red: increase; blue: decrease. (E) Percent survival of VA-s in the presence or absence of inhibitors (furfural and malonate). Overnight VA-s were incubated with caZ (5 μg/ml) plus furfural (5, 20, and 80 μM) or malonate (5, 10, and 20 mM) in M9 minimal medium plus NaAc (10 mM) at 30°C for 6 hours. (F) Enzyme activity of PDH and SDH in the presence or absence of inhibitors (furfural and malonate). Overnight VA-s were incubated with furfural (80 μM) or malonate (20 mM) in M9 minimal medium plus NaAc (10 mM) at 30°C for 6 hours. Results (A, C, E, and F) are displayed as mean ± SEM, and significant differences are identified as determined by Student’s t-test. *P<0.05, **P<0.01. At least three biologic repeats were carried out.

Abbreviations: CAZ, ceftazidime; ICL, isocitrate lyase; IDH, isocitrate dehydrogenase; MDH, malate dehydrogenase; P cycle, pyruvate cycle; PDH, pyruvate dehydrogenase; SDH, succinate dehydrogenase; TCA, tricarboxylic acid; VA-RcaZ, ceftazidime-resistant Vibrio alginolyticus; VA-S, ceftazidime-sensitive Vibrio alginolyticus.
nqrF lowers membrane potential (Figure 6C), and increases percent survival in the presence of toxic small molecule drugs, including antibiotics (Figure 6D). NaN₃ and CCCP, inhibitors of respiration and membrane potential, respectively, also increased percent survival of VA-S in a dose-dependent manner (Figure 6E). These results suggest that defects in energy metabolism and low membrane potential play roles in the mechanism of CAZ resistance in V. alginolyticus.

Elevated fatty acid metabolism in CAZ-resistant V. alginolyticus

Triclosan, an inhibitor of fatty acid biosynthesis, decreases resistance and increases dose-dependent lethal effects of CAZ on VA-R_CAZ (Figure 7A), while exogenous palmitic acid or stearic acid increases resistance of VA-S to CAZ (Figure 7B). This is consistent with the upregulation of fatty acid biosynthesis and downregulation of fatty acid degradation in VA-R_CAZ, as suggested by metabolomics data presented above. This result was confirmed using qRT-PCR to quantify expression of genes in the two pathways. The results showed upregulation of 10 genes involved in fatty acid biosynthesis (N646_1132 and N646_1439 are fabH) while expression of 13 genes in fatty acid biosynthesis were unchanged. Furthermore, expression of five of seven genes involved in fatty acid degradation was lower in VA-R_CAZ than in VA-S (N646_1311 and N646_2209 are fadA), while expression of one gene in this pathway increased and expression of one gene did not change (Figure 7C, D).

Discussion

It has been proposed that deregulation of critical cellular metabolic pathways, including carbon and energy metabolism, plays a role in or can modulate resistance of pathogenic bacteria to small-molecule drugs, including antibiotics. Here, we analyzed the metabolomes of CAZ-resistant and
CAZ-sensitive *V. alginolyticus* using GC-MS and identified pathways and biomarkers that could play a direct or indirect role in CAZ resistance in *V. alginolyticus*. More specifically, the results suggest that an inefficient P cycle, low/altered energy production/metabolism, and increased biosynthesis of fatty acids contribute to the resistance of VA-RCAZ to CAZ. Importantly, these findings could support future development of methods to restore susceptibility of *Vibrio* species to cephalosporin antibiotics.

GC-MS-based metabolomics has been used to investigate metabolic characteristics in antibiotic-resistant bacteria, and metabolic response to antibiotics in bacteria.19,22,32,33 Peng et al.
al identified alanine, aspartate, and glutamate metabolism and decreased abundance of alanine, glutamate, glucose, and fructose as the key pathways and crucial biomarkers in kanamycin-resistant *Edwardsiella tarda*, respectively. Exogenous alanine, glutamate, glucose, or fructose reprogrammed antibiotic-resistant metabolome to antibiotic-sensitive metabolome and thereby promoted antibiotic uptake, which was P cycle dependent. In addition, Ye et al indicated that alanine metabolism promoted ROS production and glucose elevated glycine, serine, and threonine metabolism, which facilitated kanamycin-mediated killing of antibiotic-resistant bacteria. The present study investigated the mechanism of CAZ resistance in VA-R<sub>CAZ</sub> and found results similar to those previously reported for kanamycin-resistant *Edwardsiella tarda*. In particular, an inefficient P cycle, decreased abundance of metabolites in the upper cycle, and increased abundance of metabolites in the lower cycle were observed in both drug-resistant strains. We confirmed low activity of PDH, CS, OGDH, MDH, and ICL, and high activity of SDH in VA-R<sub>CAZ</sub> and showed that furfural and malonate inhibit PDH and SDH and increase resistance of VA-R<sub>CAZ</sub> to CAZ.

We note that upregulated fatty acid biosynthesis is the most characteristic metabolic feature of VA-R<sub>CAZ</sub>. Our results also show decreased degradation of fatty acids in VA-R<sub>CAZ</sub>. However, we hypothesize that an inefficient P cycle, where high and low activity of SDH (to reversibly catalyze the oxidation of malate to oxaloacetate) lead to decreased abundance of succinic acid and increased abundance of fumaric acid. This promotes flux of acetyl-CoA into fatty acid biosynthesis and links it to altered energy/carbon metabolism in drug-resistant *Vibrio* species.

VA-R<sub>CAZ</sub> also show altered energy metabolism, deregulation of Na<sup>+</sup>-NQR, and low membrane potential, all factors that could potentiate and/or contribute to resistance to CAZ (note that Δ*nqrA* and Δ*nqrF* mutations are reported to decrease susceptibility to CAZ). Consistent with this, membrane potential was reduced in VA-R<sub>CAZ</sub>, Δ*nqrA*, and Δ*nqrF*, and drug resistance of VA-S was potentiated in the presence of respiratory chain inhibitors.

**Conclusion**

The present study explores the relationship between *V. alginolyticus* metabolome and CAZ resistance, and identifies an altered and inefficient P cycle, increased biosynthesis of fatty acids, deregulation of Na<sup>+</sup>-NQR and low membrane PMF (Figure 8). These findings provide insight into mechanisms of resistance to CAZ and other antibiotics, and insight into multidrug resistance in *Vibrio* and other human pathogens. Importantly, the findings may also support future development of drugs that can prevent and/or manage infections caused by drug-resistant human pathogens.

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**Disclosure**

The authors report no conflicts of interest in this work.

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