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Bayesian Modeling and Intrabacterial Drug Metabolism Applied to Drug-Resistant *Staphylococcus aureus*

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ABSTRACT: We present the application of Bayesian modeling to identify chemical tools and/or drug discovery entities pertinent to drug-resistant *Staphylococcus aureus* infections. The quinoline JSF-3151 was predicted by modeling and then empirically demonstrated to be active against *in vitro* cultured clinical methicillin- and vancomycin-resistant strains while also exhibiting efficacy in a mouse peritonitis model of methicillin-resistant *S. aureus* infection. We highlight the utility of an intrabacterial drug metabolism (IBDM) approach to probe the mechanism by which JSF-3151 is transformed within the bacteria. We also identify and then validate two mechanisms of resistance in *S. aureus*: one mechanism involves increased expression of a lipocalin protein, and the other arises from the loss of function of an azoreductase. The computational and experimental approaches, discovery of an antibacterial agent, and elucidated resistance mechanisms collectively hold promise to advance our understanding of therapeutic regimens for drug-resistant *S. aureus*.

KEYWORDS: Bayesian modeling, Staphylococcus aureus, quinoline, intrabacterial drug metabolism, YceI, AzoR

S. aureus is a coagulase, catalase, and Gram-positive cocci residing on the skin and anterior nares of about 30% of healthy individuals.^{1,2} Considered to be the most pathogenic of the Staphylococci, S. aureus is responsible for a multitude of human diseases, which are further complicated by its capability to cause infection by at least two primary mechanisms: direct tissue invasion and exotoxin production. Direct tissue invasion is the primary cause of pathogenesis in humans, of which skin and soft tissue infections, such as abscesses, have become the most common condition to be treated by physicians.^{3,4} Other significant diseases caused by S. aureus invasion of tissue include osteomyelitis, endocarditis, and pneumonia. Metastasis of any of these focal infections can lead to bacteremia and sepsis. On the other hand, toxin-mediated diseases caused by S. aureus include food poisoning, toxic shock syndrome, and scalded skin syndrome. Thus, S. aureus is a major cause of disease and burden for the population and for medical practitioners.

Traditionally, treatment for *S. aureus* infections has included β -lactams, such as penicillin G, nafcillin, and oxacillin.^{2,5} This class of antibacterials targets transpeptidation (cross-linking) of peptidoglycan stem peptides. In 1959, methicillin—another antibiotic in this family—was licensed for use against *S. aureus* in England. A year later, reports of methicillin-resistant *S. aureus* (MRSA) surfaced in Europe, and in 1968, the first

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Table 1. Biological Profiling of the Most Active Compounds versus MRSA^a

Compound ID (JSF <i>-</i> #)	Chemical Structure	Model/Score	MRSA ATCC 43300 MIC in μg/mL (μM)	MSSA ATCC 25923 MIC in μg/mL (μM)	Vero cell CC₅₀ in μg/mL (μM)
Z27692293		MRSA_1a/7.80	12	12	>50
(3145)		MRSA_1b/-17.7	(25)	(25)	(>100)
Z27646636		MRSA_1a/9.95	25	25	>50
(3144)		MRSA_1b/-12.6	(50)	(50)	(>100)
Z296972134	ON SHARE	MRSA_1b/13.2	4.0	1.6	>50
(3151)		MRSA_1a/6.77	(12)	(4.8)	(>150)
Z1542057255	on of the second	MRSA_1b/13.4	6.2	6.2	0.78
(3157)		MRSA_1a/6.94	(18)	(18)	(2.0)
Z364195972	CHO H CO H	MRSA_1b/14.2	3.1	6.2	1.6
(3153)		MRSA_1a/7.45	(10)	(20)	(5.0)

^aThe MIC values are representative of three independent experiments. The model inherent in the selection of the molecule from the top-scoring 50 compounds is listed first in the Model/Score column.

hospital outbreak of MRSA in the United States was recognized.⁶ Resistance to methicillin is conferred by the mecA gene, which encodes PBP2a, a penicillin-binding protein with low affinity for β -lactams.⁷ During the period of 1987 to 2014, surveillance data have shown that S. aureus was a significant source of hospital-acquired infections, and nearly 50% of all hospital-acquired *S. aureus* infections were due to MRSA.^{5,8–10} Nearly 20 years ago, treatment options for MRSA included vancomycin and trimethoprim/sulfamethoxazole.⁵ Unfortunately, therapeutic options for MRSA infections have changed little, and reports of vancomycin-intermediateresistant and vancomycin-resistant S. aureus (VISA and VRSA, respectively) make this notion more alarming.¹¹⁻¹³ A thorough review by Kurosu et al. shows that the pipeline for agents against MRSA contains few options.¹⁴ Therapies such as dalbavancin, oritavancin, and tedizolid, though approved for use, are not utilized in the inpatient setting due to cost and lack of efficacy data for bacteremia and sepsis.¹⁵

The majority of chemical space has yet to be explored for the seeds of new treatments for drug-resistant *S. aureus* infections. Herein, we present the application of a computational approach to advance drug discovery. Bayesian modeling, previously validated for use with data sets for the growth inhibition of *Mycobacterium tuberculosis*^{16–19} or *Neisseria* gonorrhoeae,²⁰ is leveraged to discover and validate novel hit compounds against drug-resistant *S. aureus*. We demonstrate the successful prediction of *in vitro* growth inhibitory activity for quinoline JSF-3151, we proceed to show its value as a bactericidal agent against clinically relevant drug-resistant *S. aureus* strains with *in vivo* efficacy in a mouse model of infection, and we utilize genetic methods to identify and validate two mechanisms of resistance while probing their effects on JSF-3151 through intrabacterial drug metabolism (IBDM).^{21,22}

RESULTS

Development of the Bayesian Model. The PubChem Bioassay database (https://pubchem.ncbi.nlm.nih.gov) was queried with the keywords "aureus and MIC." We ultimately selected the studies with the largest number of compounds tested against drug-resistant *S. aureus* to develop our training sets. Inspection of each data set was conducted to remove duplicates, choose the most conservative value of the MIC (minimum concentration of compound to inhibit bacterial growth *in vitro* by 90%), convert units from μ M to μ g/mL, and ensure that the strain being tested was drug-resistant. We found that the assays used *S. aureus* strains resistant to one or more of the following drugs: methicillin, linezolid, vancomycin,

or fluoroquinolones. In total, our full training set (MRSA 1a) contained 1633 compounds, of which 1043 compounds (63.8%) were designated as active (MIC $\leq 10 \ \mu g/mL$). This level of growth inhibition was deemed to be acceptable for a hit compound, based on our experience. The MRSA 1a training set encompassed a small yet fairly diverse portion of chemical space, which included known antibacterial scaffolds such as tetracyclines, fluoroquinolones, and β -lactams. However, it also included rhodanines, which are known to be promiscuous Pan Assay Interference compounds, or "PAINS".²³ Since our aim was to find a novel chemical entity with specificity against drug-resistant S. aureus, we hypothesized that removing (or pruning) these potentially problematic compounds as well as known antibacterial chemotypes from the training set would help teach the model to focus on new, high-quality chemical scaffolds. This type of structural pruning of the training set may be compared to our activity-based pruning utilized with models for mouse liver microsomal stability and Vero cell cytotoxicity.²⁴⁻²⁶ Additionally, a novel chemotype should have a heightened likelihood of operating via a novel mechanism of action. Utilizing this strategy, we manually pruned the active tetracyclines, fluoroquinolones, rhodanines, other known PAINS, and β -lactams from the actives of MRSA 1a to produce a new training set (MRSA 1b) that contained 1247 compounds, of which 657 compounds were classified as active (52.7%). These data sets were used as independent training sets to create two Bayesian models in Pipeline Pilot 9.1 (BIOVIA). We used the eight standard physiochemical descriptors plus the structural fingerprints FCFP6 (functional class fingerprints of maximum diameter 6, which utilize bits for different functional groups). For internal statistics obtained through 5-fold internal crossvalidation, MRSA_1a and MRSA_1b displayed Receiver Operator Characteristic (ROC) scores of 0.903 and 0.917, respectively. The sensitivity values were 93.5% and 96.5%, respectively, while the specificity metrics were 81.7% and 81.0%, respectively (Supplementary Table 1). The models were then assessed with an external test set of 374 in-house compounds (synthesized by the Freundlich laboratory and assayed versus MRSA ATCC 43300, of which 6.2% were active). A comparison was also included with a methicillinsensitive S. aureus (MSSA) model (Broad MSSA), constructed with screening data extracted from ChemBank (http://chembank.broadinstitute.org) (Supplementary Tables 1 and 2). The positive predictive value (Supplementary Table 2) demonstrated the order of performance to be MRSA 1a >MRSA 1b \gg Broad MSSA (Supplementary Figure 1).

The MRSA 1a and MRSA 1b models were applied in a prospective virtual screen of a library of three million commercially available compounds from Enamine (http:// www.enamine.net) accessed through the ZINC server (http:// zinc.docking.org). For each compound in the library and each of the two MRSA models, a Bayesian score was calculated, such that the value scaled with the probability of the molecule being efficacious given our activity cutoff of 10 μ g/mL. The top-scoring 15 candidates (10 from MRSA 1a and 5 from MRSA_1b) and bottom-scoring 8 candidates from MRSA_1b were assessed for in vitro growth inhibition activity. The MIC values against MRSA (heretofore referring to ATCC 43300 unless noted otherwise) and MSSA (heretofore referring to ATCC 25923 unless noted otherwise) strains were determined along with the Vero cell CC50 (minimum concentration of compound resulting in 50% growth inhibition of this model

mammalian cell line) (Supplementary Table 3). Of the 15 top scoring compounds assayed, three of the five most active compounds exhibited an MIC $\leq 10 \ \mu g/mL$ versus the MRSA strain (Table 1). All three hits were from the MRSA 1b model. Interestingly, the highest scoring compound from MRSA 1a, the fluoroquinolone JSF-3144, displayed an MIC of 25 μ g/mL (50 μ M) against both MRSA and MSSA strains. ISF-3144 and ISF-3145 did not score within the range (11.7-15.3) for the top 50 compounds from the MRSA 1b model. The three MRSA 1b selected actives either scored just outside (JSF-3151, JSF-3157) of or within (JSF-3153) the range (7.06-9.95) for the top 50 compounds from the MRSA 1a model. Of the eight MRSA_1b bottom-scoring compounds in the Enamine library, all were inactive with an MIC > 50 μ g/ mL (Supplementary Table 4). As evidenced by our results with the bottom- and top-scoring compounds, the Bayesian models are able to enrich for active compounds while triaging inactives. Furthermore, as seen with our previous Bayesian models, the results also validate the utility of pruning.²⁴⁻²⁶

Quinoline JSF-3151 proved to be the most promising of the three hits based on its MIC values of 4.0 μ g/mL (12 μ M) and 1.6 μ g/mL (4.8 μ M) versus MRSA and MSSA strains, respectively, and its Vero cell CC_{s0} value of >50 μ g/mL (>150 μ M). This gave JSF-3151 MRSA and MSSA selectivity indices (a ratio of mammalian cell cytotoxicity to antibacterial potency, SI = CC_{50}/MIC) of >12 and >31, respectively. The other two hits, JSF-3153 and JSF-3157, failed to exhibit acceptable SI values (≥ 10).²⁷ Searches of the SciFinder (http://scifinder.cas.org), PubChem (https://pubchem.ncbi. nlm.nih.gov), and ChEMBL databases (https://www.ebi.ac. uk/chembl) supported the novelty of our finding of the antibacterial efficacy of JSF-3151. To further characterize the compound, we determined the minimum bactericidal concentration (MBC) against both MRSA and MSSA strains. The MBC values of 0.39–0.78 μ g/mL (1.2–2.3 μ M) versus MSSA and 1.6-3.1 µg/mL (4.8-9.3 µM) versus MRSA demonstrated the compound is capable of killing $\geq\!\!2 \,\log_{10}$ colonyforming units (CFU) per mL (Supplementary Figure 2). Furthermore, a time-kill curve showed that JSF-3151 at $1 \times$ MIC sterilized a MRSA culture within 4 h, more than 6 times faster than vancomycin at the same fold MIC (Supplementary Figure 2). Because of the increasing emergence of vancomycin resistance in the world,^{12,13} we also quantified the MIC of JSF-3151 against VISA and VRSA strains. Gratifyingly, we found that JSF-3151 exhibited even greater potency against six clinical isolates of VISA and five clinical isolates of VRSA, as compared with MRSA (Supplementary Table 5).

In Vivo Study of JSF-3151 in a Mouse Peritonitis Model. To support mouse efficacy studies, we quantified the mouse pharmacokinetic (PK) profile of JSF-3151, demonstrating an inability to reach the MIC over a 5 h time period (Supplementary Figure 3). This outcome is consistent with the compound's poor kinetic aqueous solubility (S \leq 0.060 μ M) and low metabolic stability (half-life $t_{1/2}$ of 1.46 min) in the presence of mouse liver microsomes. Given its poor solubility, JSF-3151 provided a suspension when formulated in sterile saline. JSF-3151 was dosed in BALB/c mice at 24 mg/kg-the highest achievable dose given its solubility-via the intraperitoneal route. Nevertheless, 2 out of 12 (16.7%) mice infected with the MRSA COL strain survived 3 days postadministration of a single dose of JSF-3151, as compared with 0/12 with the vehicle-only control and 12/12 with the 100 mg/kg subcutaneous dose of vancomycin (Supplementary

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Table 2. MIC Profiling of JSF-3151 versus Spontaneous Resistant Mutants^a

strain name	strain notes	JSF-3151	ampicillin	doxycycline	rifampicin
USA300_FPR3757	parent	1.6	>50	<0.25	<0.25
JMB1100	USA300_LAC cured of pUSA03	3.1	>50	<0.25	< 0.25
USA300_FPR3757 yceI1	$T \rightarrow G$ mutation at position -43 in yceI	>50	>50	< 0.25	< 0.25
USA300_FPR3757 yceI2	adenosine base insertion at -37 in <i>yceI</i>	>50	>50	< 0.25	<0.25
USA300_FPR3757 yceI3	adenosine base deletion at position -41 in yceI	>50	>50	< 0.25	< 0.25
ATCC 43300	MRSA	3.1	32	<0.25	<0.25
ATCC 25923	MSSA	3.1	<0.25	<0.25	<0.25

^{*a*}The MIC values are representative of three independent experiments.



Figure 1. IBDM studies of JSF-3151 and JSF-3640. (A) Time course study depicting the intrabacterial accumulation of JSF-3151 within *S. aureus* strains JMB1100 (WT), *yceI2*, and *azoR::Tn* that were treated with JSF-3151. (B) Time course study depicting the intrabacterial accumulation of JSF-3640 within *S. aureus* strains JMB1100 (WT), *yceI2*, and *azoR::Tn* that were treated with JSF-3151. (C) High-resolution mass spectrometry supports assignment of JSF-3640 as the amine metabolite, formed from JSF-3151, based on its spectral signature.

Figure 3). This result, perhaps not surprising since the Bayesian model was only trained to predict *in vitro* efficacy and not aqueous solubility, mouse liver microsomal stability, and *in vivo* efficacy, suggests that analogues of JSF-3151 with optimized PK properties could demonstrate improved efficacy in this infection model.

Analysis of Drug Metabolism Demonstrates Intrabacterial Reduction of JSF-3151. We next sought to probe the mechanism of action for JSF-3151 and began with the determination as to whether JSF-3151 accumulates within the bacterium using our recently developed approach to studying intrabacterial drug metabolism (IBDM).^{21,22} Briefly, a MRSA WT strain (JMB1100; Table 2) was incubated with 10× MIC of JSF-3151 and was then sampled at 0, 15, 30, 45, and 90 min post-treatment. Following the quenching of metabolism and extraction of the cellular contents soluble in -80 °C 2:2:1 acetonitrile:methanol:water, analysis by liquid chromatography–mass spectrometry (LC-MS) showed a time-dependent accumulation of JSF-3151 within the bacterium (Figure 1A). We also noted a time-dependent increase of a new species labeled JSF-3640 (Figure 1B) whose high-resolution mass spectrometry (HRMS) (Figure 1C) data were consistent with the corresponding amine metabolite of JSF-3151:5-amino-*N*-(6-chloroquinolin-8-yl)thiophene-2-carboxamide. Critically, the synthetic metabolite coeluted with the authentic metabolite on the LC (Supplementary Figure 4) and exhibited the same low-resolution and high-resolution mass spectrum profiles. Together these results support that JSF-3640 is, indeed, the amine metabolite of JSF-3151. The amine metabolite was whole-cell inactive (MIC > 50 μ g/mL versus MRSA), despite the ability of JSF-3640 to accumulate within MRSA in a time-dependent fashion (Supplementary Figure 4).

Membrane Depolarization and NO[•] Release Are Ruled Out with Regard to the JSF-3151 Mechanism of Action. Subsequently, we sought to rule out the potential for a nonspecific mechanism of action such as membrane depolarization. MRSA was incubated with DiBAC₄, bis(1,3dibutylbarbituric acid)trimethine oxonol, which demonstrates increased fluorescence upon binding to proteins of a depolarized cell,^{28,29} and then treated with 4.5× MIC of JSF-3151. JSF-3151 failed to compromise the cell membrane, as compared with the positive control sodium deoxycholate (Supplementary Figure 5).

Given the propensity for nitroheterocycles to extrude nitric oxide (NO^{\bullet}) , 30^{-32} we examined the ability of JSF-3151 to produce NO[•]. We assessed intrabacterial NO[•] release via the dye 4-amino-5-methylamino-2',7'-difluorofluorescein diacetate (DAF-FM diacetate).³³ In addition, the Griess reagent was used to detect the presence of nitrite in the supernatant, which is the oxidation product of NO[•] that undergoes efflux from the bacterium.³⁴ Despite treatment of MRSA with 100 µM JSF-3151, neither DAF-FM (over 2 h) nor the Griess reagent (over 48 h) evidenced NO[•] production (Supplementary Figure 5), as compared with the DEA-NONOate positive control. Consistent with these experiments was our observation that the des-nitro JSF-3151 (N-(6-chloroquinolin-8-yl)thiophene-2carboxamide; synthesized independently as JSF-3587) did not accumulate within the cell according to our above IBDM studies. The NO[•] detection assays and our IBDM platform suggest that the bactericidal efficacy of JSF-3151 is not attributed to NO[•] formation.

A Bacterial Lipocalin and an Azoreductase Are Linked to Drug Resistance. Using the MRSA strain LAC (Table 2), we used a genetic screen to select for resistant mutants using 8× MIC of JSF-3151. Single colonies were apparent following 3 d of incubation with a frequency of $6 \times$ 10⁻⁸. In comparison, resistance frequencies for front-line agents such as vancomycin, ciprofloxacin, and linezolid at 8× their respective MIC value have been reported^{13,35,36} to range between 1×10^{-9} and 1×10^{-6} . Ampicillin, erythromycin, doxycycline, rifampicin, and vancomycin exhibited the same MIC versus both JSF-3151-resistant strains and the parental strain. Whole-genome sequencing of three representative mutants, demonstrating >16-fold loss of susceptibility to JSF-3151, exhibited three distinct mutations in the promoter region of the gene SAUSA300_2620 (Table 2). Previously published reports suggested that the SAUSA300 2620 gene product, annotated as a YceI-like protein, may have a similar function to the Burkholderia cenocepacia lipocalin (BcnA), which has been demonstrated to confer resistance to hydrophobic small molecules through their selective binding.^{37,38} Sanger sequencing of these representative mutants, labeled as yceI1, yceI2, and *vceI3*, further confirmed the presence of these mutations in the promoter region (Table 3). We hypothesized that the mutations in the promoter region led to an increase in production of YceI, which consequently affords resistance to JSF-3151.

To explore this hypothesis, our first goal was to observe the effect of each mutation with respect to JSF-3151 resistance. We cloned a PCR fragment containing -290 base pairs upstream of the predicted translational start site and *yceI* into a chromosomal integration vector (pLL39).³⁹ Five constructs were generated: empty vector, WT promoter+gene, and each of the three mutated promoter+gene sequences. Each of these plasmids was individually integrated into *S. aureus* strain RN4220 in the L54a *attB* locus.^{39,40} Integrates were verified using PCR (Supplementary Table 6). The individual integrated episomes were transduced into two different *S. aureus* strains: parental USA300 LAC (JMB1100) and an

Table 3. The Effect of Introducing the *yceI* Alleles into S. aureus LAC^{a}

genotype ^b	MIC ($\mu g/mL$)
pLL39	6.2
pLL39_yceI	6.2
pLL39_yce11	50
pLL39_yceI2	50
pLL39_yceI3	50
yceI::Tn pLL39	0.78
yceI::Tn pLL39_yceI	1.6
yceI::Tn pLL39_yceI1	50
yceI::Tn pLL39_yceI2	50
yceI::Tn pLL39_yceI3	50

^{*a*}The MIC values are representative of three independent experiments. ^{*b*}All strains are isogenic and constructed in the LAC background.

isogenic yceI::Tn mutant (JMB9587) (Supplementary Table 7).⁴¹ The strains afforded two functional copies of *yceI* in strain JMB1100 and only one functional copy in the *yceI::Tn* strain. We next quantified the JSF-3151 MIC values for each of these strains (Table 3; Supplementary Figure 6). As expected, the strains containing any of the three *yceI* promoter mutations led to high level resistance to JSF-3151. In addition, the yceI::Tn mutant had a lower MIC than with JMB1100. A quantitative real-time PCR (qPCR) study showed that relative to strain LAC, each of the three mutants had increased transcription of *yceI* (Supplementary Figure 6). Finally, we cloned a PCR fragment containing *yceI* into the pEPSA5.⁴² The pEPSA5_*yceI* vector contained yceI under the transcriptional control of a xylose-inducible promoter to enable gene overexpression. Following overnight induction with xylose, we witnessed a significant resistance to JSF-3151 (>64 fold) as compared with the strain carrying the empty vector (pEPSA5), as well as the no induction controls (Table 4). We hypothesized that

Table 4. The Effect of Overproducing *yceI* Gene Product on JSF-3151 Resistance^a

strain ID	growth with xylose	MIC ($\mu g/mL$)
LAC pEPSA5	+	0.78
LAC pEPSA5_yceI	+	>50
LAC pEPSA5	-	0.78
LAC pEPSA5_yceI	-	3.1
The MIC values are	representative of three	indonandant avnari

^{*a*}The MIC values are representative of three independent experiments.

increased expression of *yceI* would decrease the accumulation of JSF-3151 in cells. To test this hypothesis, IBDM analysis was performed with the *yceI2* strain. Consistent with our hypothesis, treatment of the *yceI2* strain with JSF-3151 resulted in a decreased rate of accumulation when compared with the parent strain. We also noted a decrease in the rate of JSF-3640 accumulation.

A second genetic screen was performed using the LAC *yceI::Tn* strain. Strains with heritable resistance to JSF-3151 arose at a frequency of approximately 3×10^{-8} . Whole-genome sequencing found the four strains had single nucleotide polymorphisms (SNP) in the gene SAUSA300_0206 (*azoR*). The mutations in *azoR* resulted in the following protein changes: Trp₁₀₀stop (*azoR1*), Trp₆₀stop (*azoR2*), Thr₁₂₁Ile (*azoR3*), and Trp₄₀stop (*azoR4*); respectively, they exhibited

MIC shifts of 15-, 32-, 64-, and 25-fold as compared with the *yceI::Tn* parental strain (Table 5). For comparison, a LAC

Table 5. MIC Profiling of JSF-3151 versus Spontaneous Resistant Mutants in the Background of the yceI::Tn Mutant^a

genotype ^b	MIC ($\mu g/mL$)
LAC	1.5
yceI::Tn	0.78
azoR::Tn	50
yceI::Tn azoR1	12
yceI::Tn azoR2	25
yceI::Tn azoR3	>50
yceI::Tn azoR4	25

^{*a*}The MIC values are representative of three independent experiments. ^{*b*}All strains are isogenic and constructed in the LAC background.

strain with a transposon insertion in *azoR* was constructed and demonstrated an increased MIC when compared with LAC (50 versus 1.5 μ g/mL). Moreover, returning *azoR* to the mutant strains via the plasmid pCM28 decreased MIC values to the levels of the *yceI::Tn* strain (Table 6). Overexpression of *azoR* in LAC using the pEPSA5 vector increased sensitivity to JSF-3151 (Table 6). These data verify that the lack of AzoR increases JSF-3151 resistance.

Table 6. The Effect of Overproducing *azoR* Gene Product on Resistance to $JSF-3151^{a}$

genotype	MIC ($\mu g/mL$)
LAC pEPSA5 ^b	1.5
LAC pEPSA5_azoR	0.39
yceI::Tn pCM28 ^c	0.78
yceI::Tn pCM28_azoR	0.78
yceI::Tn azoR1 pCM28	>50
yceI::Tn azoR1 pCM28_azoR	3.1
yceI::Tn azoR2 pCM28	25
yceI::Tn azoR2 pCM28_azoR	1.5
yceI::Tn azoR3 pCM28	>50
yceI::Tn azoR3 pCM28_azoR	1.5
yceI::Tn azoR4 pCM28	25
yceI::Tn azoR4 pCM28_azoR	1.5

^{*a*}The MIC values are representative of three independent experiments. ^{*b*}Strains containing pEPSA vectors were cultured with 1% xylose. ^{*c*}All strains were constructed in the LAC background.

The *azoR* gene is predicted to encode for an intracellular FMN-dependent NADH-azoreductase, catalyzing reductive cleavage of azo groups. Azoreductases from *E. coli*,^{43,44} *P. aeruginosa*,⁴⁵ *B. wakoensis* A01,⁴⁶ and *R. sphaeroides*⁴⁷ have also been reported to reduce nitro groups. We hypothesized that this enzyme is responsible for the biotransformation of JSF-3151 to JSF-3640. To test this hypothesis IBDM analysis was performed with the *azoR::Tn* strain. Interestingly, the accumulation of both JSF-3151 and metabolite JSF-3640 were decreased in the *azoR::Tn* strain as compared with the WT strain (Figure 1, Supplementary Table 8). In comparison, the intrabacterial accumulation of JSF-3151 was further depressed in the *yceI2* mutant (Figure 1, Supplementary Table 8). The relative level of accumulation of metabolite JSF-3640 was similar in the *yceI* and *azoR* mutants. With both

mutants, the intrabacterial levels of both compounds did not reach those found in the WT strain until at least 45 min postaddition of JSF-3151 to the culture.

DISCUSSION

Antibacterial resistance is a serious threat to global public health, and its burden to healthcare increases every year. The diseases caused by *S. aureus* infection are no exception, as they are among the most commonly encountered cases in the clinic today. Growing resistance to the therapeutic options currently available, such as β -lactam resistance and vancomycin resistance, necessitates new antibacterial candidates. Thus, new approaches are required to afford novel drug discovery and tool compounds, especially with time- and cost-efficiency.

Here, our Bayesian modeling approach was shown to be successful in screening millions of compounds and at a fraction of the time traditional high-throughput screening would take. The discovery of JSF-3151 via a machine learning approach validates the use of this strategy for identifying active compounds versus drug-resistant S. aureus, extending efforts from our laboratories with M. tuberculosis^{16–19} and Neisseria gonorrhoeae²⁰ and joins one publication we note that specifically focused on drug-resistant S. aureus.48 This publication from Wang et al. disclosed machine learning models for the growth inhibition of MRSA that correctly predicted (1/56 = 1.8% hit rate) an active (MIC = $4-8 \mu g/$ mL) compound ((*E*)-2-isopropyl-5-styrylbenzene-1,3-diol) versus only MRSA and MSSA strains and did not mention critical downstream studies to assess in vivo efficacy and mechanism of action/resistance. Furthermore, our Bayesian models have significantly expanded beyond methicillin resistance to include clinically relevant data sets with strains resistant to linezolid, vancomycin, or fluoroquinolones.

Our Bayesian models were built with 46 data sets pertaining to assays against drug-resistant *S. aureus*. We utilized these assays in particular because they examined the highest number of compounds against drug-resistant *S. aureus*. These data sets were crucial in developing the "prior distribution," or rather giving statistical weight to the chemical moieties and physiochemical properties relating to activity and inactivity. Our Bayesian models were successful in enriching active and inactive small molecules. Through the use of the MRSA_1b model constructed from the pruned²⁶ training set, we were led to JSF-3151, a small molecule that possesses anti-Staphylococcal activity and a novel structure for an antibacterial.

Thus, we hypothesized that the unique structure of JSF-3151 could imply a new mechanism of antibacterial activity. While nonspecific mechanisms of action were ruled out, the target remains elusive. The in vitro bactericidal activity of JSF-3151 implies that it modulates the activity of one or more targets that alone or together are essential for bacterial growth. Spontaneous-resistant mutants were raised that increased transcription of yceI, encoding a bacterial lipocalin protein that proposedly binds JSF-3151 in analogy to the elegant studies of Valvano and colleagues with Burkholderia cenocepacia.^{37,49} While their work involved complementation of the B. cenocepacia lipocalin knockout with S. aureus yceI, we are unaware of any reports in the literature detailing the role of YceI in S. aureus drug resistance. Our genetic studies have demonstrated that overexpression of yceI leads to loss of JSF-3151 in vitro activity. The analysis of IBDM data with the yceI2 mutant evidenced decreased intrabacterial exposure to both the parent JSF-3151 and identified metabolite JSF-3640 as

compared with the WT strain. Taken together, these findings support the hypothesis that YceI is decreasing the rate of JSF-3151 entry into *S. aureus* cells. Future studies will be required to determine if YceI binds JSF-3151.

Raising JSF-3151 mutants in the *yceI::Tn* background afforded mutations in the gene *azoR*. Strains with null mutations in *azoR* had increased resistance to JSF-3151. We are unaware of previous reports of drug-resistant mutants in *S. aureus* occurring through mutations in an azoreductase despite elegant studies of the role of this protein family in biotransformations of nitroaromatic and azo small molecules.^{43–47,50} It is possible that AzoR is involved in the transformation of JSF-3151 into a yet to be identified compound that has increased antibacterial activity against *S. aureus*. Alternatively, while JSF-3640 did not exhibit antibacterial activity despite its ability to enter cells, its formation from JSF-3151 as promoted by AzoR could be cidal. In support of the hypothesis, the *azoR* mutant had decreased intracellular accumulation of JSF-3640.

The drug resistance and IBDM results are supportive of decreased intrabacterial exposure to JSF-3151 correlating with reduced growth inhibition of *S. aureus*. Overexpression of *yceI* provides depressed accumulation of, and resistance to, JSF-3151. Loss-of-function mutations in *azoR* also decrease intrabacterial levels of JSF-3151 and increase resistance to JSF-3151. However, the degree of biotransformation of JSF-3151 to JSF-3640 as quantified by IBDM in these strains and in the WT does not appear to be linked to growth inhibition. JSF-3640 is a whole-cell inactive metabolite that appears to be formed by a pathway largely independent of the activity of AzoR. The AzoR-dependent biotransformation product of JSF-3151 has yet to be identified and may be linked to the mechanism of action of JSF-3151.

It is noteworthy that reports of *S. aureus* metabolomics appear to primarily focus on the effect of antibacterials on bacterial metabolite pools^{50–55} and not on intracellular drug metabolism. *In vitro* enzyme assays have been leveraged as a major approach to study enzyme-mediated small-molecule modifications^{56,57} but are potentially limited by a lack of physiological relevance to intrabacterial metabolism. We are aware of one elegant report that utilized LC-MS to quantify drug accumulation in *S. aureus*, but it is noteworthy that drug metabolism was not considered.⁵⁸ Increased emphasis on the application of IBDM to *S. aureus* programs should further enlighten our understanding of bacterial metabolic pathways, and especially those engaged in xenobiotic biotransformations, while also providing information critical to the design of more potent analogues of JSF-3151 and, thus, furthering the impact of antibacterial machine learning models.

CONCLUSIONS

The results presented herein disclose an antibacterial agent JSF-3151 with both *in vitro* and *in vivo* efficacy against clinically relevant drug-resistant *S. aureus* strains. This molecule was discovered by the successful application of a Bayesian modeling approach. The mechanism of resistance to JSF-3151 has been shown to involve (1) overexpression of the lipocalin-like protein YceI or (2) the loss of function of the azoreductase AzoR. To the best of our knowledge, neither of these resistance mechanisms has been described for staph-ylococci and both are supported by extensive genetic analyses. Furthermore, we have demonstrated how each type of mutation perturbs accumulation of JSF-3151 and its identified

metabolite JSF-3640 within the bacterium. The chemical entity, knowledge of resistance mechanisms, and platforms of Bayesian modeling and intrabacterial drug metabolism should inform further efforts to impact next-generation therapeutic strategies for drug-resistant *S. aureus* infections.

METHODS

In the detailed methods below, no unexpected or unusually high safety hazards were encountered.

Materials and Services. Quick ligase, Phusion polymerase, and restriction enzymes were purchased from New England Biolabs. Gel purification kits were purchased from Qiagen. Tryptic Soy broth (TSB) was purchased from MP Biomedicals. Mueller—Hinton broth and agar were purchased from Fisher Scientific. Whole-genome sequencing was conducted by the Genomics Center, Rutgers University-New Jersey Medical School, while Sanger sequencing was performed at Genewiz, South Plainfield, NJ.

Mice. All animal experiments were approved by the Rutgers Institutional Animal Care and Use Committee. All experiments used 6-week old female outbred CD-1 or Swiss Webster mice (Charles River Laboratories, Wilmington, MA). Mice were housed in filter-top cages and maintained in accordance with American Association for Accreditation of Laboratory Care criteria.

Bacterial Strains, Culture Conditions, Primers, and Plasmids. MRSA (43300) and MSSA (25923) strains used in this study were purchased from ATCC (Manassas, VA). VRSA and VISA strains were obtained from the Kreiswirth laboratory at the Public Health Research Institute Center (Newark, NJ). Other S. aureus strains used in this study were derivatives of the S. aureus strain LAC, which is a community acquired MRSA strain. JMB1100 is a derivative of LAC that has been cured of the pUSA03, which confers multidrug resistance.⁵⁹ S. aureus were routinely cultured in solid or liquid tryptic soy broth. For experiments, S. aureus were cultured in 5 mL of Mueller Hinton and grown at 37 °C in 30 mL capacity culture tubes with shaking at 200 rpm. Overnight cultures were diluted 1/1000 and transferred into 96-well plates containing serial dilutions of JSF-3151 in Mueller-Hinton broth. Final concentrations were calculated based on 200 μ L final volume. Plates were wrapped with parafilm from the sides to prevent dehydration and incubated statically at 37 °C. Results were measured after 18 h of incubation. When selecting for plasmids, transposon, or episome insertions, antibiotics were added at the final following concentrations: 150 μ g/mL ampicillin; 30 µg/mL chloramphenicol (Cm); 10 µg/mL erythromycin (Erm); 3 μ g/mL tetracycline (Tet). For routine plasmid maintenance, media were supplemented with 10 μ g/ mL or 3.3 μ g/mL of chloramphenicol.

Data Collection and Curation. The PubChem Bioassay Database was queried with the phrase "aureus and MIC", which yielded 22 619 for results. After sorting the largest sets with *S. aureus* results, the strain in each assay was identified and sorted into "drug-resistant," "drug-sensitive," and "unknown." In total, 139 assay results were collected. For each of these assays, the structural data files (SDF) and comma-separated value (CSV) files were downloaded. The SDF's contained every tested compound's PubChem CID and 2D structural information (but no assay results). The CSV files contained every tested compound's CID and assay results (but no structures). Additionally, for every AID, the corresponding paper was retrieved. Every assay result from a CSV file was inspected and compared to its relevant paper. Using this quality control process, any erroneous data could be noted or discarded from our study. Following the comprehensive manual inspection process, we found 72 data sets pertaining to drug-sensitive S. aureus, 46 data sets pertaining to drugresistant S. aureus, and 21 data sets that were classified as unknown. The drug-resistant sets contained assay results primarily for MRSA but also included assay results against fluoroquinolone-resistant or linezolid-resistant S. aureus, as well. Unfortunately, one of the CSV files from the drugresistant set, AID 548647, had erroneous data and was, therefore, discarded. The data within the CSV files and SDF's for each of the drug-resistant sets were collated into a spreadsheet in Discovery Studio 4.0 (BIOVIA, Inc., San Diego, CA). For consistency, the μ M sets were converted to μ g/mL. We then deleted duplicate compounds using a custom script in Pipeline Pilot 9.1 (BIOVIA, Inc., San Diego, CA) and compounds with a molecular weight >850 g/mol.

Training and Test Sets. In total, our full training set from PubChem contained 1633 compounds, of which 1043 compounds (63.8%) were designated as active (MIC \leq 10 μ g/mL). This full training set (MRSA 1a) encompassed a small yet fairly diverse portion of chemical space, which included known antibacterial scaffolds such as tetracyclines, fluoroquinolones, and β -lactams. However, it also included rhodanines (which are known to be promiscuous Pan Assay Interference compounds, called "PAINS"^{23,60}). We hypothesized that removing (or pruning) these problematic compounds as well as known antibacterial chemotypes from the training set would help teach the Bayesian to further focus on new chemical scaffolds, which should increase the likelihood that they display novel mechanisms of action.²⁶ It is important to note, however, that our M. tuberculosis Bayesian models have routinely found hits which differ significantly from the model testing set (pairwise Tanimoto similarity <0.7). Utilizing this strategy, we manually pruned the tetracyclines, fluoroquinolones, rhodanines, and β -lactams from the actives subset of MRSA 1a to produce a new training set (MRSA 1b) that contained 1247 compounds, of which 657 compounds were classified as active (52.7%).

Another training set that we used as a reference was based on the Broad Institute's assay results against methicillinsensitive *S. aureus* (Broad_MSSA). This training set has 10 934 compounds, of which 193 compounds (1.77%) were defined as "active" according to the Z-factor threshold selected by the researchers at the Broad Institute (Tali Mazor, assay names: ChemBank BacTiterGlo 1064.0004 and 1064.0005; http:// chembank.broadinstitute.org). These 193 actives also included known antibacterials, such as levofloxacin, gatifloxacin, β lactams, tetracyclines, and fluoroquinolones.

These data sets were used as independent training sets to create two different, new drug-resistant *S. aureus* machine learning models and the reference MSSA model in Pipeline Pilot 9.1. These Bayesian models utilized nine different descriptors: AlogP, molecular weight, number of rings, number of aromatic rings, number of rotatable bonds, number of hydrogen bond donors, number of hydrogen bond acceptors, molecular fractional polar surface area, and molecular function class fingerprints of maximum diameter 6 (FCFP_6, characterizing the 2D substructures up to and including 6 rings/zones/ dimensions of topology).^{61,62} Together, these descriptors define the physiochemical properties of each compound as a whole and the 2D substructures of different regions within it. These descriptors were then combined in different ways, with different weights applied to the sets of different descriptor combinations, until the most accurate model was produced, according to 5-fold internal cross-validation.

The top 100 scoring compounds from MRSA 1a and the top 50 from MRSA 1b versus an Enamine (www.enamine.net) library of 3 million compounds, accessed through Zinc (http://zinc.docking.org), were then visually inspected to identify and discard any reactive compounds or promiscuous inhibitors (e.g., Michael acceptors, rhodanines, etc.). Also removed were compounds that were visually deemed too similar to the training set to be novel (e.g., tetracyclines or almost all fluoroquinolones), or compounds with chemotypes that were not of general interest (i.e., peptides or fatty acids).^{23,60} 49 of the top 50 compounds from MRSA_1a did not pass inspection. Therefore, we also inspected the next 50 top scoring compounds from this model. Unsurprisingly, the highest-ranking compound from MRSA 1a was a fluoroquinolone which, instead of dismissing, we included in this model's predictions to be experimentally validated. We selected an additional top 9 compounds from MRSA 1a, as well as the top 5 ranking compounds from MRSA 1b that passed inspection. All candidate compounds were purchased and assessed by LC/MS for \geq 95% purity and the expected parent ion in the mass spectrum.

Drug Susceptibility Assays. MICs were performed in 96well microtiter plates. A DMSO stock solution of each compound was added to the first column and serially diluted across the columns of the plate. The last column of the plate contained no drug and served as a no-drug control. Overnight cultures of the bacterium being tested (MRSA ATCC 43300, MSSA ATCC 25923, or the aforementioned VRSA/VISA strains) were diluted 1000 fold (2 × 10³ cells), and 100 μ L was used as an inoculum in each well. MICs were determined by visual inspection for a pellet after 18 h incubation at 37 °C. Compounds were tested for cytotoxicity against Vero cells using the CellTiter 96 AQueous One Solution kit (Promega). Vero cells were seeded in 96-well plates at a density of 2×10^4 cells per well, and the plates were incubated for 4 h at 37 °C to allow attachments of the Vero cells. Compounds were then added to the wells starting from a final concentration of 50 μ g/ mL and making 12 1:2 dilutions. Cells were incubated for 72 h at 37 °C. Then 20 μ L of freshly prepared MTS:PMS reagents was added to each well. The plates were incubated for 2 h and then read at an absorbance of 490 nm.

Mouse Pharmacokinetic Study. Animals and ethics assurance: Animal studies were carried out in accordance with the guide for the care and use of Laboratory Animals of the National Institutes of Health, with approval from the Institutional Animal Care and Use Committee (IACUC) of the New Jersey Medical School, Rutgers University, Newark. All animals were maintained under specific pathogen-free conditions and fed water and chow ad libitum, and all efforts were made to minimize suffering or discomfort. Two female CD-1 mice received a single dose of experimental compound administered orally at 25 mg/kg in 20% DMA/80% PEG300, and blood samples were collected in K2EDTA coated tubes predose, 0.5, 1, 3, and 5 h postdose. Blood was kept on ice and centrifuged to recover plasma, which was stored at -80 °C until analyzed by HPLC coupled to tandem mass spectrometry (LC-MS/MS).

LC/MS-MS analytical methods: LC/MS-MS quantitative analysis for all molecules was performed on a Sciex Applied

Biosystems Qtrap 4000 triple-quadrupole mass spectrometer coupled to an Agilent 1260 HPLC system, and chromatography was performed on an Agilent Zorbax SB-C8 column (2.1 \times 30 mm; particle size, 3.5 μ m) using a reverse phase gradient elution. Milli-Q deionized water with 0.1% formic acid (A) was used for the aqueous mobile phase and 0.1% formic acid in acetonitrile (B) for the organic mobile phase. The gradient was: 5-90% B over 2 min, 1 min at 90% B, followed by an immediate drop to 5% B and 1 min at 5% B. Multiple-reaction monitoring of parent/daughter transitions in electrospray positive-ionization mode was used to quantify all molecules. Sample analysis was accepted if the concentrations of the quality control samples and standards were within 20% of the nominal concentration. Data processing was performed using Analyst software (version 1.6.2; Applied Biosystems Sciex). Neat 1 mg/mL DMSO stocks for all compounds were first serial diluted in 50/50 acetonitrile/water and subsequently serial diluted in drug free CD-1 mouse plasma (K2EDTA, Bioreclamation IVT, NY) to create standard curves (linear regression with $1/x^2$ weighting) and quality control (QC) spiking solutions. Twenty μ L of standards, QCs, control plasma, and study samples were extracted by adding 200 μ L of acetonitrile/methanol 50/50 protein precipitation solvent containing the internal standard (10 ng/mL verapamil). Extracts were vortexed for 5 min and centrifuged at 4000 rpm for 5 min. 100 μ L of supernatant was transferred for HPLC-MS/MS analysis and diluted with 100 μ L of Milli-Q deionized water.

Mouse Peritonitis-Sepsis Model. MRSA COL strain was grown overnight in Mueller-Hinton media at 37 °C. The resulting culture was diluted in 5% hog mucin and 0.9% NaCl to a challenge inoculum of approximately 1.0×10^8 CFU per mouse. The challenge inoculum was confirmed by mannitol salt agar plate spreads and administered to female outbred Swiss Webster mice (Charles River Laboratories, Wilmington, MA) via intraperitoneal injection (0.5 mL). Groups consisting of 12 mice were given single doses of vehicle (10% DMSO in sterile saline), JSF-3151 (24 mg/kg) or vancomycin (100 mg/ kg) 1 h postinfection. Vehicle and JSF-3151 were administered via intraperitoneal injection while vancomycin was administered by subcutaneous injection (final injection volume = 0.2mL). Mice were maintained in accordance with the American Association for Accreditation of the Laboratory Care criteria. All surviving animals were humanely euthanized at the end of the study (3 d) by following the American Veterinary Medical Association Guidelines for the Euthanasia of Animals. The animal study was approved by Rutgers Institutional Animal Care and Use Committee.

IBDM Studies. A single colony of *S. aureus* strain was inoculated in fresh Mueller–Hinton media and grown overnight at 37 °C. The resulting overnight was diluted 1:100 in fresh Mueller–Hinton media, and 120 mL of the culture was grown in a 250 mL flask until OD_{600} was ~0.5. Then, JSF-3151 or JSF-3640 was added at the appropriate concentration and 10 mL of the culture were sampled at the relevant time intervals post-treatment. Each aliquot was centrifuged. The supernatant fraction was immediately stored while the pellet fraction was washed once with PBS and then resuspended in ice cold 2:2:1 acetonitrile:methanol:water to halt bacterial metabolism, and immediately placed on dry ice. The cells in the pellet fraction were subjected to bead beating and all of the fractions (6 pellet, 6 supernatant) were filtered (0.22 μ m membrane) and analyzed by LC-MS. LC-MS was

performed on an Agilent 1260 HPLC coupled to an Agilent 6120 MS. The run condition for the LC-MS was 10-100% acetonitrile in water over 10 min (both acetonitrile and water contained 0.1% formic acid). The column used for the run was an EMD Millipore Chromolith Speedrod RP18-e, 50×4.6 mm.

Membrane Depolarization Assay. A single colony of S. aureus (ATCC 43300) was inoculated in fresh Mueller-Hinton media and grown to an OD_{600} of 0.5. DiBAC₄ (bis(1,3dibutylbarbituric acid)trimethine oxonol) was added to the resulting culture to afford a final concentration of 20 μ g/mL and incubated for 5 min. All reads were performed in triplicate on a Biotek Synergy Neo 2 plate reader with excitation and emission wavelengths set to 490 and 510 nm, respectively. In a Costar 96-well, black-side, clear-bottom plate, 200 µL of bacteria and dye was added and kept as a negative control. At a separate location of the plate, 200 μ L of bacteria and dye was added along with 4.5× MIC of JSF-3151 (~18 µg/mL, 54 μ M). At yet another location, 200 μ L of bacteria and dye was added along with sodium deoxycholate, the positive control, to a final concentration of 0.05% w/v. Reads were performed for 10 min with 5 s intervals.

Nitric Oxide Release Assays. (A) DAF-FM: A single colony of S. aureus (ATCC 43300) was inoculated in fresh Mueller-Hinton media to an OD₆₀₀ of 0.2. DAF-FM diacetate (4-amino-5-methylamino-2',7'-difluorofluorescein diacetate) was added to a final concentration of 10 μ M and allowed to incubate for 30 min in the dark. All reads were performed in triplicate on a Biotek Synergy Neo 2 plate reader with excitation and emission wavelengths set to 495 and 515 nm, respectively. In the first row of a Costar 96-well, black-side, clear-bottom plate, 200 µL of Mueller-Hinton media alone was added and kept as a negative control. In the second row, 200 μ L of bacteria alone were added and also kept as a negative control. In the third row, bacteria treated with DAF-FM were added. In the fourth row, bacteria treated with DAF-FM and JSF-3151 (~33 μ g/mL, 100 μ M) for 120 min were added. In the fifth row, bacteria treated with DAF-FM and 100 μ M diethylamine NONOate diethylammonium salt (DEA-NONOate) for 120 min were added. (B) Griess: Sodium nitrite stock solutions ranging from 50–0.78 μ M (2-fold dilutions) were prepared in Mueller-Hinton media. A single colony of S. aureus (ATCC 43300) was inoculated in fresh Mueller-Hinton media and grown to an OD_{600} of 0.4. The resulting culture was split into three equal volume batches. The first batch served as the negative control, the second volume was treated with JSF-3151 (~33 μ g/mL, 100 μ M), and the third volume was treated with DEA-NONOate (100 μ M). Following incubations of 24 and 48 h, 1 mL of culture was sampled and centrifuged. The supernatant was placed in the same plate as mentioned above along with the nitrite standards (180 μ L per well). To each of these wells, 20 μ L of freshly prepared Griess reagent mix was added, and the plate was allowed to sit at room temperature for 30 min. The plate was read with the above reader at an absorbance of 548 nm. Reads from the sodium nitrite standards were used to correlate absorbance to concentration, which facilitated estimation of nitrite production.

Molecular Biology. Escherichia coli DH5 α was used as a cloning host throughout. Plasmids were passaged through RN4220 and subsequently transduced into the appropriate strains using bacteriophage 80α .^{63,64} The amplicons used to create the pll39_yceI vectors were: 2620upBamHI and

2620downSalI. The pLL39 vector was linearized with Sall and BamHI and combined with similarly digested PCR fragments. The pLL39 yceI constructs were transformed into RN4220 containing pLL2787 and integrated onto the chromosome at the L54a *attB* site as previously described.³⁹ Episome integration was verified using the Scv8 and Scv9 primers.³⁹ The integrated pLL39 yceI constructs were transduced individually from RN4220 into JMB1100. The yceI::Tn and azoR::Tn alleles were acquired from the NARSA collection⁴¹ and were transduced individually into JMB1100 before further strain construction. LAC genomic DNA was used as a template and the following primer pairs were used to create amplicons for cloning: 2620uppEPBamHI and 2620downSalI; pEPSA5-0206 EcoRI for and pEPSA5-0206 BamHIrev; pCM28-0206 BamHI rev and pCM28-0206 PstI for (S6 Table). The pEPSA5 and pCM28 vectors were linearized with restriction enzymes and combined with similarly digested PCR fragments before ligation. Sanger sequencing was used to confirm vectors (Genewiz; South Plainfield, NJ). Single nucleotide polymorphisms were mapped as previously described.⁶⁵

Total RNA Extraction and Quantitative Real-Time PCR (qPCR). The *S. aureus* strains, wild type and three JSF-3151 spontaneous-resistant mutants, were grown to mid log phase ($OD_{595} = 0.4$). Bacteria were harvested and the pellets were lysed, post TRIzol (Invitrogen) addition, by bead beating. The total RNAs were extracted in chloroform and purified via RNeasy mini kit (Qiagen). The cDNA library was constructed via reverse transcription using 2 μ g total RNA as template with SuperScript III First-Strand Synthesis System (Invitrogen). The expressions of *yceI* and 16S rRNA were measured by qPCR via SuperScript III Platinum SYBR Green One-Step qPCR Kit with Rox (Invitrogen).

Compound Synthesis. Reagents and solvents for synthesis were purchased from Sigma-Aldrich, Fisher Scientific, or Enamine Ltd., and used without further purification unless noted otherwise. Analytical TLC was performed with Merck silica gel 60 F_{254} plates. Silica gel column chromatography was conducted with Teledyne Isco CombiFlash Companion or Rf+ systems. ¹H and ¹³C NMR spectra were acquired on a Bruker Avance III HD 500 MHz instrument and are listed in parts per million downfield from tetramethylsilane. LC-MS was performed on an Agilent 1260 HPLC coupled to an Agilent 6120 MS. All synthesized compounds were at least 95% pure as judged by their HPLC trace at 220 or 250 nm and were characterized by the expected parent ion(s) in the MS trace. HRMS data were acquired on an Agilent 6220 Accurate-Mass Time-of-Flight mass spectrometer.

General Procedure A: Synthesis of JSF-3151.



JSF-3151 was prepared in two steps. In short, a round-bottom flask was charged with 5-nitrothiophene-2-carboxylic acid (1.00 g, 5.78 mmol). Thionyl chloride (10 mL) was then

added with stirring. After 3 h, the product A was concentrated under reduced pressure to afford a light yellow solid (88% yield). 6-chloroquinolin-8-amine (1.03 g, 5.78 mmol) was added in a round-bottom flask followed by the addition of 15 mL dichloromethane. A (0.971 g, 5.09 mmol) was then added followed by diisopropylethylamine (1.20 equiv, 6.94 mmol, 1.21 mL). The mixture was allowed to stir at room temperature for about 6 h. The reaction mixture was washed twice with water and saturated aqueous sodium bicarbonate solution and once with saturated aqueous brine solution. The organic layer was dried over anhydrous sodium sulfate and then filtered through a pad of Celite. The filtrate was concentrated in vacuo and the product was purified by flash column chromatography (10% methanol:dichloromethane) to afford the product as a yellow solid (1.67 g, 87% yield): ¹H NMR ((CD₃)₂SO, 500 MHz) δ 10.9 (s, 1H), 9.06–9.00 (m, 1H), 8.51 (s, 1H), 8.47 (d, J = 8.2 Hz, 1H), 8.24 (d, J = 2.8 Hz, 1H), 8.16 (d, J = 3.0 Hz, 1H), 7.99 (s, 1H), 7.79–7.73 (m, 1H). Also noted 5.8 (s, DCM), 1.2 (m). ¹³C NMR ((CD₃)₂SO, 126 MHz) δ 158.8, 154.3, 150.4, 145.2, 137.9, 136.6, 135.1, 131.4, 130.7, 129.4, 129.1, 124.0, 122.8, 119.5. HRMS (ESI): Calculated for $C_{14}H_8ClN_3O_3S$ (M+H)⁺ = 334.0053; Observed 334.0069 (+4.8 ppm).

Compound 3 – JSF-3587.



JSF-3587 was synthesized following **General Procedure A** on a 0.3 mmol scale. Following flash chromatography, the product was isolated as a white solid (82 mg, 95% yield): ¹H NMR ((CD₃)₂SO, 500 MHz) δ 10.6 (s, 1H), 9.01 (dd, *J* = 4.2, 1.5 Hz, 1H), 8.61 (d, *J* = 2.3 Hz, 1H), 8.45 (dd, *J* = 8.3, 1.6 Hz, 1H), 8.03 (d, 1H), 7.99 (d, *J* = 5.0 Hz, 1H), 7.90 (d, *J* = 2.3 Hz, 1H), 7.75 (dd, *J* = 8.3, 4.2 Hz, 1H), 7.31 (t, *J* = 4.7, 3.9 Hz, 1H). Also noted 5.7 (s, DCM), 3.3 (s, H₂O). LRMS (ESI): Calculated for C₁₄H₉ClN₂OS (M+H)⁺ = 289.0; Observed 289.0.

General Procedure B: Synthesis of JSF-3640.



JSF-3151 (10 mg, 0.030 mmol) was added to a 20 mL scintillation vial. This was followed by the addition of Zn powder (9.7 mg, 0.15 mmol), ammonium formate (9.5 mg, 0.15 mmol), and 5 mL of methanol. The reaction was allowed to stir for 1 h at room temperature. Workup and purification was performed as outlined by **General Procedure A**. The reaction mixture was washed twice with water and saturated aqueous sodium bicarbonate solution and once with saturated aqueous brine solution. The organic layer was dried over anhydrous sodium sulfate and then filtered through a pad of Celite. The filtrate was concentrated *in vacuo* and the product was purified by flash column chromatography (10% methanol:dichloromethane) to afford the product.

JSF-3640 was synthesized following General Procedure B. Following flash chromatography, the product was isolated as an off-white solid (8.1 mg, 88% yield): ¹H NMR ($(CD_3)_2SO$,

500 MHz) δ 10.1 (s, 1H), 8.96 (d, J = 4.1 Hz, 1H), 8.59 (d, J = 2.1 Hz, 1H), 8.41 (d, J = 8.3 Hz, 1H), 7.78 (d, J = 2.1 Hz, 1H), 7.72 (dd, J = 8.3, 4.2 Hz, 1H), 7.53 (d, J = 4.1 Hz, 1H), 6.75 (s, 2H), 5.99 (d, J = 4.1 Hz, 1H). Also noted 5.7 (s, DCM), 3.3 (s, H₂O), 1.2 (s). HRMS (ESI): Calculated for C₁₄H₁₀ClN₃OS (M+H)⁺ = 304.0311; Observed 304.0304 (-2.3 ppm).

ASSOCIATED CONTENT

Supporting Information

The Supporting Information is available free of charge at https://pubs.acs.org/doi/10.1021/acsinfecdis.1c00265.

Comparative model external ROC curves; *In vitro* bactericidal activity of JSF-3151; *In vivo* profiling of JSF-3151; IBDM studies of JSF-3151 and its metabolite JSF-3640; JSF-3151 membrane depolarization and NO[•] release assays; Quantification of *yceI* transcription in JSF-3151-resistant strains; Internal Statistics for Bayesian Models; External Statistics for Bayesian Models; External Statistics for bayesian Models; *S. aureus* MIC values for top-scoring compounds from MRSA_1a and MRSA_1b predictions with the Enamine library; MRSA MIC values for bottom-scoring compounds from MRSA_1b predictions with the Enamine library; MIC of JSF-3151 versus VRSA and VISA strains; Primers used in this study; Strains and plasmids used in this study; Accumulation metrics for JSF-3151 and JSF-3640 in select strains (PDF)

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Author Contributions

J.S.F. conceived of the project. J.S.P., A.L.P., and SE conducted the Bayesian modeling. J.S.P., R.R., and J.O. carried out the *in vitro* biological studies which N.C. helped analyze. J.S.P. and S.S.D. executed the synthetic chemistry. J.S.P., J.N., and Y.-M.A. performed the IBDM studies. J.S.P, H.A.-T., and J.M.B. and J.N. designed, conducted, and described the drug resistance and genetic studies. M.D.Z. and H.-P.H. were responsible for the mouse pharmacokinetic studies which were analyzed by V.D.. S.P. conducted the mouse infection model studies and the resulting data were analyzed by S.P., D.S.P., J.S.P., X.W., J.S.F., J.M.B., and J.N. analyzed the generated data and wrote the manuscript. All authors contributed to editing the manuscript.

Notes

The authors declare the following competing financial interest(s): S.E. is the CEO of Collaborations Pharmaceuticals that is exploring the use of machine learning techniques to discover, optimize, and commercialize novel anti-infective agents.

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ABBREVIATIONS

IBDM, intrabacterial drug metabolism; MRSA, methicillinresistant S. aureus; VISA, vancomycin-intermediate-resistant S. aureus; VRSA, vancomycin-resistant S. aureus; MIC, minimum concentration of compound to inhibit bacterial growth in vitro by 90%; PAINS, Pan Assay Interference compounds; FCFP6, functional class fingerprints of maximum diameter 6; ROC, Receiver Operator Characteristic; MSSA, methicillin-sensitive S. aureus; CC_{50} , minimum concentration of compound resulting in 50% growth inhibition of this model mammalian cell line; MBC, minimum bactericidal concentration; CFU, colony-forming units; PK, pharmacokinetic; LC-MS, liquid chromatography-mass spectrometry; HRMS, high-resolution mass spectrometry; DiBAC₄, bis(1,3-dibutylbarbituric acid)trimethine oxonol; DAF-FM, 4-amino-5-methylamino-2',7'difluorofluorescein; qPCR, quantitative real-time PCR; SNP, single nucleotide polymorphism/s.

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