The β-lactam Ticarcillin *is a Staphylococcus aureus* UDP-N-acetylglucosamine 2-epimerase binder.

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Abstract

Infectious diseases account for 25% of the causes of death worldwide and this rate is expected to increase with the increasing rate of antibiotic resistance observed for many pathogens. Among the bacterial pathogens usually found in healthcare associated infections, *Escherichia coli* and *Staphylococcus aureus* are the most prevalent pathogens and, for the former, about 50% of the isolates are found to be methicillin resistant. Given the limited number of targets/pathways observed for the mechanism of action for the existing antibiotics, the discovery of newer targets and their evaluation becomes an urgent and necessary task. Here we describe the structure-based identification of ticarcillin as a weak binder of the UDP-N-acetylglucosamine 2-epimerase (MnaA) from *S. aureus*. After a docking-based screening of existing drugs, ticarcillin was identified as a ligand in isothermal analysis of differential scanning fluorimetry data. An equilibrium molecular dynamics simulation confirmed the docking binding mode as a stable pose, with large contributions to the binding energy coming from interactions between Arg206 and Arg207 and the carboxylate groups in ticarcillin.

1. Introduction

The Gram-positive bacteria *Staphylococcus aureus* is listed by the Center of Disease Control (CDC) at USA among those bacteria that represent a serious threat to humans due to their high rates of observed antibiotic resistance [1]. The number of infections caused by methicillin-resistant *S. aureus* (MRSA) strains per year reaches 80,000, resulting in a about 11,000 deaths [1]. Also, according to the survey taken by the National Healthcare Safety Network at the Centers for Disease Control and Prevention, USA, *S. aureus* was ranked as the second most frequent organism causing healthcare-associated infection, after *Escherichia coli*, and was responsible for about 12% of the reported infections [2].

The same report also brings an interesting, though worrying, data. According to the data collected from 2011-2014, about 50-52% of the *S. aureus* clinical isolates from central line-associate bloodstream infections were reported to be resistant to methicillin. From isolates from catheter-associated urinary tract infections, about 52-57% were reported to be methicillin-resistant. From isolates from ventilator-associated pneumonia and from surgical site infections, 42-44% were reported to be resistant [2]. The data from 2009-2010 also indicates that the rates in resistance for *S. aureus* follows a worrisome trend, with about 50% of the Staphylococcal healthcare-associated infections being caused by a resistant strain [3].

The estimated number of people with important infections caused by antibiotic resistant bacteria reached 2,000,000 in 2013, with 23,000 deaths directly caused by these infections [1]. In Europe, about 25,000 deaths were related to antibiotic resistant bacteria in 2007 by the European Center of Disease Prevention and Control [4]. The estimated number of deaths per year caused by antibiotic resistance globally reaches 700,000, highlighting the

burden caused by resistant bacteria [5]. In addition, Thorpe and coworkers estimated that

antibiotic resistance results in additional $1,383 to the cost of the treatment per patient,

resulting in an annual cost of 2.2 billion dollars in the US [6].

The bacterial cell envelope, first line of defense for bacteria against changes in the

environment, is also the target of the first-choice antibiotic for treating many *S. aureus*

infections. The beta-lactam antibiotics act by binding and inhibiting the penicillin-binding

proteins (PBPs) [7], impairing the transpeptidase activity and resulting in the inhibition of the

peptidoglycan synthesis. The cell envelope in Gram-positive bacteria such as *S. aureus*, is also

decorated with the wall teichoic acid (WTA) [8], which are polymers containing 40-60 polyol

repeats and can account for up to 60% of the cell wall mass [8]. The association between beta-

lactam antibiotic resistance and the composition of the WTA has been shown, although the

precise mechanisms underlying this association are not completely clear yet. Brown and

coworkers showed that the deletion of the glycosyltransferase (GT) TarS, responsible for the

decoration of the WTA with β-O-GlcNAc, can re-sensitize MRSA strains to beta-lactam

antibiotics [9]. Additionally, it has been shown that some PBP can interact with the WTA [10],

again suggesting some cross-talk between the WTA and the mechanism of action of the beta-

lactam antibiotics.

WTA itself is a very interesting target for the development of new antibiotic

candidates. Tarocins and tunicamycin, for example, were shown to act on the synthesis of the

WTA by inhibiting the first enzyme of the biosynthesis pathways, TarO [11], with bactericidal

effects on resistant bacteria and also with synergistic effects with beta-lactam antibiotics.

Tunicamycin, not only inhibits TarO but also inhibits MnaA, an UDP-N-acetylglucosamine 2-

epimerase that provides the appropriate concentrations of UDP-GlcNAc and UDP-ManNAc

for the further steps of the biosynthetic pathway [12,13].

As highlighted by Walsh and Wencewicz, the antibiotics under use currently are

limited to only five major validated targets/pathways [14]. So, given the scarcity of targets

and the relevance of the cell wall decoration for *S. aureus*, we report here the structure-based

screening for new MnaA binders using a catalogue of existing drugs in a repurposing strategy

[15]. After molecular docking and careful visual inspection, we identified fifteen ligand

candidates. Further experimental evaluation with quantitative DSF (qDSF) revealed ticarcillin,

a β-lactam, as a weak binder for *S. aureus* MnaA (SaMnaA). An equilibrium molecular

dynamics simulation confirmed a stable binding pose over 600 ns with strong interactions

between ticarcillin carboxylate groups and SaMnaA R206 and R207 residues.

2. Materials and Methods

2.1 Molecular Docking and Ligand Selection

The molecular model used for docking calculations was based on the crystal structure

for *S. aureus* MnaA, PDB Id 5ENZ [13], with the same structure preparation procedure as

previously described [16]. Briefly, the server H^{++} [17] was used to define the protonation

states of residues and hydrogen atoms were added accordingly. The atom types and charges

were defined according to AMBER FF14SB [18]. In this model, an UDP molecule was

considered as a part of the 'receptor'. The atomic charges were computed using a RESP fitting

procedure after initial calculations using Hartree-Fock method with 6-31G* basis, as available

in Gaussian 09 [19]. The substrate UDP-GlcNAc was prepared using the same method and

considered as the 'reference ligand' during the docking calculations. The ZINC Drug Database

(ZDD) was used for the screening. The MOL2 files were used as provided by the database

using the default ZINC methods for ligand preparation [20]. The docking calculations were set

up using `LiBELa`, a ligand- and receptor-based docking software developed in our group

[21], using a pre-computed `DelPhi` [22,23] electrostatic potential for modeling the polar

interactions and a 6-12 Lennard Jones potential to model van der Waals interactions [24]. The

docking results were analyzed using the ViewDock tool available in UCSF Chimera [25] and

promising candidates were selected based on the visual inspection of favorable interactions

involving the ligand and the enzyme.

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111   2.2 SaMnaA expression and purification
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S. aureus MnaA was cloned from a S. aureus SA16 gDNA sample kindly provided by Prof.

Dr. Ilana Camargo. The SaMnaA gene was amplified using the primers forward 5'-

CAGGGCGCCATGATGAAAAAGATTATGACCATATTTG-3' and reverse 5'-GACCCGACGCGGTTA

TCAGAAATCGCTTGGTTTT-3'. The amplified gene was inserted in the pET-Trx1A vector using

a strategy described elsewhere [26]. *E. coli* Rosetta (DE3) cells were transformed with the

resulting plasmid and used for protein expression. A starting culture was grown overnight in

LB medium supplemented with containing 50 μg/ml kanamycin. Afterward, the starting

culture was used to inoculate LB medium containing 50 μg/ml kanamycin. The cells were

incubated in a shaker at 18°C and, after reaching an Optical density of 0.8 at 600 nm, the

protein expression was induced with 0.1 mM IPTG. After 16 hours at 18°C, the cells were

harvested by centrifugation at 7800 x g for 30 min and disrupted by sonication in lysis buffer

containing 50 mM NaH_2PO_4, pH 8.0, 300 mM NaCl, 10 mM imidazole, 1 mM PMSF, 1 mg/mL

lysozyme. The soluble proteins were isolated from the cell debris by centrifugation at 18,000

x g for 30 min at 4°C and SaMnaA was purified by IMAC. Briefly, the soluble fraction was

incubated with 5 ml of the Ni resin equilibrated with washing buffer containing 50 mM

NaH_2PO_4, pH 8.0, 300 mM NaCl, 20 mM imidazole, 10% glycerol. After extensive washing,

MnaA was eluted in the same buffer supplemented with 500 mM imidazole. After dialysis to

remove the imidazole, and concentration up to about 2 mg/ml and incubated with a Tobacco

Etch Virus (TEV) protease for 16 hours at 4°C, to remove the fusion protein. A second IMAC

step was carried out to remove the His-tagged fusion protein thioredoxin. For crystallization

trials, an additional purification step was carried out with size exclusion chromatography

using a Superdex 75 column.

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136 2.3 Differential Scanning Fluorimetry (DSF)
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Fourteen ligands were initially selected from docking calculations and were screened for

their ability to bind to *S. aureus* MnaA *in vitro* using a DSF assay. Briefly, SaMnaA at 2μM was

incubated with each ligand at a concentration of 6 μM (3:1 ratio) for 30 minutes. Afterwards,

the complex was incubated with SYPRO Orange dye (ThermoFisher) for DSF experiment using

a Bio-Rad CFX96 real-time PCR Detection System equipment using a temperature range of 25-

95 °C with a heating rate of 1°C every 30 seconds. The obtained data was adjusted according

to the unfolding model proposed by Bai and coworkers [27], that relates the measured

fluorescence intensity *Y* to the temperature *T* [27]:

$$Y(T) = \left[\frac{1}{1+K_U}(m_F T + b_F)\right] + \left[\frac{K_U}{1+K_U}(m_U T + b_U)\right] \text{(1)}$$

where

$$K_U = e^{\frac{-\Delta G}{RT}} \text{(2)}$$

and

$$\Delta G = \Delta H \left(1 - \frac{T}{T_m}\right) - \Delta C_P \left(T_m - T + T \ln\left[\frac{T}{T_m}\right]\right) \text{(3)}$$

As previously described, the two terms in first equation account for the contributions of the

folded and unfolded protein, m_F/m_U and b_F/b_U account for the dependence of the dye when

the protein is folded/unfolded, thus incorporating the baselines and slopes, while K_U gives the

unfolded/folded equilibrium constant. T_m values were extracted from this initial fitting to the

DSF data, using ΔC_P as constant at 3.22 kcal/(mol.K), as estimated by the ScooP server [28],

and used to evaluate actual binders among the ligands selected from docking calculations.

For affinity constant calculations, the same protocol was used, varying the ligand

concentration from 0 to 2,000 µM. The thermal unfolding data was adjusted in the model

proposed by Bai and coworkers [27]. In this model, the fraction of unfolded protein f_U is given

as:

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$$f_U = \frac{[U]}{[U]+[F]+[FL]}\ (4)$$

165 Where [U] is the concentration of the unfolded protein, [L] is the free ligand concentration,

166 [F] and [FL] are the concentration of the folded in unbound and folded and bound protein,

167 respectively. The model assumes that only the folded species bind the ligand in the following

168 equilibrium:

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$$[U] + [L] \underset{K_U}{\leftrightarrow} [F] + [L] \underset{K_d}{\leftrightarrow} [FL]\ (5)$$

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$$[P] = [F] + [U] + [FL]\ (6)$$

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$$[L]_T = [L] + [FL]\ (7)$$

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$$K_U = {[U]}\big/{[F]}\ (8)$$

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$$K_d = \frac{[F][L]}{[FL]}\ (9)$$

175 Using equations 6-9, equation 4 can be rewritten as:

$$f_U = \frac{1}{1+[(1/K_U)\times(1+[L]/K_d)]} \quad (10)$$

After redefining the free ligand concentration [L] as a function of the total ligand

concentration $[L]_T$, f_U can be analyzed in terms of only two parameters, K_U and K_d, since $[L]_T$

and [P] are known [27]. All the data analysis was done using `DSF_fitting.py` python

script as described by Bai and coworkers [27]. All the tested compounds were purchased from

Merck-Sigma.

Molecular Dynamics Simulations

The docking pose found for Ticarcillin bound to *S. aureus* MnaA was used for

equilibrium MD simulations. Here, the same receptor preparation described was used. For

the ligand, RESP charges were computed after electrostatic potential calculation using HF/6-

31G*, using the same protocol as used for UDP and UDP-GlcNAc. Ligand atom types were

attributed according to GAFF2 [29]. The complex, containing MnaA and ticarcillin was

immersed in OPC solvation box with 150 mM NaCl and neutralized using tLeap. The system

was used for energy minimization, heating and equilibration using the same protocol as

previously described [16]. After, 5,000 minimization steps, the system was heated to 300 K in

50 ps using an NVT ensemble simulation, followed by 50 ps of an NPT simulation to equilibrate

the system density. For these preliminary equilibration steps, harmonic restraints were kept

in the initial coordinates of the ligands and the protein with an spring constant of 2.0

kcal/(mol.Å^2). Finally, 500 ps of NPT simulation was carried out for final equilibration,

followed a 600 ns productive simulation (NPT). For the final equilibration and productive

simulation, a weak positional restraint was kept at all α carbon atoms to avoid enzyme

opening with an spring constant of 1.0 kcal/(mol.Å^2). All the MD simulations were run with

SANDER, PMEMD and PMEMD.CUDA programs from the AMBER 20 package [30–33]. The MD trajectory was analyzed using AMBERENERGY++ [34], a locally written program to compute interaction energies, VMD [35], Chimera [25] and CPPTRAJ [36].

Results

Given the relevance of *MnaA* in the synthesis of the WTA and the antibiotic relevance, we used a structure-guided screening of compounds that could inhibit the enzymatic activity. For this purpose, we used the ZINC Drug Database (ZDD), a catalogue of almost 3,000 purchasable drugs that were approved for use in humans in different countries [20]. The rationale here involved the identification of binders that are already proved to be non-toxic and with a tolerable pharmacokinetic profile. In a typical drug discovery pipeline, the compounds could be further improved to improve its properties, including the affinity for the target. In this context, the identification of a binder in the narrow chemical space would be a challenge *per se*.

After visual inspection of the docking results, the following compounds were selected for experimental evaluation: methocarbamol, pantoprazole, 5-azacytidine, pyridoxal 5'-phosphate, sulfamethizole vetranal, suprofen, sitaxentan, cytarabine, (e)-5,5'-(diazene-1,2-diyl)bi, ticarcillin, nilutamide, adenosine, thymidine, tolbutamide, L-Histidine, sulfadiazin vetranal and sulfamethoxazole. The selection criteria involved the number of favorable interactions with polar residues in the MnaA active site, the absence of ligand polar groups lacking interactions with the receptor and a reasonable match in the occupied volume as compared to the substrate pose.

224   Experimental Screening of Ligand Candidates
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A rapid assay for screening of the ligand candidates was set up using a DSF assay.

Initially, the enzyme melting curve was characterized in the apo state (unbound) as well as

bound to UDP, UDP-N-Acetylglucosamine and UDP + UDP-N-acetylglucosamine together. As

indicated in Figure 1, the melting temperatures (T_M) for the apo MnaA and for the enzyme

bound to UDP do not change significantly (54.4 and 55.2°C, respectively), suggesting that (i)

the enzyme binds very weakly to UDP in the absence of the substrate or (ii) the enzyme could

be pre-loaded with UDP after bacterial expression [13]. When bound to UDP-GlcNAc, in the

other hand, an important shift in the melting temperature was observed, as shown in Figure

1, with a T_M of 58.2°C. Interestingly, when both UDP and UDP-N-acetylglucosamine were

incubated with MnaA, the observed T_m was 58.7°C, very close to the T_m observed for UDP-N-

acetylglucosamine alone.

Figure 1. Melting curves for unbound (apo) MnaA (black squares) and in complex with UDP (red squares), UDP-GlcNAc (green
squares) and UDP and UDP-GlcNAc simultaneously (blue squares). The red lines indicate the fit of the experimental data to
equations (1)-(3).

This preliminary result indicated that, even the enzyme substrate, binds with a

relatively small change in the T_m, of about 4 degrees. So, in order to identify potential MnaA

binders, we searched for ligands that resulted in a T_m change of at least 2 degrees in a DSF

screening experiment. From the 17 initially purchased compounds, nine showed a T_M shift of

2 degrees or higher and were selected for further experimental evaluation. The ligand

candidates initially tagged as potential hits were methocarbamol, pantoprazole,

sulfamethiazole, suprofen, cytarabine, olsalazine, ticarcillin, nilutamide and sulfadiazine,

while sulfamethoxazole, tolbutamide, histidine, thymidine, pyridoxal 5'-phosphate were

discarded as potential ligands. 5-azacytidine revealed fluorescence signal in the absence of

protein and was removed from our analysis.

Figure 2. Initial hits in the experimental ligand screening. SaMnaA (2μM) and the ligands identified in the virtual screening
(6μM) were preliminary tested for their binding properties using the DSF assay. A thermal shift of 2 degrees (red line) was
considered an indication of binding. In all cases, the incubation of the tested ligand with UDP increased the observed T_M.

For a further exploration of the binding potential of the selected compounds,

quantitative DSF assay was setup, as previously described by Bai and coworkers [27], where

the change in T_M is monitored as a function of the ligand concentration. First, a positive

control experiment was set up, using MnaA substrate, UDP-GlcNac. The analysis of the

melting curves over a concentration range of 0 - 2,000 μM revealed a clear right-shift in the

melting temperature (T_m) for an increasing ligand concentration, as shown in Figure 3.

According to the global fitting of the data, measured in two independent replicates, at

$T_m=54°C$, a K_U of 1.08 was achieved with a K_D of 33 μM. At $T_m=56°C$, a K_U of 3.22 with a K_D of

31 μM were observed. Here, K_U is an equilibrium constant between the unfolded and folded

enzyme. So, a KU of 1 gives equal concentrations of folded and unfolded species. Interestingly,

the K_M for UDP-GlcNAc measured for SaMnaA was reported to be 411 $\pm$ 57μM according to

Mann and coworkers [13]. Clearly, the differences can be due to the assumptions of an

equilibrium interaction rather than a kinetic (enzymatic) model. In line with this observation,

a plot of $\ln K$ as function of $1/T$ does not show a typical linear dependency expected for the

van't Hoff relationship, indicating that the enthalpy may not be temperature independent.

This could be related to the change in enzyme activity over the temperature range, resulting

in this non-linear relationship.

Figure 3. Isothermal analysis of UDP-GlcNAc binding to SaMnaA based on DSF data. (A) Thermal denaturation fluorescence data and fitting curve (red lines) for different ligand concentrations. (B) Fraction of unfolded protein as a function of UDP-GlcNac at different temperatures. (C) Shifts observed in the melting temperature T_M as a function of ligand concentration. (D) Overlay of the data shown in panel (B) of fraction of unfolded protein as a function of ligand concentration. In all cases, the red lines show the best fit obtained from the models shown in equations (1) to (10).

For most of the ligands selected in the initial screen, a clear T_M shift as a function of the ligand concentration could not be observed. Instead, a constant Tm was observed as a function of the ligand concentration, with a sharp increase in fluorescence in higher concentrations. This sharp increase in fluorescence can indicate some interaction between the ligand and fluorescent probe, for example.

For ticarcillin, however, the melting curves were measured in two replicates over a concentration range of 0-1,300 μM, limited by the ligand solubility, and at T_M=54°C a K_U of

289   1.08 was observed with a K_D of 421 μM (Figure 4). For ticarcillin, in contrast to what was

290   observed for UDP-GlcNAc, a linear relationship was observed for $\ln K$ as a function of $1/T$.

Figure 4. Isothermal analysis of ticarcillin binding to SaMnaA based on DSF data. (A) Thermal denaturation fluorescence data
and fitting curve (red lines) for different ligand concentrations. (B) Fraction of unfolded protein as a function of ticarcillin at
different temperatures. (C) Shifts observed in the melting temperature T_M as a function of ligand concentration. (D) Overlay
of the data shown in panel (B) of fraction of unfolded protein as a function of ligand concentration. In all cases, the red lines
show the best fit obtained from the models shown in equations (1) to (10).

The quantitative DSF data indicates that ticarcillin is a weak MnaA binder with a K_D in

the range of 400 μM, in our experimental conditions, what is about thirteen times greater

than the K_D measured for the substrate UDP-GlcNAc in our conditions. Although the

quantitative DSF approach does not measure the binding constant in an equilibrium situation,

previous data have shown that binding affinities agree very closely to the binding affinities

determined by other experimental approaches, such as Isothermal titration calorimetry (ITC)

[27].

Molecular Dynamics Simulation of the SaMnaA-Ticarcillin Complex
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The docked pose from ticarcillin bound to SaMnaA was used to generate the initial

coordinates of the solvated complex for an equilibrium MD simulation. The obtained 600 ns

trajectory was used for a cluster analysis in CPPTRAJ [36] and also in UCSF Chimera. In both

cases, two major cluster were identified with minor differences among them. Figure 5 show

the ligand bound to the SaMnaA, as observed in the most populated cluster in the MD

simulation.

Figure 5. Representative frame of the most populated cluster during the MD simulation of the SaMnaA-ticarcillin complex.
Solvent and non-polar hydrogen atoms were omitted for clarity. Ticarcillin is shown in gray sticks.

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As shown in Figure 5, several polar residues interact with ticarcillin in the binding site,

contributing to its binding affinity. Most importantly, H40, N240, H238, H205, R207 and R206

directly Interact with ticarcillin carboxylate groups as well as with carbonyl oxygen. As

previously described, it is interesting to note that many positively charged residues surround

the binding site, what would be expected for an enzyme that binding phosphorylated

substrates [16]. That said, ticarcillin seems to be a reasonable binder, given its total negative

charge.

The individual contributions of each polar residue in the vicinity of the catalytic cleft

were computed with `AmberEnergy++` and is shown in Figure 6 below. As evidenced in the

left panel, the charged arginine residues have the largest contribution to the binding energy,

with about -130 kcal/mol per residue, while the entire interaction energy between ticarcillin

and the environment (MnaA and explicit solvent) sums about -450 kcal/mol (Figure 6, right

panel). This finding is also in line with our previous analysis that showed that R206, R207 and

K239 were the residues with largest contribution to UDP-GlcNac binding to MnaA [16]. Here,

K239 contributes to about -50 kcal/mol to the ticarcillin binding energy.

Figure 6. Interaction energies among ticarcillin and MnaA. (left) Decomposition analysis of the individual contributions to the
binding energy for the polar residues in the vicinity of the active site. (right) Interaction energies computed for ticarcillin and
the entire protein, as well as its electrostatic and van der Waals terms. The blue line shows the interaction energy of the
ligand (ticarcillin) with the environment, i.e., protein and the explicit solvent.

Additional analysis of the interaction energy with MMPBSA.py [37] revealed an

interaction energy of -47 ± 6 kcal/mol, using a Generalized Born (GB) model from Hawkins,

Cramer and Truhlar [38]. In line with the previous analysis, ticarcillin binding to SaMnaA has

a van der Waals contribution of -35 ± 4 kcal/mol, while the electrostatic contributions sum -

239 $\pm$ 28 kcal/mol. The GB correction for the (implicit) solvent adds +232 $\pm$ 24 kcal/mol,

resulting in a final binding energy if -47 kcal/mol.

Discussion

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The limited number of molecular targets/pathways explored by existing antibiotics is

worrying [14]. The resistance problem is even more severe for some pathogens where high

resistance rates are observed. The data reported to the National Healthcare Safety Network

2015-2017 indicated, for example, that 48.4% of the *S. aureus* isolates from device-associated

healthcare associate infections in acute-care hospitals were found to be resistant to

methicillin (MRSA) [39]. 82.1% of the *Enterococcus faecium* isolates were found to be

vancomycin resistant and 43% of *Acinetobacter spp* were found to be resistant to multidrug

resistant [39]. These alarming trends highlight the need for new antibiotic targets and for the

basic and translational research on their mechanisms.

Here we described the structure-based identification of ticarcillin as a weak MnaA

binder. According to the docking model and to the binding mode observed in the MD

simulations, ticarcillin binds to the substrate pocket, putatively acting as a competitive

inhibitor, although this proposal still awaits for further experimental validation. This finding

has at least two interesting implications. The first implication is the usage of ticarcillin as a

scaffold for the development and optimization of compounds aiming higher affinity to

SaMnaA. Those compounds can be useful as chemical probes to better understand the effects

of SaMnaA inhibition in *S. aureus* cultures. Since ticarcillin is a β-lactam antibiotic, there is no

sense in testing it since a reduction in cell proliferation is already expected to be observed.

The second implication is associated with ticarcillin mechanism of action.

Interestingly, this β-lactam is expected to reach serum concentrations of up to 124 μg/ml

after an intravenous injection of 1 g [40], what means serum levels even between 300 and

400 μM. So, at these concentrations, MnaA could be expected to be partially bound to

ticarcillin. This off-target interaction for ticarcillin could explain, at least partially, the good

activity of ticarcillin on the Gram-negative bacteria *Pseudomonas aeruginosa*. *P. aeruginosa*

has a UDP-N-Acetylglucosamine 2-epimerase that shares about 51% in sequence identity with

S. aureus MnaA. The same is true for species of the *Acinetobacter* genus. Thus, it is tempting

to speculate that the MnaA inhibition may be associated to the ticarcillin activity on Gram-

negative bacteria. In line with this observation, it is interesting to note that the activity on

Gram-negative bacteria is only achieved at high concentrations [40], again suggesting a low-

affinity interaction with the target.

If this hypothesis is correct, is it possible to anticipate other pathogens that could be

susceptible to ticarcillin? According to a sequence comparison in BLAST [41], other UDP-N-

acetylglucosamine 2-epimerases are close orthologues of *S. aureus* MnaA, including a

Mycoplasma epimerase. Since *Mycoplasma* species are usually unresponsive to beta-lactams,

it may be a feasible experimental model to assess whether the UDP-N-acetylglucosamine can

inhibit the growth and, maybe, indicate a new therapeutic indication for ticarcillin. Although

this experiment is not accessible to us, we foresee very promising consequences of the

findings reported here.

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Author contributions

ASN wrote the original draft and supervised and administrated the project. ASN and ECA

involved in the hypothesis generation, definition of the methodology, data acquisition and

analysis. The final manuscript was revised by ASN and ECA.

Data accessibility

The data generated in this study are available from our group's GitHub account at

https://github.com/alessandronascimento/LabData/tree/main/MnaA-Ticarcillin.

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Figure 1. Melting curves for unbound (apo) MnaA (black squares) and in complex with UDP

(red squares), UDP-GlcNAc (green squares) and UDP and UDP-GlcNAc simultaneously (blue

squares). The red lines indicate the fit of the experimental data to equations (1)-(3).

Figure 2. Initial hits in the experimental ligand screening. SaMnaA (2 μM) and the ligands

identified in the virtual screening (6 μM) were preliminary tested for their binding properties

using the DSF assay. A thermal shift of 2 degrees (red line) was considered an indication of

binding. In all cases, the incubation of the tested ligand with UDP increased the observed T_M.

Figure 3. Isothermal analysis of UDP-GlcNAc binding to SaMnaA based on DSF data. (A)

Thermal denaturation fluorescence data and fitting curve (red lines) for different ligand

concentrations. (B) Fraction of unfolded protein as a function of UDP-GlcNac at different

temperatures. (C) Shifts observed in the melting temperature T_M as a function of ligand

concentration. (D) Overlay of the data shown in panel (B) of fraction of unfolded protein as a

function of ligand concentration. In all cases, the red lines show the best fit obtained from the

models shown in equations (1) to (10).

Figure 4. Isothermal analysis of ticarcillin binding to SaMnaA based on DSF data. (A) Thermal

denaturation fluorescence data and fitting curve (red lines) for different ligand

concentrations. (B) Fraction of unfolded protein as a function of ticarcillin at different

temperatures. (C) Shifts observed in the melting temperature TM as a function of ligand

concentration. (D) Overlay of the data shown in panel (B) of fraction of unfolded protein as a

function of ligand concentration. In all cases, the red lines show the best fit obtained from the

models shown in equations (1) to (10).

Figure 5. Representative frame of the most populated cluster during the MD simulation of

the SaMnaA-ticarcillin complex. Solvent and non-polar hydrogen atoms were omitted for

clarity. Ticarcillin is shown in gray sticks.

Figure 6. Interaction energies among ticarcillin and MnaA. (left) Decomposition analysis of

the individual contributions to the binding energy for the polar residues in the vicinity of the

active site. (right) Interaction energies computed for ticarcillin and the entire protein, as well

as its electrostatic and van der Waals terms. The blue line shows the interaction energy of the

ligand (ticarcillin) with the environment, i.e., protein and the explicit solvent.