

RSC Advances

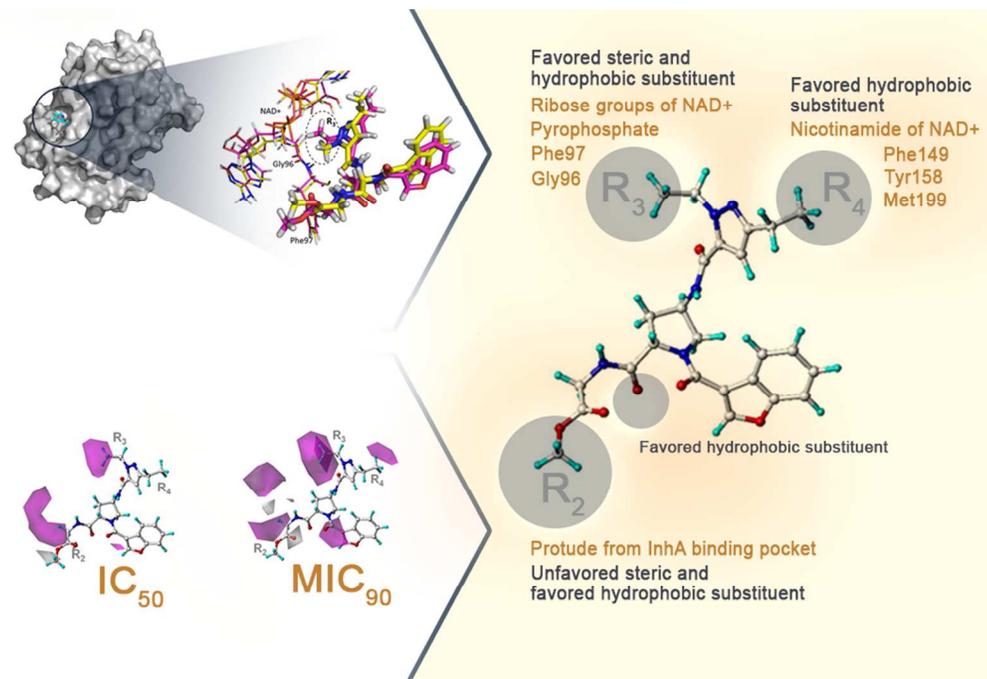


This is an *Accepted Manuscript*, which has been through the Royal Society of Chemistry peer review process and has been accepted for publication.

Accepted Manuscripts are published online shortly after acceptance, before technical editing, formatting and proof reading. Using this free service, authors can make their results available to the community, in citable form, before we publish the edited article. This *Accepted Manuscript* will be replaced by the edited, formatted and paginated article as soon as this is available.

You can find more information about *Accepted Manuscripts* in the [Information for Authors](#).

Please note that technical editing may introduce minor changes to the text and/or graphics, which may alter content. The journal's standard [Terms & Conditions](#) and the [Ethical guidelines](#) still apply. In no event shall the Royal Society of Chemistry be held responsible for any errors or omissions in this *Accepted Manuscript* or any consequences arising from the use of any information it contains.



The structural concept for enhancing both IC_{50} and MIC_{90} activities summarized from MD simulations and CoMSIA results
 90x60mm (300 x 300 DPI)

1 Structural Requirements of Benzofuran Pyrrolidine Pyrazole
2 Derivatives as Highly Potent Anti-tuberculosis Agents for
3 Good Correlation of IC₅₀-MIC Based on Integrated Results
4 from 3D-QSAR and MD Simulations
5

6 Pharit Kamsri^a, Auradee Punkvang^b, Supa Hannongbua^c, Patchreenart
7 Saparpakorn^c and Pornpan Pungpo^{a,*}
8

9 *^aDepartment of Chemistry, Faculty of Science, Ubon Ratchathani University,
10 Ubonratchathani, Thailand*
11

12 *^bFaculty of Science, Nakhon, Phanom University, Nakhon Phanom, Thailand*
13

14 *^cDepartment of Chemistry, Faculty of Science, Kasetsart University, Bangkok, Thailand*
15

16 ** All correspondence should be addressed to P. Pungpo: E-mail: pornpan_ubu@yahoo.com*
17 *Tel.: +664 535 3400 ext.4124, Fax: +664 528 8379*
18
19
20
21
22
23
24
25
26
27
28

ABSTRACT

A 2-*trans* enoyl-acyl carrier protein (ACP) reductase or InhA of *M. tuberculosis* is a drug target of isoniazid (INH), the first-line drug for tuberculosis treatment. Many series of compounds have been developed as novel inhibitors of this enzyme. However, they lack good potency against purified InhA and activity against intact *M. tuberculosis* cells. Benzofuran pyrrolidin pyrazole derivatives are potent direct InhA inhibitors. These compounds show high potency for InhA inhibition with IC₅₀ values at nanomolar levels. However, their activities against *M. tuberculosis* cells in terms of MIC₉₀ were about one-thousand fold than IC₅₀. Accordingly, in this work, IC₅₀ and MIC₉₀ values of benzofuran pyrrolidin pyrazole derivatives were subjected to CoMFA and CoMSIA studies in order to investigate the structural basis required for good activity against both purified InhA and *M. tuberculosis* cells. Moreover, MD simulations were employed to evaluate key interactions for binding benzofuran pyrrolidin pyrazole derivatives in InhA. Based on MD results, the core structure of these compounds is the key portion for binding in the InhA pocket. Alternatively, **R** substituents showed weak interactions with the InhA pockets. Interpretation of IC₅₀ and MIC₉₀ CoMSIA contour maps revealed the structural requirements in terms of steric, electrostatic, hydrophobic and hydrogen donor and acceptor for IC₅₀ and MIC₉₀ values of InhA inhibitors. Finally, the integrated results obtained from MD simulations and graphic interpretation of CoMSIA models provided a structural concept for rational design of novel InhA inhibitors with better potency against both the InhA enzyme and intact *M. tuberculosis* cells.

Keyword: *M. tuberculosis*; InhA; MD simulations, 3D-QSAR

1 1. Introduction

2 Tuberculosis (TB) is an infectious disease caused by *Mycobacterium tuberculosis*
3 (*M. tuberculosis*) and remains one of the world's deadliest infectious diseases. The World
4 Health Organization (WHO) reported that an estimated 9.0 million people developed new TB
5 cases and 1.5 million people died from this disease in 2013. Moreover, the incidence of new
6 TB cases and deaths in 2013 was higher than those reported previously.¹ The high mortality
7 rate of TB is caused by multi drug-resistant tuberculosis (MDR-TB),²⁻⁷ extensively drug-
8 resistant tuberculosis (XDR-TB),⁸⁻⁹ totally drug-resistant tuberculosis (TDR-TB)¹⁰⁻¹¹ and
9 human immunodeficiency virus (HIV) co-infection.¹ A NADH-dependent 2-*trans* enoyl-acyl
10 carrier protein (ACP) reductase or InhA has been identified as potential drug target for
11 tuberculosis treatment.¹² This enzyme catalyzes the reduction of α,β -unsaturated fatty acids,
12 the last step in fatty acids biosynthesis in *M. tuberculosis*.¹²⁻¹⁴ InhA was reported as the drug
13 target of isoniazid (INH), the first-line drug against tuberculosis.¹⁵⁻²³ Since INH is a prodrug,
14 it requires the activation process of catalase-peroxidase (KatG) to generate the acyl radical
15 active form. This radical is then covalently bound to nicotinamide adenine dinucleotide
16 (NAD^+) to produce an active INH-NAD adduct acting as a potent InhA inhibitor.¹⁸⁻²³ The
17 high potency of INH against InhA was lost by mutations in KatG. Therefore, many
18 researchers aimed to discover novel inhibitors that can directly inhibit InhA without the KatG
19 activation process. Inhibitors that can act like this are called direct InhA inhibitors. A class of
20 *N*-((3*R*, 5*S*)-1-(benzofuran-3-carbonyl)-5-carbamoylpyrrolidin-3-yl)-1*H*-pyrazole-5-
21 carboxamide derivatives (benzofuran pyrrolidin pyrazole derivatives) have been identified as
22 potent direct InhA inhibitors.²⁴ The majority of benzofuran pyrrolidine pyrazole derivatives
23 show high potency against purified InhA with inhibitory concentration of compound required
24 to inhibit InhA at 50% (IC_{50}) values at the nanomolar level. However, these compounds show
25 weak cellular activity against *M. tuberculosis*, with the minimum inhibitory concentration of
26 compound that resulted in complete inhibition in growth of *M. tuberculosis* 90% (MIC_{90}) at
27 the micromolar level. These results show poor correlation between IC_{50} and MIC_{90} values of
28 benzofuran pyrrolidine pyrazole derivatives. In this work, IC_{50} and MIC_{90} values of
29 benzofuran pyrrolidine pyrazole derivatives were used for comparative molecular field
30 analysis (CoMFA) and comparative molecular similarity indices analysis (CoMSIA) studies
31 in order to investigate the structural basis of these compounds for good activity against both
32 InhA and *M. tuberculosis*. Moreover, molecular dynamics (MD) simulations were employed
33 to evaluate the key interactions for binding of benzofuran pyrrolidin pyrazole derivatives in

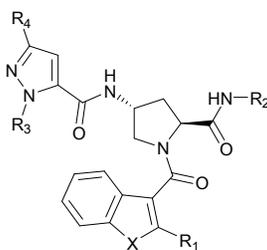
InhA. Therefore, the integrated results obtained from MD simulations and graphic interpretation of quantitative structure activity relationship (QSAR) models should provide crucial structural concepts for improving the correlation between IC₅₀ and MIC₉₀ values of benzofuran pyrrolidin pyrazole derivatives.

2. Material and Methods

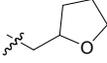
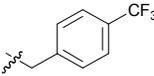
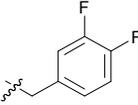
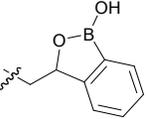
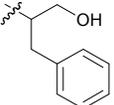
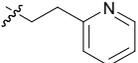
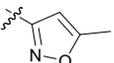
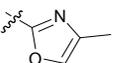
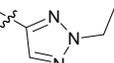
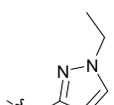
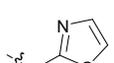
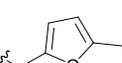
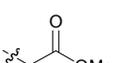
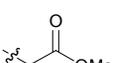
2.1 Data sets and biological activities

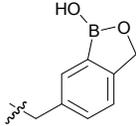
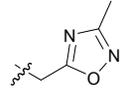
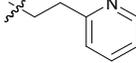
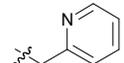
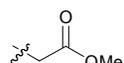
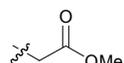
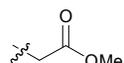
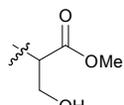
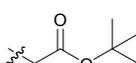
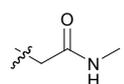
Thirty-four benzofuran pyrrolidin pyrazole derivatives used for CoMFA and CoMSIA studies were identified from the published literature.²⁴ Chemical structures and experimental biological activities in terms of MIC₉₀ and IC₅₀ values of these compounds are shown in **Table 1**. MIC₉₀ and IC₅₀ values were nominally converted into log (1/MIC₉₀) and log (1/IC₅₀) values for CoMFA and CoMSIA studies. Based on the diversity of structures and wide range of activities, the data set of compounds was divided into 30 training set compounds for final model development and 4 test set compounds for model validation. All chemical structures of benzofuran pyrrolidin pyrazole derivatives were constructed using the standard tools available in the GaussView 3.07 program and were then fully optimized using the HF/6-31G method implemented in the Gaussian 09 program.²⁵ The harmonic vibrational frequencies of the optimized geometries have also been calculated. All elements in the calculated Hessian matrix are positive, which indicate that the structures are true minima on the potential energy surface.

Table 1 The chemical structures and activities against InhA and *M. tuberculosis* of thirty-four benzofuran pyrrolidin pyrazole derivatives.



Cpd.	X	R ₁	R ₂	R ₃	R ₄	IC ₅₀ (μM)	MIC ₉₀ (μM)	log(1/IC ₅₀)	log(1/MIC ₉₀)
1	O	H	Et	Me	Et	0.034	8.00	7.47	5.10
2*	O	H		Me	Et	0.005	0.50	8.30	6.30
3	O	H	H	Me	Et	0.012	3.00	7.92	5.52

4	O	H	CH ₂ CF ₃	Me	Et	0.046	4.00	7.34	5.40
5	O	H	CH ₂ CH ₂ CH ₃	Me	Et	0.021	15.60	7.68	4.81
6	O	H	CH ₂ CH ₂ OMe	Me	Et	0.014	4.00	7.85	5.40
7*	O	H	CH ₂ CH ₂ COOEt	Me	Et	0.022	4.00	7.66	5.40
8	O	H		Me	Et	0.045	4.00	7.35	5.40
9	O	H		Me	Et	0.040	4.00	7.40	5.40
10	O	H		Me	Et	0.042	16.00	7.38	4.80
11*	O	H		Me	Et	0.009	2.00	8.05	5.70
12	O	H		Me	Et	0.035	3.00	7.46	5.52
13	O	H		Me	Et	0.112	1.00	6.95	6.00
14	O	H		Me	Et	0.025	1.00	7.60	6.00
15	O	H		Me	Et	0.018	16.00	7.74	4.80
16	O	H		Me	Et	0.009	8.00	8.05	5.10
17	O	H		Me	Et	0.003	4.00	8.52	5.40
18	O	H		Me	Et	0.032	4.00	7.49	5.40
19	O	H		Me	Et	0.005	1.00	8.30	6.00
20	O	H		Me	Et	0.021	1.50	7.68	5.82
21	O	H		Me	Cyclopropyl	0.015	1.00	7.82	6.00
22	O	H		Et	Et	0.003	0.05	8.52	7.30
23	O	H	H	Et	Et	0.004	0.50	8.40	6.30

24	O	H		Et	Et	0.002	0.20	8.70	6.70
25	O	H		Et	Et	0.004	0.50	8.40	6.30
26	O	H		Et	Et	0.004	0.50	8.40	6.30
27	O	H	CH ₂ CH ₂ OH	Et	Et	0.003	1.00	8.52	6.00
28	O	H		Et	Et	0.002	0.70	8.70	6.15
29	O	Et		Me	Et	0.005	0.70	8.30	6.15
30	S	H		Me	Et	0.029	1.00	7.54	6.00
31*	O	Ph		Me	Et	0.003	1.50	8.52	5.82
32	O	H		Me	Et	0.018	2.00	7.74	5.70
33	O	H		Me	Et	0.007	2.00	8.15	5.70
34	O	H		Me	Et	0.008	2.00	8.10	5.70

1 *test set

2

3 2.2 Molecular docking calculations

4 In this study, molecular docking calculations using the GOLD Program²⁶⁻³⁰ were
 5 employed with the aims of generating the initial structure for MD simulations and performing
 6 molecular alignment to set up CoMFA and CoMSIA models. The available X-ray structure of
 7 InhA in a complex with compound **1** (PDB code 4COD) was used as an initial structure for
 8 molecular docking calculations. All atoms of the protein were kept rigid, whereas ligand was
 9 flexible during the molecular docking calculations. The number of Genetic Algorithm (GA)
 10 runs was set to 15 runs with the default search algorithm parameters. The docking
 11 calculations were validated using the root-mean-square deviation (RMSD) value between the
 12 docked and observed X-ray conformations of compound **1** in its pocket. A RMSD value
 13 lower than 1 Å was acceptable. Then, molecular docking calculations with validated

1 parameters were used to dock all remaining compounds into the InhA binding pocket. The
2 binding mode that showed the lowest binding energy was selected for each compound and
3 was used to set up CoMFA and CoMSIA models. It was then used as the initial structure for
4 MD simulations of compounds **2**, **22**, **23** and **28**.

6 **2.3 Molecular dynamics simulations**

7 Compound **28**, with the best IC₅₀ value, was selected to investigate its binding mode
8 in InhA. Moreover, the binding modes of compounds **2**, **22** and **23** were modelled by MD
9 simulations in order to investigate the effect of **R**₂ and **R**₃ substituents on the IC₅₀ value. The
10 AMBER12 program³¹ was employed to perform molecular dynamics simulations. The
11 complex structures of compounds **2**, **22**, **23** and **28** in InhA obtained from molecular docking
12 calculations were used as the initial structure in MD simulations. The Amber *ff03* force field
13 was used for the physical description of InhA³². The general Amber force field (GAFF)³³⁻³⁴
14 and restrained electrostatic potential (RESP) partial charges³⁵⁻³⁸ of ligands and NAD⁺ were
15 generated by the *antechamber* module implemented in the AMBER12 package. To generate
16 the system for MD simulations, the initial complex structure was solvated by TIP3P water³⁹
17 in a truncated octahedral box extending up to 10 Å from the solute species. Five Na⁺ ions
18 were added to neutralize the system charge. Initially, the energy of system was minimized
19 using a steepest decent method followed by the conjugate gradient method. Then, the system
20 was gradually warmed from 0 K to 300 K in 30 ps by restraining all atoms of the complex
21 with a restraint weight of 2 kcal/molÅ². This was followed by 70 ps of the position-restrained
22 dynamics simulations with a restraining weight of 2 kcal/molÅ² at 300 K under an isobaric
23 condition. Finally, 10 ns MD simulations without any restraints were performed using the
24 same conditions. Long-range electrostatic interactions were applied using the Particle Mesh
25 Ewald method (PME)⁴⁰ during the simulations. The cut-off distance for the long-range van
26 der Waals interaction was set to 8 Å. The SHAKE method⁴¹ was applied to constrain the
27 bond lengths of hydrogen atoms attached to heteroatoms. Coordinates and energy outputs
28 during MD simulations were recorded at 2 ps intervals.

30 **2.4 Binding free energy calculations**

31 The Molecular Mechanics/Poisson–Boltzmann Surface Area (MM-PBSA) method⁴²⁻⁴⁵
32 was employed for calculating the binding free energy of compounds **2**, **22** and **23** in InhA. In
33 this calculation, 250 snapshots of the complex, receptor and ligand were extracted every 8 ps

1 from the last nanosecond of the MD trajectory, which represents the equilibrium state. The
 2 binding free energy (ΔG_{bind}) of compounds **2**, **22** and **23** complexed with InhA were
 3 estimated from equation 1, where ΔG_{vacuum} and ΔG_{solv} were the binding free energy of the
 4 complex in vacuum and the solvation free energy, respectively. In the MM-PBSA approach,
 5 the solvation free energy was calculated by solving a linearized Poisson-Boltzman equation.
 6 ΔG_{vacuum} was obtained by calculating the interaction energy between InhA and compounds **2**,
 7 **22** and **23** (ΔE_{MM}) and taking the entropy change ($T\Delta S$) as shown in equation 2. ΔE_{MM} is
 8 divided into three components, non-covalent van der Waals energy (ΔG_{vdW}), electrostatic
 9 energy (ΔG_{ele}) and internal energy (ΔG_{int}), as shown in equation 3. ΔE_{MM} and ΔG_{solv} were
 10 calculated using the SANDER module and a PBSA program of the AMBER suite,
 11 respectively. The entropy contribution was estimated using normal mode analysis with the
 12 NMODE module.⁴⁶ The entropy contribution was estimated using 250 snapshots for the
 13 binding free energy calculation.

14

$$15 \quad \Delta G_{\text{bind}} = \Delta G_{\text{vacuum}} + \Delta G_{\text{solv}} \quad \mathbf{1}$$

$$16 \quad \Delta G_{\text{vacuum}} = \Delta E_{\text{MM}} - T\Delta S \quad \mathbf{2}$$

$$17 \quad \Delta E_{\text{MM}} = \Delta G_{\text{vdW}} + \Delta G_{\text{ele}} + \Delta G_{\text{int}} \quad \mathbf{3}$$

18

19 **2.5 CoMFA and CoMSIA methods**

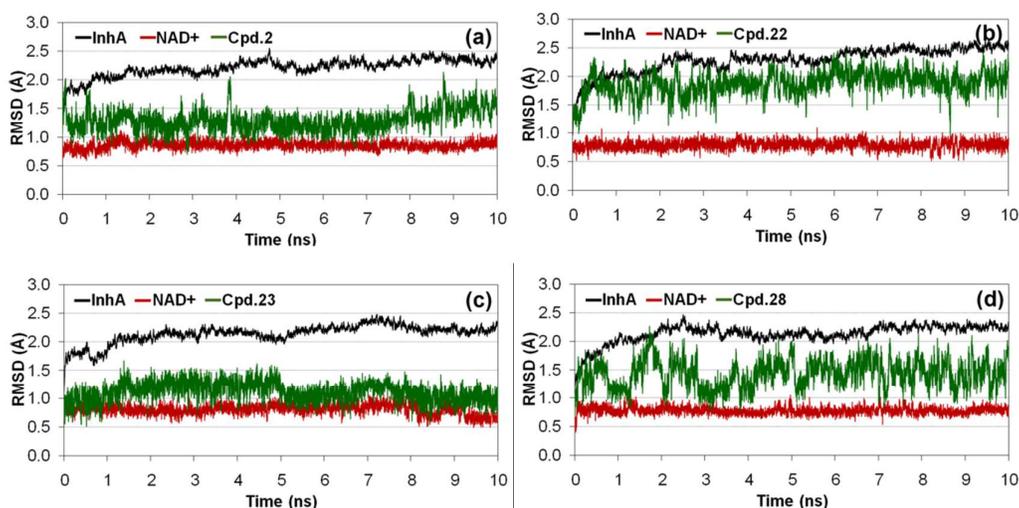
20 IC_{50} and MIC_{90} values of compounds were used to set up CoMFA⁴⁷ and CoMSIA⁴⁸
 21 models in order to evaluate the key structural features relating to the activity against both
 22 InhA and *M. tuberculosis*. The predicted binding modes of training set compounds obtained
 23 from molecular docking calculations were used for molecular alignment to set up CoMFA
 24 and CoMSIA models. SYBYL 8.0 molecular modelling software was used to run CoMFA
 25 and CoMSIA models. Partial least square (PLS) analysis was employed to derive a linear
 26 relationship between CoMFA and CoMSIA descriptor fields and activities. The PLS analysis,
 27 using the leave-one-out (LOO) cross-validation method, was performed to determine the
 28 optimal number of components. Sequentially, a final analysis with the optimal number of
 29 components was performed to construct CoMFA and CoMSIA models that were not
 30 cross-validated. The non-cross-validated correlation coefficient (r^2) and the leave-one-out
 31 cross-validated correlation coefficient (q^2) were used to evaluate the predictive ability of
 32 CoMFA and CoMSIA models. Selected CoMFA and CoMSIA models were employed to

1 predict IC_{50} and MIC_{90} values of test set compounds that were not used to construct models.
2 This was done to evaluate the external predictive ability of these models.

3. Results

3.1 Stability of the complex models

6 To reveal the structural stability of simulation system, the RMSD values for the
7 position of all solute species were separately analyzed. The RMSD plots for the four
8 simulation systems over 10 ns are shown in **Figure 1**. Convergent RMSD plots indicated that
9 the equilibrium state was reached for each system during this simulation period. As shown,
10 the RMSDs for compounds **2**, **22**, **23** and **28** in InhA converged after approximately 2 ns.



11
12
13
14 **Figure 1.** RMSD plots of compounds **2** (a), **22** (b), **23** (c), and **28** (d) complexed with InhA.

3.2 Reliability of the calculation methods

17 MD simulations were employed to model the binding modes of compounds **2**, **22**, **23**
18 and **28** in the InhA pocket. The experimental binding free energy (ΔG_{exp}) lying within the
19 experimental error of the calculated values (ΔG_{bind}) considered as the correlation between
20 the experimental binding free energy and the calculated values was used to indicate the
21 reliability of the modelled binding modes of these compounds. ΔG_{bind} values of compounds
22 **2**, **22**, **23** and **28** were close to their ΔG_{exp} values (Table 2). Therefore, we concluded that
23 MD simulations reliably modelled binding modes of compounds **2**, **22**, **23** and **28** in the InhA
24 pocket.

1 **Table 2** ΔG_{bind} and ΔG_{exp} of compounds **2**, **22**, **23** and **28** in InhA (kcal/mol).

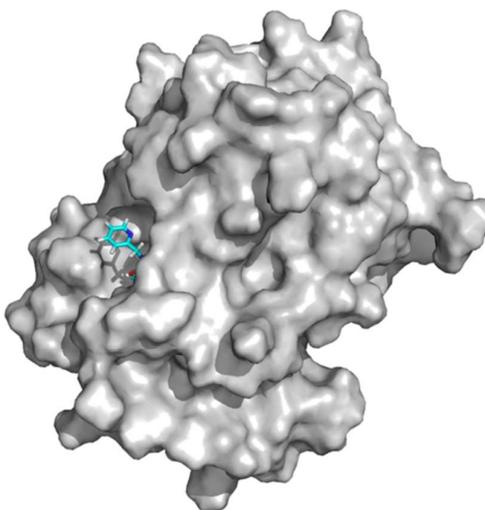
2

Cpd.	ΔH	$-T\Delta S$	ΔG_{bind}	ΔG_{exp}
2	-46.91±5.08	-31.03±6.06	-15.88±5.14	-15.52
22	-49.69±3.87	-33.15±6.41	-16.54±4.80	-15.82
23	-49.61±3.71	-32.79±5.57	-16.82±4.79	-15.65
28	-49.26±4.45	-32.52±6.58	-16.74±5.34	-16.07

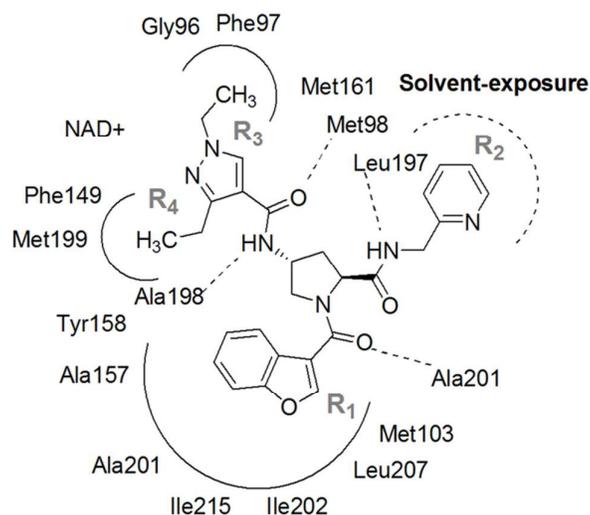
3

4 **3.3 Binding mode of compound 28**

5 The binding mode of compound **28** complexed with InhA obtained from MD
 6 simulations is shown in **Figure 2**. Residues located near each substituent and the core
 7 structure are listed in **Figure 3**. A hydrogen atom (the **R₁** substituent) is near the carbonyl
 8 backbone of Met103. 2-pyridinyl methyl (the **R₂** substituent) protrudes from the InhA pocket
 9 and interacts with the solvent (**Figure 2**). The ethyl moiety (the **R₃** substituent) is located near
 10 backbones of Gly96, Phe97 and pyrophosphate and ribose groups of NAD⁺. The ethyl group
 11 (the **R₄** substituent) was located in the hydrophobic side chains of Phe149, Tyr158, Met199
 12 and nicotinamide of NAD⁺. With regard to the core structure, the pyrazole ring in the core
 13 structure was sandwiched between two hydrophobic side chains of Met161 and Ala198. CO
 14 and NH of pyrazole amide formed hydrogen bonds with the backbones of Met98 and Ala198,
 15 respectively. The benzofuran core was buried in the hydrophobic side chains of Ile215,
 16 Ala157, Ile202 and Ala201, and was sandwiched between the hydrophobic side chains of
 17 Leu207 and Met103. The carbonyl of benzofuran core formed a hydrogen bond with the NH
 18 backbone of Ala201. NH of pyrrolidine amide formed a hydrogen bond with the CO
 19 backbone of Leu197.



1
2 **Figure 2.** Compound **28** (cyan) in its complex with whole InhA (grey) obtained from MD
3 simulations.



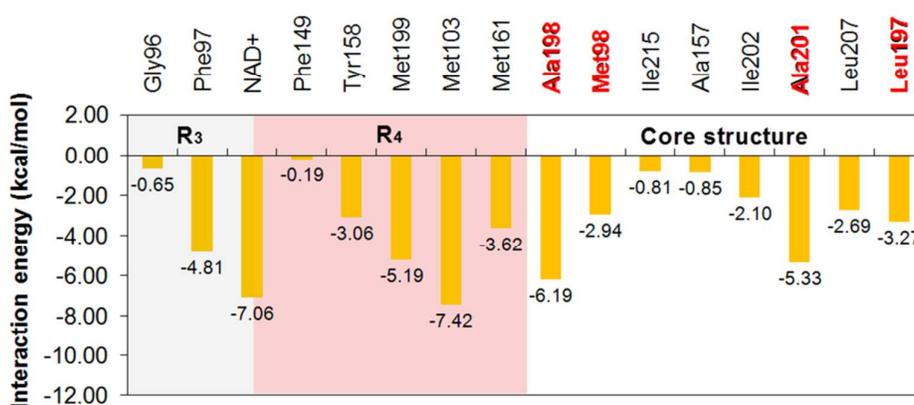
4
5 **Figure 3.** List of residues surrounding within 4 Å from compound **28**.

6

7 **3.4 Interaction energy**

8 Free-energy decomposition calculations were used to investigate the interaction
9 energies between compound **28** and each residue in the InhA pocket. **Figure 4** shows these
10 interaction energies obtained from free-energy decomposition calculations. The lowest
11 interaction energy (-7.42 kcal/mol) was observed for Met103, indicating that this residue had
12 the largest contribution to binding of compound **28** in the InhA pocket. As previously
13 mentioned, Met103 and Leu207 were sandwiched in the benzofuran core. Another
14 remarkable interaction energy (-7.06 kcal/mol) was found for NAD⁺. This was responsible

1 for van der Waal and electrostatic interactions with the **R**₃ and **R**₄ substituents of compound
 2 **28** (Figure 3). Ala198 showed an interaction energy (-6.16 kcal/mol), comparable with those
 3 of Met103 and NAD⁺. This residue formed hydrogen bonds with the NH of pyrazole amide
 4 and sandwiched the pyrazole ring (Figure 3). Met98, Leu197 and Ala201 formed other
 5 hydrogen bonds with the core structure with interaction energies of -2.94, -3.27 and -5.33
 6 kcal/mol, respectively. Based on interaction energy profile of compound **28**, the core
 7 structure formed more attractive interactive energies with surrounding residues than **R**
 8 substituents (Figure 4). This result indicates that the core structure is the key fragment for
 9 binding of this compound in the InhA pocket.



10

11 **Figure 4.** Interaction energy profile of compound **28** and surrounding residues within 4 Å.

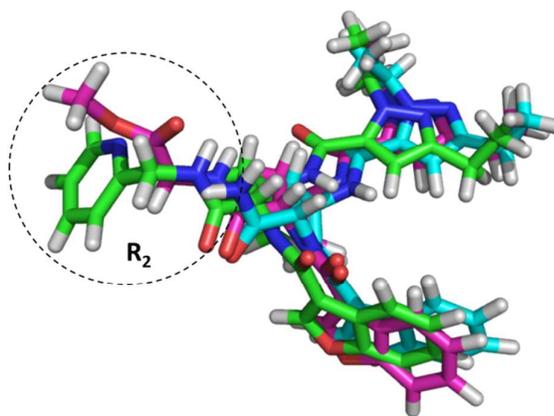
12

13 3.5 The effect of the **R**₂ substituent on IC₅₀ and MIC₉₀ values

14

15 As compared with the positions of other **R** substituents, the **R**₂ position had the most
 16 varied substituents (Table 1). Compound **28** exposing the 2-pyridylmethyl at the **R**₂ position
 17 showed the best activity for InhA inhibition with an IC₅₀ of 0.002 μM. When the **R**₂
 18 substituent of this compound was replaced by CH₂COOMe (compound **22**), the IC₅₀ value
 19 was slightly changed to 0.003 μM. In contrast, the MIC₉₀ value against whole
 20 *M. tuberculosis* cell was greatly changed from 0.7 μM to 0.05 μM (Table 1). To reveal the
 21 effect of the **R**₂ substituent on the IC₅₀ value, the binding modes of compounds **28** and **22**
 22 were compared (Figure 5). The binding modes of these compounds in the InhA pocket were
 23 similar, and the **R**₂ substituents occupied in the same positions. Moreover, the interaction
 24 energy profiles of compounds **28** and **22** with residues in InhA pocket were similar (Figure
 25 6). As discussed above, the **R**₂ substituent of compound **28** protruded from the InhA pocket
 leading to weak interaction of this substituent with the pocket. Therefore, the IC₅₀ value

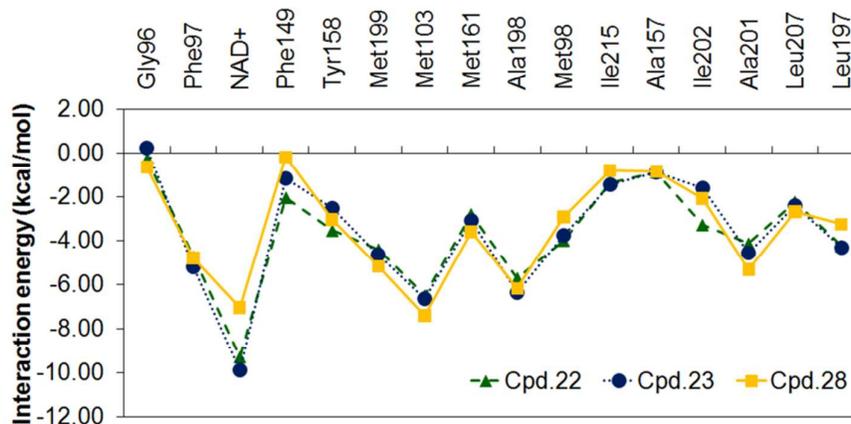
1 against InhA was not significantly changed when the R_2 substituent was varied. When the R_2
2 substituent was replaced by a hydrogen atom (compound **23**), the binding mode and
3 interaction energy profile of this compound were similar to those of compounds **22** and **28**
4 (**Figures 5** and **6**). With regard to IC_{50} values, compound **23** showed a comparable IC_{50} value
5 with those of compounds **22** and **28**. However, the MIC_{90} value of this compound ($0.5 \mu M$)
6 was largely increased over that that of compound **22** ($0.05 \mu M$). These results indicate that
7 the R_2 substituent had a small effect on the IC_{50} value against InhA due to its weak
8 interaction with the InhA pocket. Alternatively, this substituent is crucial to controlling the
9 MIC_{90} against intact *M. tuberculosis* cells.



10

11

12 **Figure 5.** The superimposition of binding modes of compounds **22** (pink), **23** (cyan) and **28**
13 (green).



14

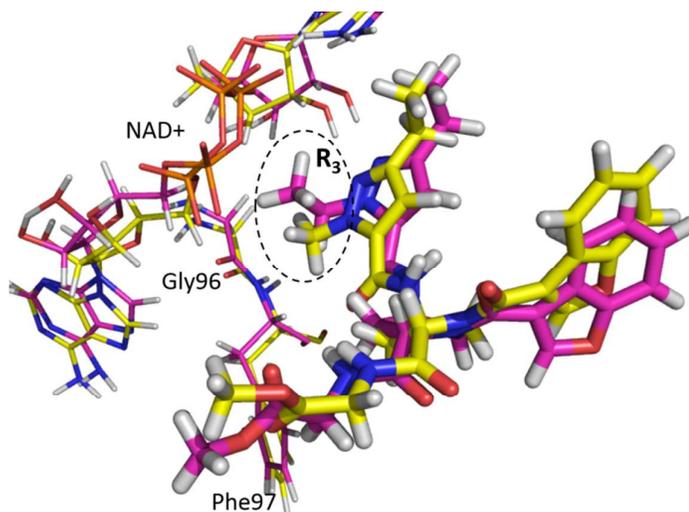
15 **Figure 6.** Comparison of the interaction energy profiles of compounds **22** (green), **23** (blue)
16 and **28** (yellow) with surrounding pocket within 4 Å.

17

18

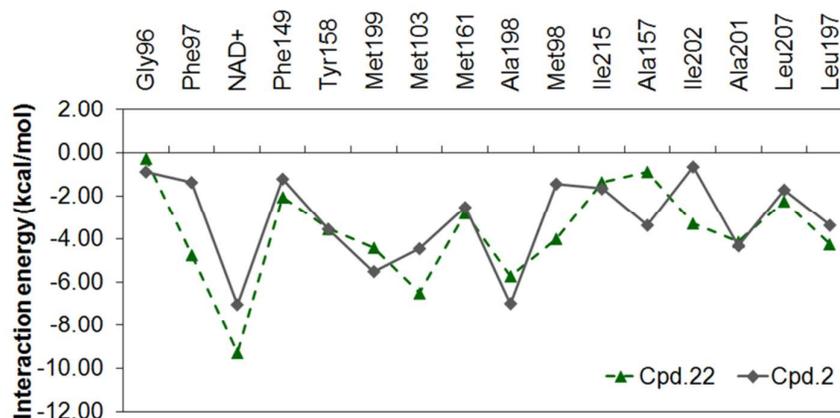
3.6 The effect of the R_3 substituent on IC_{50} and MIC_{90} values

The R_3 substituent of compounds in the data set was varied as ethyl (Et) or methyl (Me) groups (Table 1). Compounds 2 and 22 with structural differences at the R_3 substituent were selected to show the effect of the R_3 substituent on IC_{50} and MIC_{90} values. IC_{50} values of these compounds (0.005 and 0.003 μ M, respectively) were not significantly different, but their MIC_{90} values were tenfold different (0.5 and 0.05 μ M, respectively). Figure 7 shows the binding modes of compounds 2 and 22 in InhA obtained from MD simulations. The R_3 substituents of these compounds were located in the same position and surrounded by backbones of Gly96, Phe97 as well as pyrophosphate and ribose groups of NAD^+ . The ethyl group (The R_3 substituent) of compound 22 is close to Phe97 and pyrophosphate and ribose groups of NAD^+ more than the methyl group of compound 2. Therefore, interaction energies of compound 22 with Phe97 and NAD^+ had greater attraction than those of compound 2 (Figure 8). Moreover, the presence of a methyl group at the R_3 position of compound 2 shifted the position of benzofuran core surrounded by Met103 and Ile202, and disrupted hydrogen bond interaction with Met98. Accordingly, interaction energies of compound 2 with Met98, Met103 and Ile202 showed less attraction than those of compound 22 (Figure 8). These results indicate that compound 22 should have a better IC_{50} against InhA compared to compound 2. However, other than the interaction energies of Met98, Met103, Ile202, Phe97 and NAD^+ , compounds 2 and 22 are comparable. The IC_{50} value for InhA inhibition by compound 22 was slightly better than that of compound 2. However, its MIC_{90} value was tenfold better than that of compound 2. The results indicated that the ethyl group at the R_3 position is more conducive to favorable IC_{50} and MIC_{90} values than the methyl group.



23

24 **Figure 7.** The superimposition of binding modes of compounds 2(yellow) and 22 (pink).



1
2 **Figure 8.** Comparison of the interaction energy profiles of compounds **2** (gray) and **22**
3 (green) with surrounding pocket within 4 Å.
4

5 **3.7 CoMFA and CoMSIA models**

6 In this study, CoMFA and CoMSIA models were constructed from IC₅₀ and MIC₉₀
7 where prefixed with IC₅₀ and MIC₉₀, respectively. IC₅₀ and MIC₉₀ CoMSIA models were
8 constructed based on various combinations of molecular descriptor fields, in order to develop
9 a highly predictive CoMSIA model (**Tables 3 and 4**). An IC₅₀ CoMSIA model constructed
10 from the combination of steric (S), electrostatic (E), hydrophobic (H) and hydrogen acceptor
11 (A) fields⁴⁸ gave the highest q^2 (0.646), whereas an MIC₉₀ CoMSIA model including steric,
12 electrostatic, hydrophobic and hydrogen donor (D) fields⁴⁸ showed the highest q^2 (0.639).
13 Therefore, these models were selected for graphical interpretation of IC₅₀ and MIC₉₀
14 CoMSIA contour maps. In order to assess the predictive abilities of IC₅₀ and MIC₉₀ CoMSIA
15 models, IC₅₀ and MIC₉₀ values of the test set were predicted. Both IC₅₀ and MIC₉₀ CoMSIA
16 models showed good ability to predict IC₅₀ and MIC₉₀ values of the test set data as shown in
17 **Figure 9**. In case of IC₅₀ and MIC₉₀ CoMFA models, they had poor predictive ability with q^2
18 values of 0.464 and 0.432, respectively. Accordingly, these CoMFA models were not used
19 further in this work.
20
21
22
23
24
25
26

1 **Table 3** Statistical results of IC₅₀ CoMFA and CoMSIA models.

2

Models	Statistical parameters						Fraction
	q ²	r ²	s	SEE	N	F	
CoMFA							
S/E	0.464	0.996	0.392	0.035	6	909.618	60.3/39.7
CoMSIA							
S/E	0.084	0.977	0.512	0.081	6	162.845	32.1/67.9
S/H	0.465	0.950	0.383	0.118	5	90.431	29.1/70.9
S/D	0.624	0.923	0.321	0.145	5	57.579	54.3/45.7
S/A	0.146	0.970	0.495	0.093	6	123.724	39.7/60.3
S/E/H	0.260	0.981	0.460	0.074	6	194.704	16.6/44.5/38.9
S/E/D	0.592	0.980	0.342	0.076	6	185.576	21.0/53.7/25.3
S/E/A	0.281	0.975	0.454	0.085	6	149.701	22.5/42.8/34.7
S/E/H/D	0.646	0.990	0.318	0.055	6	363.962	13.1/35.8/28.5/22.6
S/E/H/A	0.336	0.983	0.436	0.070	6	222.520	12.3/31.5/29.4/26.8
S/E/H/D/A	0.610	0.991	0.334	0.050	6	437.341	10.0/25.4/22.6/20.7/21.4

3

4 Bold values indicate the best CoMSIA model

5 *N* optimum number of components; *s* standard error of prediction; *SEE* standard error of
6 estimate; *F* F-test value; *S* steric field; *E* electrostatic field; *H* hydrophobic field; *D* hydrogen
7 donor field; *A* hydrogen acceptor field

8

9

10

11

12

13

14

15

16

17

18

1 **Table 4** Statistical results of MIC₉₀ CoMFA and CoMSIA models.

2

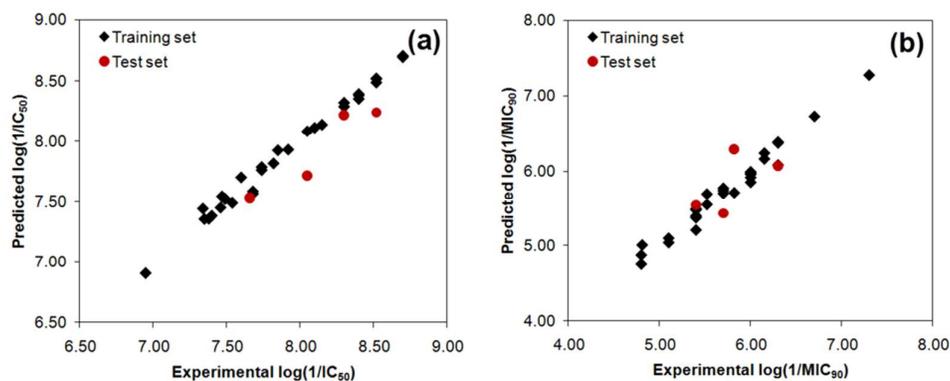
Models	Statistical parameters						Fraction
	q ²	r ²	s	SEE	N	F	
CoMFA							
S/E	0.432	0.853	0.442	0.225	2	78.451	53.2/46.8
CoMSIA							
S/E	0.456	0.949	0.469	0.143	6	71.455	25.1/74.9
S/H	0.459	0.780	0.432	0.275	2	47.970	34.4/65.6
S/D	0.261	0.732	0.514	0.310	3	23.642	52.7/47.3
S/A	0.602	0.978	0.401	0.093	6	174.060	46.3/53.7
S/E/H	0.477	0.961	0.460	0.126	6	93.558	13.8/52.8/33.4
S/E/D	0.210	0.912	0.553	0.184	5	49.990	17.7/64.4/18.0
S/E/A	0.550	0.955	0.426	0.134	6	82.091	19.9/48.1/32.0
S/E/H/D	0.415	0.938	0.476	0.155	5	72.712	10.9/45.8/29.3/13.9
S/E/H/A	0.639	0.973	0.382	0.105	6	136.014	12.5/35.6/42.2/27.7
S/E/H/D/A	0.494	0.961	0.442	0.123	5	118.951	9.3/33.4/22.8/10.4/24.2

3

4 Bold values indicate the best CoMSIA model

5 *N* optimum number of components; *s* standard error of prediction; *SEE* standard error of
 6 estimate; *F* F-test value; *S* steric field; *E* electrostatic field; *H* hydrophobic field; *D* hydrogen
 7 donor field; *A* hydrogen acceptor field

8



9

10 **Figure 9.** The plot of experimental and predicted activities of the training and test data sets
 11 derived from IC₅₀ (a) and MIC₉₀ (b) CoMSIA models.

1

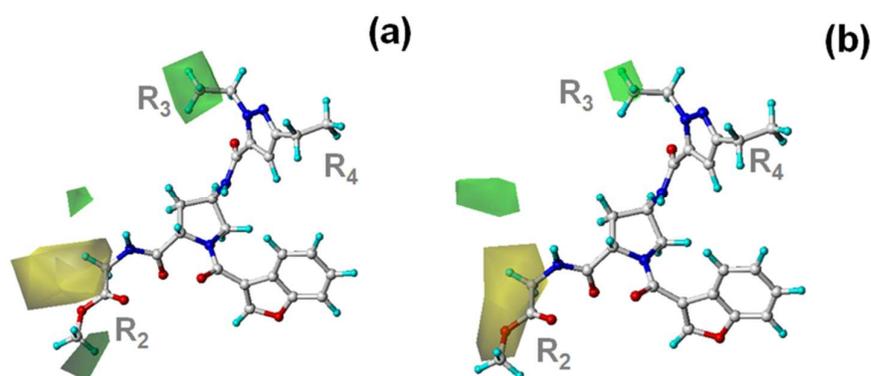
2 **3.8 CoMSIA contour maps**

3 To reveal the importance of molecular descriptor fields in both IC_{50} and MIC_{90} values
4 of InhA inhibitors, CoMSIA contour maps were established. Compound **22** presented the best
5 MIC value. Graphical interpretation of its IC_{50} and MIC_{90} CoMSIA contour maps was done.
6 Interpretation of its IC_{50} and MIC_{90} CoMSIA contour maps revealed structural requirements
7 in terms of steric, electrostatic, hydrophobic and hydrogen donor and acceptor fields for IC_{50}
8 and MIC_{90} values of InhA inhibitors.

9

10 **3.9 Steric requirements for IC_{50} and MIC_{90} values**

11 **Figure 10** shows the CoMSIA steric contour maps obtained from selected IC_{50} and
12 MIC_{90} CoMSIA models. These contours highlight the steric requirements for IC_{50} and MIC_{90}
13 values of benzofuran pyrrolidine pyrazole derivatives. Both IC_{50} and MIC_{90} CoMSIA models
14 show a green contour at the R_3 substituent. These results indicated that a bulky R_3 substituent
15 is favourable for both IC_{50} and MIC_{90} values. Accordingly, an ethyl group is more preferred
16 for the steric requirement of the R_3 substituent than a methyl group. This is consistent with
17 the MD simulations since an ethyl group can form more interactions with InhA. At the R_2
18 position, IC_{50} and MIC_{90} CoMSIA models present a large yellow contour. However, IC_{50}
19 CoMSIA model shows a favorable green steric contour at the terminal of the R_2 substituent
20 (**Figure 10a**). Based on MD simulations results, the R_2 substituent had weak interaction with
21 the InhA pocket leading to less influence on the IC_{50} value. Therefore, the steric requirement
22 of R_2 substituent should be based on the MIC_{90} CoMSIA steric contour that presented only a
23 yellow contour near this substituent (**Figure 10b**).



24

25 **Figure 10.** Steric contour maps of IC_{50} (a) and MIC_{90} (b) CoMSIA models in combination
26 with compound **22**.

27

3.10 Electrostatic requirements for IC_{50} and MIC_{90} values

Electrostatic requirements for IC_{50} and MIC_{90} values of benzofuran pyrrolidine pyrazole derivatives are visualized in **Figure 11**. Both IC_{50} and MIC_{90} CoMSIA contours show only an electrostatic requirement at the R_2 substituent. The IC_{50} CoMSIA shows a red contour at the ester moiety of R_2 substituent, whereas MIC_{90} CoMSIA presents a blue contour at this position. These results show different electrostatic requirements for IC_{50} and MIC_{90} values of benzofuran pyrrolidin pyrazole derivatives. However, the R_2 substituent has weak influence on the IC_{50} value. Therefore, the electrostatic requirement of R_2 substituent for MIC_{90} values should take more priority.

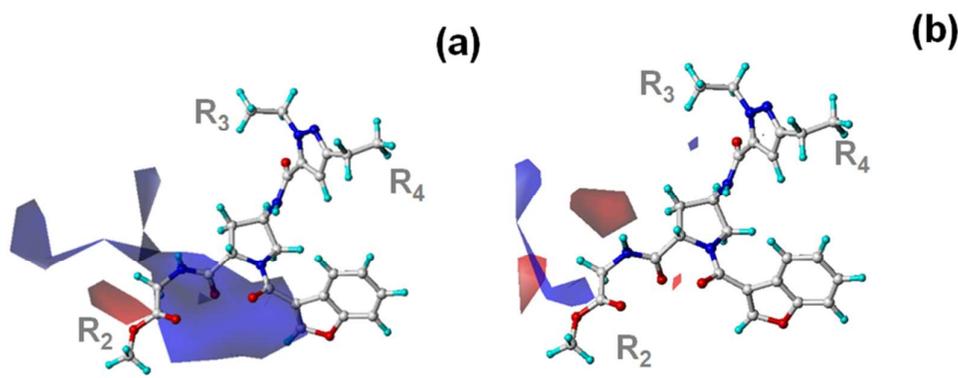
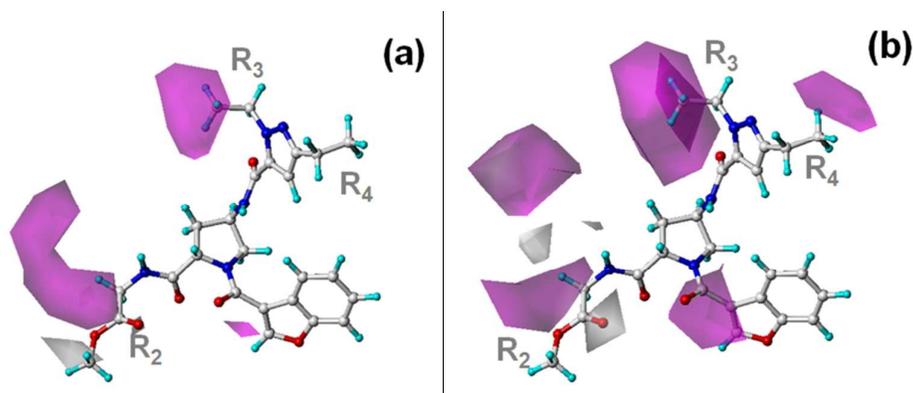


Figure 11. Electrostatic contour maps of IC_{50} (a) and MIC_{90} (b) CoMSIA models in combination with compound **22**.

3.11 Hydrophobic requirements for IC_{50} and MIC_{90} values

Both IC_{50} and MIC_{90} CoMSIA contours show a purple contour at the R_3 substituent of compound **22** (**Figure 12**). This shows that the hydrophobic requirements of the R_3 substituent for both IC_{50} and MIC values were similar. The R_3 substituent was either a methyl or ethyl group. As seen in **Figure 12**, the terminal of ethyl group was buried in a purple R_3 contour. Therefore, the ethyl group was preferable for the hydrophobic requirement of the substituent. IC_{50} and MIC_{90} values of compound **2** with the methyl group at the R_3 substituent were weaker than those of compound **22** containing an ethyl group. At the R_2 substituent, both IC_{50} and MIC_{90} CoMSIA contours display a purple contour at this position (**Figure 12**). Therefore, the presence of a hydrophobic substituent at this purple region should enhance both IC_{50} and MIC_{90} values. The grey contour located at the carbonyl moiety of the R_2 substituent in both IC_{50} and MIC_{90} CoMSIA contours indicated that this moiety is important for both IC_{50} and MIC_{90} values. Another important hydrophobic contour is located at the R_4 substituent. The MIC_{90} CoMSIA shows a purple region near the R_4 substituent (**Figure 12b**),

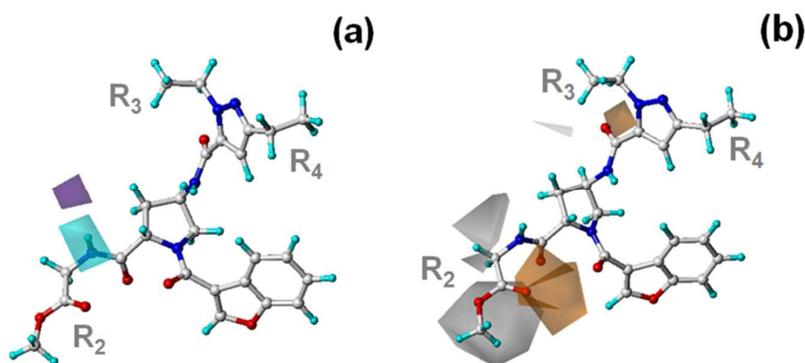
1 but this contour disappeared in the IC_{50} CoMSIA contour (**Figure 12a**). Therefore, a
2 hydrophobic moiety could be presented at purple region to enhance the MIC_{90} value without
3 a negative contribution to the IC_{50} value.



4
5 **Figure 12.** Hydrophobic contour maps of IC_{50} (a) and MIC_{90} (b) CoMSIA models in
6 combination with compound **22**.

8 *3.12 Hydrogen donor and acceptor requirements for IC_{50} and MIC_{90} values*

9 The hydrogen donor field was included in the selected IC_{50} CoMSIA model, but this
10 molecular descriptor was instead changed to a hydrogen acceptor field in the selected MIC_{90}
11 CoMSIA model. The IC_{50} CoMSIA model did not show any hydrogen donor contour near
12 any **R** substituents. However, this model showed a favourable hydrogen donor contour at the
13 amide moiety of the core structure. The amide moiety appears to impact the IC_{50} value.
14 Consistent with the MD simulations results, this moiety can form hydrogen bonds with
15 Leu197. The MIC_{90} CoMSIA model shows a favourable hydrogen acceptor contour at the
16 carbonyl moiety of **R**₂ substituent, indicating that this moiety is essential to a good MIC_{90}
17 value.

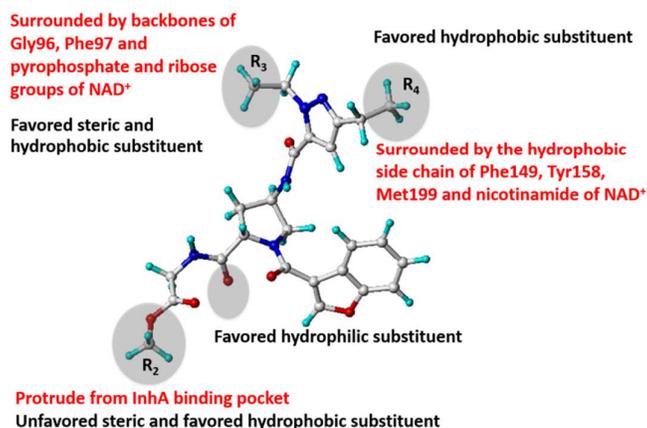


18
19 **Figure 13.** Hydrogen donor contour of IC_{50} CoMSIA model (a) and hydrogen acceptor
20 contour MIC_{90} CoMSIA model (b) in combination with compound **22**.

1

2 **3.13 The structural concept for good IC_{50} and MIC_{90} correlation**

3 Based on the MD simulations results, the core structure of benzofuran pyrrolidine
4 pyrazole derivatives is of key importance for binding in the InhA pocket. Therefore, this
5 fragment is crucial for favorable IC_{50} values. Among all **R** substituents, the **R₂** substituent has
6 the least interaction with the InhA pocket because it protrudes from the pocket. Modifications
7 of the **R₂** substituent did not significantly change IC_{50} values, but rather produced a tenfold
8 increase in MIC_{90} values (compounds **22** and **23**). Accordingly, the **R₂** substituent is a key
9 group that can be used to adjust the MIC_{90} value without negative contribution to the IC_{50}
10 value. Based on the results obtained from our MD simulations and CoMSIA studies, the
11 structural concept to correctly balance IC_{50} and MIC_{90} values of benzofuran pyrrolidin
12 pyrazole derivatives is summarized in **Figure 14**. New compounds designed based on this
13 concept should show better IC_{50} and MIC_{90} values.



14

15 **Figure 14.** The structural concept for good IC_{50} and MIC_{90} correlation summarized from MD
16 simulations and CoMSIA results. Red and black letters indicate the results
17 obtained from MD simulations and CoMSIA results, respectively.

18

19 **4. Conclusion**

20 The combination of MD simulations and graphical interpretation of IC_{50} and MIC_{90} CoMSIA
21 models highlight the structural concept to correctly balance IC_{50} and MIC_{90} values of
22 benzofuran pyrrolidin pyrazole derivatives. The core structure of template compound is
23 crucial to attaining favorable IC_{50} values, whereas the **R₂** substituent is a key group to
24 enhance MIC_{90} values without negative effects on IC_{50} values. Modifications of **R**
25 substituents following the structural concept suggested here should allow design of novel

1 InhA inhibitors with better potency against both the InhA enzyme and intact *M. tuberculosis*
2 cells.

3

4 **Acknowledgements**

5 This research was supported by the Thailand Research Fund (DBG5680003, MRG5680169)
6 the National Research Council of Thailand and Higher Education Research Promotion. The
7 financial support from Royal Golden Jubilee Ph.D. Program (PHD/0004/2554) to P. Kamsri
8 is gratefully acknowledged. Faculty of Science, Ubon Ratchathani University, Kasetsart
9 University and NECTEC are gratefully acknowledged for supporting this research.

10

11 **References**

- 12 1. Global Tuberculosis Report; World Health Organization: Geneva, 2014.
- 13 2. M. C. Becerra, J. Bayona, J. Freeman, P. E. Farmer, J. Y. Kim, *Int. J. Tuberc. Lung Dis.*,
14 2000, **4**, 387–394.
- 15 3. C. Dye, M. A. Espinal, C. J. Watt, C. Mbiaga, B. G. Williams, *J. Infect. Dis.*, 2002, **185**,
16 1197–1202.
- 17 4. M. A. Espinal, *Tuberculosis (Edinburgh)*, 2003, **83**, 44–51.
- 18 5. A. Wright, M. Zignol, A. Van Deun, D. Falzon, S. R. Gerdes, K. Feldman, S. Hoffner, F.
19 Drobniowski, L. Barrera, D. van Soolingen, F. Boulabhal, C. N. Paramasivan, K. M.
20 Kam, S. Mitarai, P. Nunn, M. Raviglione, *Lancet*, 2009, **373**, 1861–1873.
- 21 6. C. Y. Chiang, R. Centis, G. B. Migliori, *Respirology*, 2010, **15**, 413–432.
- 22 7. Multidrug and Extensively Drug-Resistant TB (M/XDR-TB):2010 Global Report on
23 Surveillance and Response (WHO/HTM/TB/2010.3). World Health Organization:
24 Geneva, 2010.
- 25 8. N. S. Shah, A. Wright, G. H. Bai, L. Barrera, F. Boulabhal, N. Martin-Casabona, F.
26 Drobniowski, C. Gilpin, M. Havelkova, R. Lepe, R. Lumb, B. Metchock, F. Portaels, M.
27 F. Rodrigues, S. Rüscher, A. Van Deun, V. Vincent, K. Laserson, C. Wells, J. P.
28 Cegielski, *Emerging Infect. Dis.*, 2007, **13**, 380–387.
- 29 9. M. Berry, O. M. Kon, *Eur. Respir. Rev.*, 2009, **18**, 195–197.
- 30 10. A. A. Velayati, P. Farnia, M. R. Masjedi, T. A. Ibrahim, P. Tabarsi, R. Z. Haroun, H. O.
31 Kuan, J. Ghanavi, M. Varahram, *Eur. Respir. J.*, 2009, **34**, 1202–1203.
- 32 11. Z. F. Udhwadia, R. A. Amale, K. K. Ajbani, C. Rodrigues, *Clin. Infect. Dis.*, 2012, **54**,
33 579–581.

- 1 12. A. Quemard, J. C. Sacchettini, A. Dessen, C. Vilcheze, R. Bittman, W. R. Jacobs, J. S.
2 Blanchard, *Biochemistry*, 1995, **34**, 8235–8241.
- 3 13. C. Vilcheze, H. R. Morbidoni, T. R. Weisbrod, H. Iwamoto, M. Kuo, J. C. Sacchettini,
4 W. R. Jacobs Jr., *J. Bacteriol.*, 2000, **182**, 4059–4067.
- 5 14. D.A. Rozwarski, C. Vilchèze, M. Sugantino, R. Bittman, J.C. Sacchettini, *J. Biol.*
6 *Chem.*, 1999, **274**, 15582–15589.
- 7 15. D.A. Rozwarski, G.A. Grant, D.H. Barton, W.R. Jacobs Jr., J.C. Sacchettini, *Science*,
8 1998, **279**, 98–102.
- 9 16. C. Vilcheze, F. Wang, M. Arai, M.H. Hazbon, R. Colangeli, L. Kremer, T.R. Weisbrod,
10 D. Alland, J.C. Sacchettini, W.R. Jacobs Jr., *Nat. Med.*, 2006, **12**, 1027–1029.
- 11 17. A. Dessen, A. Quemard, J.S. Blanchard, W.R. Jacobs Jr., J.C. Sacchettini, *Science*, 1995,
12 **267**, 1638–1641.
- 13 18. K. Johnsson, P.G. Schultz, *J. Am. Chem. Soc.*, 1994, **116**, 7425–7426.
- 14 19. B. Lei, C.J. Wei, S.C. Tu, *J. Biol. Chem.*, 2000, **275**, 2520–2526.
- 15 20. K. Johnsson, D.S. King, P.G. Schultz, *J. Am. Chem. Soc.*, 1995, **117**, 5009–5010.
- 16 21. Y. Zhang, B. Heym, B. Allen, D. Young, S. Cole, *Nature*, 1992, **358**, 591–593.
- 17 22. A. Quemard, A. Dessen, M. Sugantino, W.R. Jacobs Jr., J.C. Sacchettini, J.S. Blanchard,
18 *J. Am. Chem. Soc.*, 1996, **118**, 1561–1562.
- 19 23. A. Banerjee, E. Dubnau, A. Quemard, V. Balasubramanian, K.S. Um, T. Wilson, D.
20 Collins, G. de Lisle, W.R. Jacobs Jr., *Science*, 1994, **263**, 227–230.
- 21 24. L. Encinas, H. O'Keefe, M. Neu, M.J. Remuiñán, A.M. Patel, A. Guardia, C.P. Davie, N.
22 Pérez-Macías, H. Yang, M.A. Convery, J.A. Messer, E. Pérez-Herrán, P.A. Centrella, D.
23 Alvarez-Gómez, M.A. Clark, S. Huss, G.K. O'Donovan, F. Ortega-Muro, W. McDowell,
24 P. Castañeda, C.C. Arico-Muendel, S. Pajk, J. Rullás, I. Angulo-Barturen, E. Alvarez-
25 Ruíz, A. Mendoza-Losan, L. Ballell Pages, J. Castro-Pichel, G. Evindar, *J. Med. Chem.*,
26 2014, **57**, 1276–1288.
- 27 25. M. Frisch, G. Trucks, H. Schlegel, G. Scuseria, M. Robb, J. Cheeseman, G. Scalmani, V.
28 Barone, B. Mennucci, G. Petersson, H. Nakatsuji, M. Caricato, X. Li, H. Hratchian, A.
29 Izmaylov, J. Bloino, G. Zheng, J. Sonnenberg, M. Hada, M. Ehara, K. Toyota, R.
30 Fukuda, J. Hasegawa, M. Ishida, T. Nakajima, Y. Honda, O. Kitao, H. Nakai, T. Vreven,
31 J. Montgomery, J. Peralta, F. Ogliaro, M. Bearpark, J. Heyd, E. Brothers, K. Kudin, V.
32 Staroverov, R. Kobayashi, J. Normand, K. Raghavachari, A. Rendell, J. Burant, S.
33 Iyengar, J. Tomasi, M. Cossi, N. Rega, J. Millam, M. Klene, J. Knox, J. Cross, V.

- 1 Bakken, C. Adamo, J. Jaramillo, R. Gomperts, R. Stratmann, O. Yazyev, A. Austin, R.
2 Cammi, C. Pomelli, J. Ochterski, R. Martin, K. Morokuma, V. Zakrzewski, G. Voth, P.
3 Salvador, J. Dannenberg, S. Dapprich, A. Daniels, Farkas, J. Foresman, J. Ortiz, J.
4 Cioslowski and D. Fox, Gaussian 09, Gaussian, Inc., Wallingford CT, 2009.
- 5 26. G. Jones, P. Willett, R.C. Glen, *J. Mol. Biol.*, 1995, **245**, 43-53.
6 27. G. Jones, P. Willett, R.C. Glen, A.R. Leach, R. Taylor, *J. Mol. Biol.*, 1997, **267**, 727-748.
7 28. J.W. Nissink, C. Murray, M. Hartshorn, M.L. Verdonk, J.C. Cole, R. Taylor, *Proteins*,
8 2002, **49**, 457-471.
9 29. M.L. Verdonk, J.C. Cole, M.J. Hartshorn, C.W. Murray, R.D. Taylor, *Proteins*, 2003, **52**,
10 609-623.
11 30. M.L. Verdonk, G. Chessari, J.C. Cole, M.J. Hartshorn, C.W. Murray, J.W. Nissink, R.D.
12 Taylor, R. Taylor, *J. Med. Chem.*, 2005, **48**, 6504-6515.
13 31. D.A. Case, T.A. Darden, T.E. Cheatham III, C.L. Simmerling, J. Wang, R.E. Duke, R.
14 Luo, R.C. Walker, W. Zhang, K.M. Merz, B. Roberts, S. Hayik, A. Roitberg, G. Seabra,
15 J. Swails, A.W. Götz, I. Kolossváry, K.F. Wong, F. Paesani, J. Vanicek, R.M. Wolf, J.
16 Liu, X. Wu, S.R. Brozell, T. Steinbrecher, H. Gohlke, Q. Cai, X. Ye, J. Wang, M.-J.
17 Hsieh, G. Cui, D.R. Roe, D.H. Mathews, M.G. Seetin, R. Salomon-Ferrer, C. Sagui, V.
18 Babin, T. Luchko, S. Gusarov, A. Kovalenko, P.A. Kollman, AMBER 12, University of
19 California, San Francisco.
20 32. Y. Duan, C. Wu, S. Chowdhury, M.C. Lee, G. Xiong, W. Zhang, R. Yang, P. Cieplak, R.
21 Luo, T. Lee, J. Caldwell, J. Wang, P.A. Kollman, *J. Comput. Chem.*, 2003, **24**, 1999–
22 2012.
23 33. J. Wang, R.M. Wolf, J.W. Caldwell, P.A. Kollman, D.A. Case, *J. Comput. Chem.*, 2004,
24 **25**, 1157-1174.
25 34. J. Wang, W. Wang, P. A. Kollman, D.A. Case, *J. Mol. Graphics Model.*, 2006, **25**, 247-
26 260.
27 35. P. Cieplak, W.D. Cornell, C. Bayly, P.A. Kollman, *J. Comput. Chem.*, 1995, **16**, 1357–
28 1377.
29 36. W.D. Cornell, P. Cieplak, C.I. Bayly, P.A. Kollmann, *J. Am. Chem. Soc.*, 1993, **115**,
30 9620–9631.
31 37. C.I. Bayly, P. Cieplak, W. Cornell, P.A. Kollman, *J. Phys. Chem.*, 1993, **97**, 10269–
32 10280.
33 38. J. Wang, P. Cieplak, P.A. Kollman, *J. Comput. Chem.*, 2000, **21**, 1049–1074.

- 1 39. M.W. Mahoney, W.L. Jorgensen, *J. Chem. Phys.*, 2000, **112**, 8910–8922.
- 2 40. T. Darden, D. York, L. Pedersen, *J. Chem. Phys.*, 1993, **98**, 10089–10092.
- 3 41. J.P. Ryckaert, G. Ciccotti, H.J.C. Berendsen, *J. Comp. Phys.*, 1977, **23**, 327–341.
- 4 42. N. Homeyer, Gohlke, H. *Mol. Inf.*, 2012, **31**, 114–122.
- 5 43. J. Wang, T. Hou, X. Xu, *Curr. Comput. Aided Drug Des.*, 2006, **2**, 95–103.
- 6 44. J. Wang, P. Morin, W. Wang, P.A. Kollman, *J. Am. Chem. Soc.*, 2001, **123**, 5221–5230.
- 7 45. T. Hou, J. Wang, Y. Li, W. Wang, *J. Chem. Inf. Model.*, 2011, **51**, 69–82.
- 8 46. M. Kaledin, A. Brown, A.L. Kaledin, J.M. Bowman, *J. Chem. Phys.*, 2004, **121**, 5646–
9 5653.
- 10 47. R. D. Cramer, III, D. E. Patterson, J. D. Bunce, *J. Am. Chem. Soc.*, 1988, **110**, 5959
11 5967.
- 12 48. G. Klebe , U. Abraham , T. Mietzner, *J. Med. Chem.*, 1994, **37**, 4130–4146.