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Kinase Inhibitors as Potential Drugs: A Molecular Dynamics Simulation Study

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Kinase Inhibitors as Potential Drugs: A Molecular Dynamics Simulation Study

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ABSTRACT

Molecular dynamics simulation and binding free energy (ΔG_{bind}) calculations were done to inspect the interaction between five inhibitors and PDK-1 kinase. The free energy (ΔG_{bind}) values were computed using MM-GBSA and MM-PBSA free energy calculation methods.

The entropic contribution of the binding free energy ΔS was computed using normal mode (NMODE) method. The change of enthalpy (ΔH) was calculated using the equation $\Delta G=\Delta H$ -T. ΔS .

There is a noticeable difference in the values of ΔG depending on the calculation method whether MM-PBSA or MM-GBSA, and this is due to the calculation different approach in each case.

PDK-1 kinase is a well validated anticancer target. The results gave the binding modes between PDK-1 kinase and the five inhibitors, which can be used in the future in the drug design processes for cancer treatment. The placement of water molecules in the binding sites are known. This can be used to design better inhibitors through adding substituents to the inhibitor to replace a water molecule that binds kinase in the active site based on the creation of an inhibitor that includes a structural water mimic. Through Molecular dynamics simulation, we identify potency PDK-1 inhibitor (5) that have unique binding to the inactive kinase conformation (DFGout). On the other hand, inhibitors (1-4) are consider as classical ATP-competitive kinase inhibitors (Type I) which are bind to the active conformation DFG-in.

It was reported that **type I kinase inhibitors** form water-mediated hydrogen bond networks (both water molecules W1 and W2 are commonly observed) and the ligand does not extend to the water-filled cavity. These two features distinguish type I from type II inhibitors and these two features were obvious in our study in the binding modes of inhibitors (1-4) with the PDK-1 kinase.

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منخص

PDK في هذه الدراسة قمنا باستخدام التمثيل الجزيئي من اجل حساب قيم طاقة الربط لخمس مثبطات لانزيم الكاينيز PDK 1 –باستخدام الطريقتين (Molecular Mechanics/Generlized-Born surface area) 2 – باستخدام الطريقة (Molecular Mechanics/Poisson Boltzmann surface area) 2 – 4 – 1.05 عند المعادلة AG = ΔH – 1.05 باستخدام هذه المعادلة AG = ΔH – 1.05.

هناك فرق ملحوظ في قيم التغير في طاقة الربط اعتمادا على طريقة الحساب سواء باستخدام (MM-PBSA or MM-GBSA) بحيث كانت نتائج MM-GBSA افضل من نتائج MM-PBSA وهذا يرجع الي الاختلاف في طريقة الحساب التي يتم اعتمداها من قبل MM-PBSA and MM-GBSA.

كذلك وضحت نتائج هذه الدراسة كيفية ارتباط هذه المثبطات بانزيم الكاينيز I-DDK والتي يمكن استخدامها لعلاج السرطان, بحيث أن موقع جزيئات الماء كانت موجودة في مواقع الربط ويمكن استخدام هذه النتائج في تصميم مثبطات أفضل من خلال إضافة ذرات مثل الأوكسجين والهيدروجين للمثبط لكي يكون قادرا على تكوين روابط هيدروجينية لكي يحل مكان جزيئات الماء التي كانت موجودة في مواقع الربط من خلال الاستناد على مبدا تكوين مثبطات جديدة محاكية لجزيتات الماء.

من خلال محاكاة الديناميات الجزيئية قمنا بتحديد مثبط (المركب 5) والذي يرتبط بشكل فريد بأنزيم الكاينيز PDK-1 لتكوين كاينيز غير نشيط على نقيض المركبات الأخرى(1-4) والتي يتم اعتبراها مثبطات كاينيز كلاسيكية منافسة ل ATP . ومن الجدير بالذكر ان المثبط (المركب 5) هو دواء فعال وذلك بسبب قيمة طاقة الربط الكبيرة Cal/mol هذه القيمة الكبيرة ناتجة من تكوين خمس روابط هيدروجينية قوية بين هذا المثبط وأنزيم الكاينيز PDK-1. وفقا لما ذكرته التقارير فان النوع الأول من مثبطات الكاينيز (مثبطات كلاسيكية منافسة ل ATP) تتميز عن النوع الثاني من المثبطات من خلال تشكيل شبكات من الروابط الهيدروجينية بواسطة جزيئين من الماء وهذا يتوافق مع نتائج هذه الدراسة بحيث انه تم ملاحظة هذه الميزة في المثبطات (المركب4) و (المركب3).

Chapter 1

1. INTRODUCTION

In this work, structure-based drug design that employs molecular dynamics, binding free energy calculations is used to investigate anti-cancer inhibitors of 3-phosphoinositide dependent kinase-1 (PDK-1)¹. This computational drug design approach has been successfully applied to both lead optimization and hit identification against PDK-1¹.

Fast expansion in this area has been made possible by advances in software and hardware computational power and sophistication, identification of molecular targets, and an increasing database of publicly available target protein structures. CADDD is being utilized to identify hits (active drug candidates), select leads (most likely candidates for further evaluation), and optimize leads i.e. transform biologically active compounds into suitable drugs by improving their physicochemical, pharmaceutical, ADMET/PK (pharmacokinetic) properties.²

The purpose of usages of computational tools as we hope to improve effectiveness and efficiency of drug discovery and development process, decrease use of animals, and increase predictability.²

1.1 Computer-aided drug design

A general strategy for drug discovery efforts can be summarized by the following steps: Identifying a target, screening for inhibitors, isolating hits, optimization and selection of a candidate molecule for clinical studies¹. Drug discovery is a very long and expensive process in which twelve to twenty-four years are needed to discover and develop a new drug. In addition, the average cost to develop a new drug into markets is more than \$1 billion¹.

Historically, Paul Ehrlich³ was the first person to postulate on the existence of chemoreceptors that can be exploited therapeutically. It was reported that initial stages in chemotherapy was focused on the isolation and purification of active ingredients from natural products such as plants. This was followed by rational approach which was based on understanding the mechanisms of action and the drugreceptor interactions³. Despite using the newer approach, it was reported that only 18 new chemical entities (NCE) were approved in the year 2005 and 2006⁴. For this reason, a newer approach was developed to increase the efficiency of drug discovery process.

Computer-aided drug design (CADD) is an example on the newer approaches that is using computer based techniques to analyze molecules and molecular systems to predict their biological properties.⁵ This approach helps in the identification and optimization of new potential drugs⁵. Computational drug design has played an important role in the successful development of marketed drugs such as saquinavir, ritonavir, and indinavir were utilized in the treatment of human immunodeficiency virus (HIV)⁶.

There are two broad strategies in computational drug design:⁵

- 1. Using Ligands for drug design
- 2. Receptor-based drug design Related to structure

The first strategy is a ligand-based drug design (LBDD) that is usually applied if a number of biologically active compounds are characterized and the target 3D structure is unknown⁵. By analyzing the physico-chemical properties of these active molecules, LBDD aims at predicting new chemical structures that are likely to have better biological properties⁶. There are many methods that use active known molecules to predict new ligands including Quantitative structure-activity

relationship, Pharmacophore modeling and shape-based screening methods⁷. For example, Quantitative structure-activity relationship (QSAR) method is based on the regression analysis of relationship between biological activity of set of homologous compounds and their various physico-chemical descriptors such as hydrophobic properties, electrostatic properties, steric factors, donor-acceptor⁵. Another example is the Pharmacophore modeling which based on accounting for direct protein–ligand binding, and ignores other interactions outside the pharmacophore region⁷. A Pharmacophore is defined as the three-dimensional representation of active chemical features of active compounds⁷.

The second strategy is Structure-based drug design methods are usually used when the X-ray crystal/NMR structure of the target protein is available⁵. The core strategy of this approach is based on analyzing the active ligand interaction with the binding site on the target protein. This means that ligands that exhibit similar interactions to the active ones will have similar biological effects⁶.

Ligand Docking and de novo drug design are two examples on the SBDD methods. Docking methods require the structure of the target protein to estimate the binding energy of a number of ligands and rank them according to their estimated binding free energies. In comparison, De novo drug design require the structure of the active site as starting point⁵.

1.2 Protein Kinases

There are more than 500 protein kinases known in the human genome. They are the second largest group of currently investigated drug targets ^{8,9}. Protein kinases main biological function is to catalyze the transfer of phosphoryl group of ATP to a hydroxyl group of threonine, tyrosine or serine residues¹⁰. Signal-transduction

pathways are activated by the phosphorylation of certain proteins which are responsible for the transition of cellular signals throughout the cells and to the nucleus¹⁰.

The deregulation of phosphorylation reaction due to mutations in kinase genes are known to cause 218 diseases. For this reason, protein kinases are considered as an important therapeutic target in different diseases and viral infections. Therefore, the protein kinases are considered as important effectors in human pathology¹⁰ and thereby a highly attractive therapeutic target in drug discovery⁹.

1.3 Role of the kinase enzymes in cancer

It was reported that abnormality in protein kinases can lead to the development of several reported disorders and major diseases such as, endocrine disorders, cardiovascular disease are due to malfunction of phosphorylation process¹¹.

There are 500 genes that encode kinases are involved in cancer, while other oncogenes activate kinases or are phosphorylated by other kinases. This what makes kinases as potential targets for drug development¹¹.

There are three specific sites in protein kinase that are involved in phosphorylation: an ATP binding site, a domain catalyzing the transfer of phosporyl group from ATP (phosphate pocket) and a substrate- binding site (PIF-pocket).¹²

Perturbed signal transduction provokes deregulation of different processes in cell migration, which can lead to malignant phenotype. 50% or more of receptor tyrosine kinase (RTK) and several serine/threonine kinases have been perturbed in different human malignancies. Irregular activity of a protein kinase which occur by genomic rearrangements result in hybrid proteins with catalytic domains of a protein kinase and another unrelated protein.¹² A second mechanism that damages the normal function of protein kinases is the mutations responsible for kinase constitutive activity. The third mechanism is explained by increasing expression of protein kinases. Finally, deregulation of kinase activity by activation of oncogenes can also contribute to tumorigenesis.

The chemotherapy treatment of breast cancer by using cyclophosphamide causes weight gain, ovarian failure, cardiac toxicity and Probability of developing a second cancers¹³. Studies estimated that in 2050 the global cancer will increase to 27million new cases¹⁴. Therefore, there are urgent need to discover a potent and selective cancer drug with no side effects.

1.4 3-Phosphoinositide dependent kinase-1 (PDK-1)

PDK-1 is a 556 amino acids serine/threonine kinase that belongs to the AGC protein kinase family. It plays an important role in the phosphorylation and activation of a number of proteins such as protein kinase B (PKB), protein kinase

C isoforms, the p70 ribosomal S6 kinase and serum and glucocorticoid-induced kinase¹⁵.

Structure of PDK-1 is consisting of two lobes: C-terminal lobe and N-terminal lobe and is similar in overall structure to PKA¹⁶. C-terminal pleckstrin homology domain (PH) is essential for interaction of PDK-1 with the cell membrane because it binds with phosphoinositide lipids of the plasma membrane¹⁵.

PDK-1 residues Val124, Val127 on the α -helix, Lys115, Ile119, Ile118 on the B-helix, and Leu155 on B-sheet form a hydrophobic pocket (PIF) pocket (Fig 1.1) ¹⁶. Since Leu155 is presented at the center of this pocket, whereas the other residues form a lining of the inside wall of the pocket¹⁷.



Figure 1.1: Structure of PDK-1 kinase domain with ATP molecule. The C-terminal lobe (in blue), the C-helix in green, the N-terminal lobe (in green), and the pSer241 in the T-loop (in purple/red spheres)¹⁶

S6K1 substrate interacts with the PIF-pocket of PDK-1 with higher affinity when it is phosphorylated at its hydrophobic motif. This indicated that the phosphate-binding site may be located close to the PIF-pocket.¹⁷

It was reported that the mutation of Leu155 to Glu canceled the ability of PDK-1 to interact with a peptide (PIFtide) substrates such as PRK2, S6K1 and SGK1.¹⁷ Whereas mutation of Ile119, Lys115, Leu155, and Glu150 to Ala decreased the affinity of PDK-1 to PIFtide binding substrates but did not abolish the ability of PDK-1 for phosphorylation and activation of PIFtide substrates such as S6K1 and SGK1.¹⁷



Figure 1.2: PIF-binding pocket of PDK-1 kinase

Phosphate-docking site is another small pocket lined with basic residues. This pocket is located in close vicinity to the PIF pocket (Fig 1.2). In the crystal structure

shown in figure 1.1, this pocket was occupied by a sulfate-anion that interacts with four residues lining the phosphate pocket, namely Gln150, Arg131, Lys76, and Thr148.¹⁶,¹⁷

The αC-helix (residues 129–131) is an important element in the core of PDK-1 structure formed from residues 124–136. It links both the N-terminal lobe and the C-terminal lobe with the active site as well as the phosphopeptide pocket with the phosphoserine in the T-loop. In particular, Val127 and Val124 are involved in formation of the hydrophobic pocket (PIF-pocket).

Arg129 and Arg131 form two hydrogen bonds with the phosphorylated Ser241 and sulfate in the phosphate pocket, respectively (Fig. 1.1). In addition, each of Glu130 and Lys111 forms a hydrogen bond with the phosphate of bound ATP that are crucial for kinase activation. Finally, Tyr126 forms a hydrogen bond with the phosphorylated Ser241.¹⁶

1.5 Identification of residues in the ATP pocket

The ATP binding pocket as described in Figure 1.3 consists of multiple regions¹⁸ as described below:

1) The Adenine region which is a conserved hydrophobic region. It is made up of residues at positions P2 (residue 88), P10 (residue 96), P13 (residue 109), P17 (residue142), P35 (residue 212). The adenine ring of ATP makes hydrophobic contacts with these five residues. In addition, it makes two hydrogen bonds with the

backbone of the hinge region residues (P20-P27). A third hydrogen bond occurs between two C-H groups of pyrimidiene ring with the carbonyl group of P23. The adenine-binding region is not characterized by large variability of amino acids, as a result of this it is not a good site for high degree of specificity.¹⁸



Figure 1.3: ATP binding pocket region: phosphate region (in magenta); sugar region (in green); Adenine region (in cyan); buried region (in violet) and solvent accessible region in (yellow)¹⁸

and P27 (residue 164).¹⁸ In 80.7% of protein kinases P27 residue is a serine, a glutamate, an aspartate or glutamine. The variability in P27 allows for the development of selective and potent inhibitors as demonstrated in the EGFR family of kinases where a unique cysteine placed in P27 position.¹⁸

3) Phosphate region which contains many highly polar residues and it consists of

two parts: (a) glycine-rich loop (GXGXXGXV: P3 -P10) lies on the N-terminal lobe.

It is the only one that shows significant conformational flexibility.¹⁸

(b) alpha-helix which consists an essential and conserved residues: which is made up of residues at position P14 (residue 111), P15 (residue 129), P33 (residue 209), P37 (residue 223) and P38 (residue 225).¹ P15 (residue 129) with three-dimensional location that makes an essential indicator of the active state of any kinase. P37 (residue 223) and P38 (residue 225) are conserved in all protein kinases and these are essential for the transfer of phosphate group from ATP to the substrate. This part gives an indication whether the kinases are in their active or inactive conformations.¹⁸

4) Buried region: the largest sequence diversity in the ATP pocket residues are found in this region, this region is not occupied by ATP, which is made up of residues at position P16 (residue 133), P17(residue 142), P18 (residue 144), P19 (residue 156), P20 (residue 158), P36 (residue 222).¹⁸ The residue in position P20 is important in determining the size of this specific region in the ATP binding pocket. P20 is often a bulky amino acid (40% methionine, 15% phenylalanine). It acts as a "molecular gate" to the buried ATP binding region. The introduction of a group to the buried region increases potency and, increases selectivity compared to that of kinases when this region is smaller.¹⁸

5) Solvent accessible regions: this region is important in exploited to increasing the binding affinity and to modulate ADME (toxicity) properties of ligands. The major difference in shape of solvent accessible area is contributed to the presence or

absence of glycine residue in positon P26. Often the NH of glycine forms intramolecular hydrogen bond with the carbonyl of P23 residue.¹⁸

1.6 Inhibition of PDK-1 Kinase enzyme

It was reported that overexpression of PDK-1 resulted in vitro and the PDK-1 phosphorylation was reported to suffer a high elevated levels in vivo breast cancers.¹⁹ This explained that there are a strong relationship between PDK-1 and malignant phenotype.

The main strategy of developing kinase inhibitors is to reduce ATP binding and/ or inhibit kinase activity¹². ATP and PDK-1 inhibitors compete in binding to the PDK-1 active site. When the PDK-1 inhibitors bind to PDK-1 active site they act to stop the transmission of phosporyl group from ATP to different amino acids. As a result, PDK-1 signal transduction is blocked. Development of PDK-1 inhibitors could lead to development of better treatment options for cancer.

specificity would not be a challenge if the target protein has unique catalytic functions and active site structures.²⁰ All 500 protein kinases encoded in the human genome have similar ATP-binding site structure.²¹ In the last decade, more than 50 patents of PDK-1 inhibitors were published in which the ATP-pocket within the kinase domain was the target.²²

Therefore, it is easy to establish the reason of kinase inhibitors being very specific and why the off-target effects are inevitable.²⁰ Nonetheless, Off-target effects are sometimes advantageous in clinical drugs. For example, Gleevec (Imatinib) was developed for treatment of chronic leukemia as an oral inhibitor of BCR-Abl. Currently, it has been approved for treatment of gastrointestinal stromal tumor (GIST).²⁰ This work demonstrates that the context of cells determine specificities of chemical inhibitors in vivo conditions.

Figure 1.4 shows that the specificities of inhibitors depend on cellular context. For example, in vitro standardized conditions were employed for enzyme-substrate and ATP concentrations, in addition to 'standard temperature and pressure' in physical chemistry. These conditions do not reflect the situation in living cells.²⁰ This reflect the facts that even we discovered the excellent drug in vitro condition, it may be not become an excellent one in vivo condition.



Figure 1.4: Context of target cells determined target specificities of inhibitors²⁰

1.7 Known potent drugs for cancer diseases

BX-320, BX-795 and BX-912 (Figure 1.5) are considered as potent and selective competitive inhibitors of PDK-1 enzyme activity with respect to its substrate (ATP). BX-320 which inhibit the PDK-1 signaling pathway in different cancer cell lines including MDA-453 (breast), U87-MG (glioblastoma), PC-3 (prostate), HCT-116 (colon), MiaPaCa (pancreatic) and LOX (melanoma)cells. BX-795 and BX-912 potently inhibited the growth of PC-3, U87-MG and MDA-453 cancer cell lines only.²³



Figure 1.5: Example on potent inhibitors of PDK-1: BX-795, BX-912 and BX-320, respectively²³

The high potency and selectivity of BX-320 is due to the formation of two hydrogen bonds between two nitrogens of amino-pyrimidine group with Ala162, which lies in the hinge region of the PDK-1 (Figure 1.6).²³



Figure 1.6: Structure of BX-320 bound to the ATP binding pocket of PDK-1²³

Singh et al²⁴ reported using molecular docking that myricetin (3,5,7-Trihydroxy-2-(3,4,5-trihydroxyphenyl)-4-chromenone) acts as a probable anticancer agent. Myricetin is naturally occurring flavanol and it is a polyphenolic compound.²⁴

Myricetin is considered as a potent PDK-1 inhibitor because of: (1) negative Docking energy of Myricetin-PDK-1 complex (-41 Kcal/mol), which indicates a favorable binding of Myricetin at the binding site of the PDK-1 kinase. (2) Formation of the most essential type of interaction between PDK-1 receptor and myricetin molecule (hydrogen bonding). The residues involved in formation of hydrogen bonds were Thr 222, Ala 162, Lys111, Asp 223, Ser 160, and Glu 130. Ala 162 and Ser 160 among these amino acid residues lie in the Hinge of PDK-1 protein. This type of interaction confirms that Myricetin fits into the active pocket of PDK-1receptor tightly (Table 1.1).²

Myricetin	Docking energy (Kcal/mol)	Hydrogen bonding residues	Hydrogen bond distances (Å)
	-41	O5: H-Lys ¹¹¹	2.44
НО ОН ОН		O5: H-Lys ¹¹¹	2.19
		O8: H-Ala ¹⁶²	2.34
но		H28: O-Glu ¹³⁰	2.31
о́н		H30: O-Asp ²²³	2.09
		H23: O-Ser ¹⁶²	2.14
		H33: O-Ser ¹⁶²	1.98

Ong et all²⁵ reported that Myricetin possesses both antioxidant properties and prooxidant properties, it also has a therapeutic potential in cancer treatment, cardiovascular diseases and diabetes mellitus. Benzo(a)- pyrenes cause cancer of the skin and lungs. Myricetin reduces the risk of skin cancer caused by polycyclic aromatic hydrocarbons.²⁵ Polycyclic aromatic hydrocarbons when metabolized produce carcinogenic metabolites.²⁵ Myricetin was found to inhibit the hydroxylation of benzo(a)pyr- ene in the human liver microsomes.²⁵

Virtual screening, NMR-based fragment screening, and ultrahigh throughput screening (UHTS) led to the identification of diverse chemicals as PDK-1 inhibitors which bind the PDK-1 kinase in the ATP-site with at least one H-bond towards the hinge region. The first four inhibitors that are used in this investigation were identified using a combined screening method (HTS and virtual screening).²²

The inhibitors studied in this work are 6-methoxy-2-(1H-pyrazol-5-yl)-1Hbenzimidazole (inhibitor 1), 4-dicarboxylicacid diamide (inhibitor 2), 4-butyl-6-(1H-pyrrolo[2,3-b]pyridin-3-yl)pyrimidin- 2-amine (inhibitor 3), 4-ethyl-6-[5-(1Hpyrazol-4-yl)-1H-pyrrolo[2,3-b]pyridin-3-yl]pyrimidin-2-amine (inhibitor 4)²² and 1-(3,4-difluorobenzyl)-2-oxo-N-{(1R)-2-[(2- oxo-2,3-dihydro-1H-benzimidazol-5yl)oxy]- 1-phenylethyl}-1,2-dihydropyridine-3-carboxamide (inhibitor 5) which are shown in figure 1.7.



Figure 1.7: The five PDK-1 inhibitors that used in this investigation²²

Inhibitor (5) is an example of pyridinonyl-based PDK-1 inhibitors described by Sunesis and Biogen Idec.²⁶ The concept of these inhibitors is mainly dependent on

the presence of a flexible linker to a hinge binding moiety (HBM) bearing neighboring H bond donor (HD) and H bond acceptor (HA) groups.²⁶

1.8 Classification of inhibitors

Protein kinase inhibitors are classified according to their binding modes as follows:

Type I inhibitors: which are classical ATP-competitive and bind the ATPbinding site when the protein kinase is an activated state. They bind the hinge region with at least one hydrogen bond. FDA recently approved Type I anticancer kinase inhibitors: gefitinib, dasatinib, sunitinib, lapatinib, ruxolitinib, pazopanib, vemurafenib, crizotinib, erlotinib, and bosutinib.²⁷

Type II inhibitors: are also ATP-competitive with binding to the extended ATP-binding site of protein kinase in an inactive state. If a significant change in the protein conformation occurs, it means that the inhibitor belongs to type II. Conformational changes in the protein kinase structure open a new hydrophobic pocket in the back of the protein that is called the *Deep Pocket* (also called the Phe pocket or allosteric pocket). These inhibitors, usually are hydrogen bonded to the hinge region but this is not a requirement for their action.²⁸

Type III inhibitors (Allosteric binders): are ligands that target allosteric binding sites of protein kinase, therefore they are non-ATP-competitive. As

allosteric binding sites are highly specific for a protein kinase, this means a high degree of selectivity can be achieved. A specific feature that characterized this type of inhibitors no hydrogen bond in the hinge region. At present no drug on the market belongs to this type.²⁸ Traxler has developed a pharmacophore model for ATP-competitive inhibitors (type I) that identifies five different regions within the ATP-binding site.²⁸

1.9 Drug design and drug properties

The drug-likeness of oral small molecules were evaluated by several guidelines. Up to 2015, a total of 28 small molecule kinase inhibitors SMKIs are FDA approved for the treatment of human cancer. Lipinski's Rule of Five (ROF) has been used as a rule of thumb to evaluate their absorption, permeability and solubility of drugs. The Veber Rules (Number of rotatable bonds (NRB) \leq 10 and polar surface area (PSA) \leq 140 Å²) to evaluate oral bioavailability. Analysis of the number of rings (NOR) as included in the MDDR Rule (NOR \geq 3).²⁹

The physicochemical properties of more than 2,000 drugs and candidate drugs in clinical trials were analyzed by Lipinski: A compound has drug-like properties if it matches the following criteria (The Lipinski rule of five).³⁰

•Its molecular weight (M.W) < 500.

•The lipophilicity property of compound (logP) which is the logarithm of the partition coefficient between 1-octanol and water³¹ \leq 5.

• The number of atoms in the molecule that donate hydrogen atoms to form hydrogen bonds (OH & NH) ≤ 5

•The number of atoms that can accept hydrogen atoms to form hydrogen bonds

 $(O \& N) \le 10$

Poor absorption or permeability is possible when the compound properties not obey the rule of five.³²

Analysis of 28 FDA approved SMKIs revealed that 28 SMKIs were fitted well with ROF (HBD \leq 5 and HBA \leq 10) and the molecular weight of 28 SMKIs is in the range 400 to 600. It is worth noting that with the exception of exitinib all inhibitors have at least six HBAs this reflect the fact that nitrogen and oxygen atoms are beneficial for kinase inhibitors. The Veber Rule³³ are abided by all inhibitors except dabrafenib. The analysis of number of rings as included in MDDR Rule³⁴ showed good adherence, NOR was no more than five for all SMKIs.²⁹

1.10 Computational approach for binding free energy calculation using MM-GBSA (or MM-PBSA) Several computational methods are available for calculating the binding free energy of protein-protein interactions, protein-DNA interactions and ligand-protein interactions.³⁵ Some methods are more accurate but computationally intensive such as the thermodynamic integration (TI) and the free energy perturbation (FEP) methods.³⁶ On the other hand, less accurate methods such as molecular mechanics/Poisson Boltzmann surface area (MM/PBSA) and molecular mechanics/Generalized Born surface area (MM/GBSA) are less time-consuming methods.³⁷ MM/PBSA and MM/GBSA methods are end point methods because they calculate the binding affinity through simulations of only two end states (unbound and bound states of a ligand and its protein target).³⁷

1.10.1 Binding free energy of ligand-protein complex using MM-GBSA

In the MM-GBSA formulation, the binding free energy of a ligand to a protein is calculated as the difference between the free energy of protein-ligand complex and the sum of the free energies of protein and ligand separately as follows.³⁸

$$\Delta G^0$$
 binding, solvated = G^0 complex, solvated - [G^0 receptor, solvated + G^0 ligand, solvated] (1.1)
From the thermodynamic cycle shown in Figure 1.8. The binding free energy calculated as illustrated in equation (1.2).

$$\Delta G^{0}_{\text{binding, solvated}} = \Delta G^{0}_{\text{binding, vacuum}} + \Delta G^{0}_{\text{solv, complex}}$$
$$- (\Delta G^{0}_{\text{solv, receptor}} + \Delta G^{0}_{\text{solv, ligand}})$$
(1.2)



Figure 1.8: Thermodynamic cycles for binding free energy calculations for complex solvated systems (in blue boxes), whereas systems in the gas phase (in white boxes)

In the calculation of the solvation free energy term polar and nonpolar contributions are considered. For the polar contribution, the change in the free energy resulting from transfer of a charged molecule from gas-phase (modeled as a homogeneous medium with dielectric constant=1) to solvent (modeled as a homogeneous medium with=80), equation (1.4) γ and β values are constants dependent on the applied method.^{39,40}

$$\Delta G_{sol}^0 = \Delta G_{polar}^0 + \Delta G^0 \tag{1.3}$$

$$\Delta G_{sol(nonpolar)}^{0} = \gamma(SASA) + \beta \tag{1.4}$$

$$\Delta G_{sol(polar)}^{0} = G_{electrostatic,\varepsilon=80}^{0} - G_{electrostatic,\varepsilon=1}^{0}$$
(1.5)
The approximation formula of the electrostatic contribution appears in equation (1.6), but extended Generalized Born model consists of a set of radii (ai) and charges contributions from Equation 1.7 for each particle.⁴¹

$$\Delta G_{elec} = -\frac{q^2}{2a} \left(1 - \frac{1}{\varepsilon} \right) \tag{1.6}$$

$$\Delta G_{elec} = -\left(1 - \frac{1}{\epsilon}\right) \sum_{i=1}^{N} \sum_{i=i+1}^{N} \frac{q_i q_j}{r_{ij}} - \frac{1}{2} \left(1 - \frac{1}{\epsilon}\right) \sum_{i=1}^{N} \frac{q_i^2}{a_i}$$
(1.7)

$$\Delta G^{0}_{\text{bind, vacuum}} = \Delta E^{0}_{\text{MM}} - T.\Delta S^{0}_{\text{Nmode}}$$
(1.8)

$$E_{MM} = E_{bond} + E_{Angle} + E_{Torsion} + E_{Van der Waals} + E_{electrostatioc}$$
(1.9)

Protein -inhibitor average binding energy estimated by in the gas phase by molecular mechanics. Two types of energy are involved: first are non-covalent energies consisting of van der Waals energy and electrostatic energy. The second type are the covalent energies represented by bonds, angles and dihedral energies.³⁹

Another way to calculate binding affinity is by molecular mechanics-Poisson Boltzmann surface area(MM-PBSA). Both MM-GBSA and MM-PBSA use the same previous equations to calculate the binding free energy, but the difference in the calculation of the electrostatic solvation energy G_{sol} (polar contribution).⁴²

$$\Delta G_{\rm sol} = \Delta G_{\rm PB/GB} + \Delta G_{\rm SA} \tag{1.10}$$

Where ΔG_{sol} is the sum of electrostatic solvation energy (polar contribution), $\Delta G_{PB/GB}$, and the nonelectrostatic solvation component (nonpolar contribution), ΔG_{SA} .⁴² The electrostatic energy (ΔG_{PB}) is calculated by solving the Poisson-Boltzmann numerically. By combining Poisson's equation (1.11) for the electrostatic potential with Boltzmann's equation (1.12) that gives the charge distribution, you end up with the Poisson-Boltzmann equation (1.13).⁴¹

$$\nabla^2 \phi(\mathbf{r}) = -\frac{4\pi\rho(\mathbf{r})}{\epsilon} \tag{1.11}$$

$$n(r) = Ne^{-\frac{V(r)}{k_b T}} \tag{1.12}$$

$$\Delta G \ solv = \frac{1}{2} \sum_{i} qi \left(\emptyset i^{\varepsilon=80} - \emptyset i^{\varepsilon=1} \right)$$
(1.13)

Chapter 2

Computational Methods

2.1 Protein- Inhibitor structures

The crystal structure of PDK-1 complex with five inhibitors were taken from the Brookhaven Protein Data Bank. The PDB codes of the PDK-1 with inhibitors 6methoxy-2-(1H-pyrazol-5-yl)-1H-benzimidazole (inhibitor 1), 4-dicarboxylicacid diamide (inhibitor 2), 4-butyl-6-(1H-pyrrolo[2,3-b]pyridin-3-yl)pyrimidin- 2-amine (inhibitor3),4-ethyl-6-[5-(1H-pyrazol-4-yl)-1H-pyrrolo[2,3-b]pyridin-3yl]pyrimidin-2-amine (inhibitor 4), and 1-(3,4-difluorobenzyl)-2-oxo-N-{(1R)-2-[(2-oxo-2,3-dihydro-1H-benzimidazol-5-yl)oxy]-1-phenylethyl}-1,2dihydropyridine-3-carboxamide (inhibitor 5) are 5HNG, 5HO7, 5HO8, 5HKM, and 3NAX respectively. Water molecules and two sulfate groups were removed from the PDB files.

It was reported that part of N-terminal lobe of PDK-1 (residues 1-50) interact with Ralguanine nucleotide exchange factors. This region was not present in the PDB file of PDK-1 structure, because this region assumed a unique conformation in PDK-1.¹⁷ PDK-1 protein consists of 556 amino acids, the Phosphoserine residue (SEP) is in position 241 is linking L-peptide ($C_3 H_8 N O_6 P$). It was reported that 3-Phosphoinositide-dependent protein kinase-1 (PDK-1) expressed in 293 cells was phosphorylated at Ser25, Ser241, Ser393, Ser396 and Ser410. Mutation of Ser241 to Ala canceled PDK-1 activity, whereas mutation of the other phosphorylation sites individually to Ala did not aff ect PDK-1 activity. Also it was reported that PDK-1 can phosphorylate itself at Ser241, leading to its own activation.⁴³ The pdb files of PDK-1 structure in inhibitors (1-4) show a phosphorylated T-loop in Ser241 therefore, it is in an active state.¹⁷

2.2 Equilibration of the solvated system

Four steps were used to equilibrate the system: minimization, heating, density equilibration and unrestrained equilibration.

(a) Relaxation of the solvated system.

We used *sander* to minimize our system in order to remove any bad contacts as a results of the hydrogenation steps in *xLeap*. Minimization (imin=1) was done in two steps: The first step involves the relaxation of water molecules only, whereas protein and inhibitor atoms were fixed by using a harmonic restraint (restraint_wt=2.0). The second step involves the minimization of the whole system using Sander. The input file min.in was used to perform the first step and min_all.in was used to perform the second step of minimization (see Appendix B). Minimization was performed using 500 steps of the steepest descent method, then switching to conjugate gradient algorithm for the remaining steps (maxcyc=1000, ncyc=500). Constant volume periodicity was applied (ntb=1).

(b) Heating the solvated system.

The system was then heated (imin=0) using langevin thermostat (ntt=3) to maintain the temperature of our system to 300K, with a collision frequency 2 ps⁻¹. This method is more efficient than Berendsen method (ntt=1) due to hot solvent, cold solute phenomena.⁴⁴ The file titled heat.in (see Appendix B) was used to perform the heating process.

(c) Density equilibration.

The file titled density.in (see Appendix B) was used to perform this step. The system was equilibrated at 300 K with constant pressure periodic boundary (ntp=1) using Particle mesh Ewald (PME) method and positional restrains of 2 Kcal/mol. $Å^2$ was applied.

(d) Unrestrained equilibration

The file titled equil.in (see Appendix B) was used to perform this step. The unrestrained system was equilibrated at 300 K with constant pressure periodic boundary (ntp=1). The *SHAKE* method⁴⁵ was applied (ntc=2, ntf=2) to hold all covalent bonds containing hydrogen atoms.

2.3 Production step of the solvated system

The production simulation time is 2ns that run using the same conditions as in the final phase of equilibration to prevent any sudden jump in the potential energy due to a change in simulation conditions. The production run was carried out over four sequential steps using the input file prod.in (see Appendix B). During all the MD simulations, the Particle Mesh Ewald (PME) method was utilized with 10Å cutoff for long-range interactions.

2.4 Calculating the binding free energy of the protein-inhibitor complex

We carried out the binding free energy calculation using both the MM-GBSA method and the MM-PBSA method for comparison. This is achieved using input file for mmpbsa.in (see Appendix B).

2.5 Calculating the entropic contribution

Normal Mode Analysis (*Nmode*) was used to calculate the entropic contribution.⁴⁶ The file titled mmpbsa_nm.in (see Appendix B) was used to do this step.

Chapter 3

Results and discussion

3.1 Analysis of simulations

The system reached state of equilibrium after different stages of simulation. This was checked by monitoring of different properties during the simulation. The system properties were extracted from the output files, and were ploted versus time. Figures 3.1 to 3.4 show plots for inhibitor 1-protein complex.



Figure 3.1: Density of protein-inhibitor (1) complex system during equilibration

As shown in Figure 3.1, the first 50 ps of the simulation represents the heating stage. There was no density data recorded due to the constant volume condition that took place until 50 ps. After that the density increased up to 1.02 g/ml and stayed around this number until the last 550 ps. This is reasonable because the density of pure water at 300 K is 1.00 g/ml, so adding inhibitor 1-protein complex lead to a rise in density of the system by 4%.⁴⁷



Figure 3.2: Temperature of protein-inhibitor (1) complex system during equilibration runs

In Figure 3.2, the temperature rises regularly from 0 K to 300 K. After that the temperature of the system reached an equilibrium value of 300 K over the last

stage of simulations, indicating that Langenvin dynamics applied successfully in this case.⁴⁷



Figure 3.3: Pressure of protein-inhibitor (1) complex system during equilibration

In the pressure versus time plot (Figure 3.3). It shows that in the time interval between 0-50 ps the pressure was zero, because it was running at constant volume. At 50 ps the system changed to constant pressure, the volume of the box changed and the pressure dropped sharply becoming negative.⁴⁸ Positive values of pressure reflect a force trying to make the water box larger, whereas negative pressure values reflect a force trying to reduce the volume of the water box.⁴⁷ While the pressure plot shows that the pressure fluctuated during the simulation, pressure stabilized at1 atm, this indicates a successful equilibration.



Figure 3.4: Total, kinetic and potential energy of protein-inhibitor (1) complex system during equilibration runs. The kientic energy (in red line); the potential energy (in black line); the total energy (in green line) which is the sum of kinteic and potential energy

According to the energy plot versus time (Figure 3.4) the first 50 ps of simulation there was An increased in all energies corresponding to heating from 0 K to 300 K. The kinetic energy remained constant in the last stages indicating a successful performance of temperature thermostat.⁴⁷

The potential energy and the total energy initially increased, then during the constant volume stage (0 to 20 ps) there was a plateau, then at 20 to 40 ps there was a decrease in the energy values because, at this stage we switched off the protein-ligand restraints and moved to constant pressure. After that the potential energy

leveled off for the reminder of our simulation indicating stability and a relaxed system.⁴⁷



Figure 3.5: RMSD of protein backbone during unrestrained equilibration run of protein-inhibitor (1) complex

In order to quantify the similarity between a native inhibitor1-protein complex (com_wat.inpcrd) and a generated inhibitor1-protein complex (equil.mdcrd), the mass weighted RMSD (Root-mean square deviation) can be calculated between these two structures. Figure 3.5 shows that the root mean square deviation (RMSD) values increased rapidly in the first 75 ps, then it fluctuated around a value of 1.4 Å until the last 250 ps, which is an acceptable value. RMSD

values were around 1.4 Å, this reflects an acceptable conformational changes in the protein backbone .

3.2 Study of binding energies of kinase – inhibitors3.2.1 Binding free energies of protein-inhibitor complexes: MM-PBSA versus MM-GBSA

MM-PBSA and MM-GBSA are direct methods for the quantitative prediction of binding free energy of ligand-protein complex.⁴⁹

Both methods are used in this work to calculate the binding free energy of PDK-1 kinase with five inhibitors (Table 3.1). As shown in Figure 3.6 there is a good correlation between binding free energies were calculated by MM-GBSA, and experimental values of binding free energies which are derived from the experimental reported IC₅₀ values ($R^2 = 0.54$).



Figure 3.6: Correlation between ΔG calculated by MM-GBSA and ΔG experimental values

To the contrary, the correlation between the binding free energies were calculated by MM-PBSA and experimental values of binding affinity (Fig 3.7) which are derived from the experimental reported IC₅₀ values is weaker ($R^2 = 0.06$).



Figure 3.7: Correlation between ΔG calculated by MM-PBSA and ΔG experimental values

It is observed that the calculated binding free energies using MM-GBSA method were closer to the experimental values than those calculated using MM-PBSA method. In principle, PB is more theoretically rigorous than GB, but it does not mean that MM/PBSA can give better predictions than MM/GBSA.⁵⁰ Our result agrees with some of the reports that MM-GBSA based on GB^{OBCI} is considered a better approach than the MM-PBSA in calculating the binding free energies when heterocyclic and aromatic system is present.⁵¹

	Experimental		Calculated	
Inhibitor	IC 50 (μM)	$\Delta \mathbf{G_{exp}}^{22}$ (Kcal/mole)	$\begin{array}{c} \Delta G_{calc} \\ \text{(MM-GBSA)} \\ \text{(Kcal/mole)} \end{array}$	$\begin{array}{c} \Delta G_{calc} \\ \text{(MM-PBSA)} \\ \text{(Kcal/mole)} \end{array}$
62O	93	-5.5	0.3 ± 1.6	26.0 ± 1.7
$63L \qquad \overset{CH_3}{\overset{CH_3}{1}}$	17	-6.5	-1.6 ± 2.0	18.1 ± 2.0
MOL H ₃ C	1.1	-8.1	-21.0 ± 1.5	23.5 ± 2.3
HNNN	0.012	10.0	15.0 1.0	25.2.2.6
61Y	0.013	-10.8	-15.0 ± 1.8	25.2 ± 2.6
	-	-	-52.3 ± 2.8	-11.4 ± 3.0

Table 3.1: Binding free energies (kcal/mol) calculated at T= 300 K and P= 1 atm for PDK-1 binding with the four inhibitors

The experimental binding free energies (ΔG_{bind}) were calculated from the experimental values of IC₅₀, by using this equation:

$\Delta \mathbf{G}_{\text{bind}} = \mathbf{RTln} \mathbf{K}_{\mathbf{D}} = \mathbf{RTlnIC}_{50} \tag{3.1}$

The reported IC₅₀ values are concentrations at which the PDK-1 kinase activity is inhibited by 50% of the initial concentration.⁵² The kinetic study of enzyme-inhibitor reaction in the absence of inhibitor follows a simple Michaelis-Menten equation (3.2).⁵³ The following equation assumes that the concentration of enzyme is sufficiently low (neglected).

$$V0 = \frac{V\max S}{Km + S} \tag{3.2}$$

$$VI = \frac{Vmax \ S}{Km \left(1 + \frac{I}{KI}\right) + S}$$
(3.3)

Where, V_{max} = maximum velocity; V_0 = velocity in the absence of the inhibitor; K_m =Michaelis constant of the substrate; V_I = velocity in the presence of inhibitor; I = concentration of inhibitor; S= substrate concentration; K_I = dissociation constant of enzyme-inhibitor complex (EI).

When $I = I_{50}$, $V_0 = 2V_I$ then⁵³

$$\frac{2 Vmax S}{Km \left(1 + \frac{I50}{KI}\right) + S} = \frac{Vmax S}{Km + S}$$
(3.4)

By rearranging equation 3.4:

$$I50 = KI \ (1 + \frac{S}{Km}) \tag{3.5}$$

In the case of a competitive inhibitor, S << Km, then Ki ~ IC_{50}.

3. 3 Analysis of the binding mode of inhibitor (1)-PDK-1 complex

In this section we discussed the binding mode of inhibitor (1) with PDK-1 kinase complex (Figure 3.8(a) and (b)). The non-covalent interaction of inhibitor to the proteins is governed by different interactions including van der Waal and hydrogen bond interaction.²⁴

Inhibitor (1) makes two strong hydrogen bond interactions with the backbone oxygen atom of Ser⁹⁰ in the adenine region of the kinase with a distance of 2.1 Å and 1.8 Å (Table 3.2). The benzimidazole ring is in a buried region. It is surrounded by residue Thr¹⁵² (4.6 Å not considered as hydrogen bond).

Other weak interactions were formed between the inhibitor (1) and the PDK-1 (Figure 3.8(c)), C-H π interaction⁵⁴ between carbon hydrogen atom of Leu¹⁸ and the center of benzimidazole ring of inhibitor (3.1 Å is the average distance from the hydrogen atom to the center of ring). The same type of interaction was made by Ala⁹² in the hinge region, but the average distance is 3.9 Å.



Figure 3.8: (a) Inhibitor (1)-PDK-1 complex, (b) graphical representation of inhibitor (1) and (c) other weak interactions between inhibitor (1) and kinase



 Table 3.2: Hydrogen bond analysis of inhibitor (1) in 62O-PDK-1complex

3.4 Analysis of the binding mode of inhibitor (2)-PDK-1 complex

In this section we discussed the binding mode of inhibitor (2) with PDK-1 kinase complex (Figure 3.9(a) and (b)). Inhibitor (2) has hydrogen bond interaction with the backbone carbonyl group of Ser⁹⁰ with 2.1 Å distance, and another two hydrogen bonds are formed with Ala⁹² at 3.2 Å and 2.1 Å distances in the hydrophobic adenine pocket (Table 3.3).

A strong hydrogen bond interaction with the carbonyl group of Thr¹⁵² in the buried region (length = 1.8 Å). It is worth noting that this inhibitor has an intramolecular hydrogen bond between hydrogen atom of amino group (HO2) and oxygen of carbonyl group (O14) as shown in table 3.3.

The highest frequency of intramolecular hydrogen bonds for planer, six membered rings stabilized by conjugation with a π -system. The formation of an intramolecular hydrogen bond result in an increased lipophilicity and membrane permeability accompanied by reduced aqueous solubility. These are due to the removal of one donor and one acceptor function from the surface of a molecule.⁵⁵

Replacing real rings by such pseudo rings to form pseudo six-membered ring is a new and non-conventional strategy and the new classes of kinase inhibitors follow this approach.⁵⁶ We noted that inhibitor (2) interacts with the active site in water-mediated hydrogen bonds with active-site residues. A water-mediated network of hydrogen bonds is formed by 2 water molecules to inhibitor (2) as shown in table 3.3.

In addition to all of these interactions, other weak interactions were formed between inhibitor (2) and PDK-1 (Figure 3.11(c)), C-H.....C=O interaction⁵⁴ between hydrogen atom of Tyr⁹¹ and the carbonyl group of inhibitor (2.6 Å).

Another weak interaction was formed of the type C-H..... π interaction⁵⁴ between carbon hydrogen atom of Val²⁶ with the center of pyrazole ring (3.7 Å is average distance between the center of the pyrazole ring and the hydrogen atom) as shown in figure 3.9(c).

It is worth noting that Leu¹⁸ is close to inhibitor due to the C-H.....C=O weak interaction between hydrogen atom of Leu¹⁸ and the carbonyl group of Val²⁶ (2.1 Å) as shown in figure 3.9(c).



Figure 3.9: (a) Inhibitor (2)-PDK-1 complex, (b) graphical representation of inhibitor (2) and (c) other weak interactions between inhibitor (2) and kinase



 Table 3.3: Hydrogen bond analysis of inhibitor (2) in 63L-PDK-1complex

Atom of	Atom of	Comment
inhibitor	protein/H ₂ O	
O14	H-63L ²⁸² (Intramolecular hydrogen bond)	
HO5	O-Asp ¹⁵³ (Water mediated)	Water Molecules Molecules

Table 3.3: Hydrogen bond analysis of inhibitor (2) in 63L-PDK-1complex



 Table 3.3: Hydrogen bond analysis of inhibitor (2) in 63L-PDK-1complex

3.5 Analysis of the binding mode of inhibitor (3)-PDK-1 complex

In this section we discussed the binding mode of inhibitor (3) with PDK-1 kinase complex (Figure 3.10(a) and (b)). Inhibitor (3) is located in the ATP-binding site (which lies between the N-terminal and C-terminal lobes of kinases).⁵⁷ The 7-azaindole ring mimics the interactions of the adenine base in ATP with the protein backbone, where two conserved hydrogen bonds are formed between the 7-azaindole nitrogen N3 in inhibitor (3) and the backbone-hydrogen of Ala⁹², and the 7-azaindole hydrogen HO5 and the backbone-oxygen of Ser⁹⁰ as shown in table 3.4.

In addition to the presence of direct hydrogen bonds, there are water-mediated hydrogen bond interactions. The water mediated hydrogen bond interaction occur between inhibitor (3) and Lys⁴¹ in the phosphate region, and Thr¹⁵² in the buried region as shown in table 3.4.

It was reported that the discovery of aminoindazole ring and the addition of one heterocyclic ring which is involved by using its nitrogen atoms in the hydrogen bond interaction with inhibitor. This is critical for binding. Overall cumulative data confirm that each nitrogen in the aminoindazole positively contributes to PDK-1 binding and inhibition activity.⁵⁸

Other weak interactions were formed between inhibitor (3) and the PDK-1,

C-H.... π interaction⁵⁴ between carbon hydrogen atom of Leu¹⁸ and the center of 7azaindole ring (3.3 Å is average distance between the hydrogen atom and the center of the ring) as shown in figure 3.10(c).





Figure 3.10: (a) Inhibitor (3)-PDK-1 complex, (b) graphical representation of inhibitor (3) and (c) other weak interactions between inhibitor (3) and kinase

Atom of	Atom of	Comment
inhibitor	protein/H ₂ O	
N16	H-Lys ⁴¹	
	(Water	N3 Water Water
	mediated)	c2 c6 c8 molecule
N16	H-Lys ⁴¹	H02 H02 H01 H01 H01 H01 H02 H02 H02 H02 H02 H02 H02 H02 H02 H02
		LYS 41
N15	H-Thr ¹⁵²	THR 152 Hos Water molecule N7 H08
HO8	O-Thr ¹⁵²	N3 C6 ME 152 2.0
	(Water mediated)	H03 C2 C5 N15 N16 C1 C4 C1 C1 V13 H02 H04 C12

 Table 3.4: Hydrogen bond analysis of inhibitor (3) in MOL-PDK-1complex

Atom of	Atom of	Comment
inhibitor	protein/H ₂ O	
HO5	O-Ser ⁹⁰	SER 90
N3	H-Ala ⁹²	ALA 92

Table 3.4: Hydrogen bond analysis of inhibitor (3) in MOL-PDK-1complex

3.6 Analysis of the binding mode of inhibitor (4)-PDK-1 complex

In this section we discussed the binding mode of inhibitor (4) with PDK-1 kinase complex (Figure 3.11(a) and (b)). Inhibitor (4) forms four hydrogen bonds with the protein. The N12 atom binds the hydroxyl group hydrogen of Thr¹⁴⁷ in the buried region, N4 atom binds hydroxyl group hydrogen of Ala⁸⁷ in the adenine region, N23 atom binds the hydrogen atom of Lys⁹⁴, N23 atom binds hydroxyl group hydrogen atom of Glu⁹¹ in the sugar region and HO5 hydrogen atom binds to the backbone carbonyl group of Ser⁸⁵ in the adenine region (Table 3.5).

Other weak interaction was formed between inhibitor (4) and the PDK-1, C-H.... π interaction⁵⁴ between carbon hydrogen atom of Leu¹³ and the center of 7azaindole ring as shown in figure 3.11(c).





Figure 3.11: (a) Inhibitor (4)-PDK-1 complex, (b) graphical representation of inhibitor (4) and (c) other weak interactions between inhibitor (4) and kinase

It was reported that **type I kinase inhibitors** form water-mediated hydrogen bond networks (both water molecules W1 and W2 are commonly observed) and the ligand does not extend to the water-filled cavity. These two features distinguish type I from type II inhibitors.⁵⁹

Figure 3.12 illustrates the typical distribution of ligand-W1 hydrogen bond distances for 180 ATP-binding site ligands.⁵⁹ According to inhibitor (4), the ligand-W1 hydrogen bonds was 3.0 Å which is agree with the typical distribution of hydrogen bonds.



Figure 3.12: The distribution of hydrogen bond lengths for 180 ATP-binding site ligand-W1 hydrogen bonds⁵⁹



 Table 3.5: Hydrogen bond analysis of inhibitor (4) in 61Y-PDK-1complex

Atom of	Atom of	Comment
inhibitor	protein/H ₂ O	
N23	H-Lys ⁹⁴	LYS 94
N23	H-Lys ⁹⁴	
N23	H-Lys ⁹⁴	
N17	H-Lys ³⁶	
	(Water mediated)	LYS 36 2.8 HOS HOS NI7 HIB 3.0
		C ¹ Cl1 Nl4 C ¹ Cl5 Water molecule

 Table 3.5: Hydrogen bond analysis of inhibitor (4) in 61Y-PDK-1complex

Atom of	Atom of	Comment
inhibitor	protein/H ₂ O	
HO5	O-Ser ⁸⁵	ER-26 HG N17 FR-26 C13 N14 1.9 HG N12 N14 1.9 HG N8 C10 C16 C12 SER 85 N4 O O O O O O HG N8 C10 C16 C12 HG HG
HO8	O-Asp ¹⁴⁸ (Water mediated)	ASP 148

 Table 3.5: Hydrogen bond analysis of inhibitor (4) in 61Y-PDK-1complex

Atom of	Atom of	Comment
inhibitor	protein/H ₂ O	
N23	H-Glu ⁹¹	GLU 91

 Table 3.5: Hydrogen bond analysis of inhibitor (4) in 61Y-PDK-1complex

Dunitz et. al⁶⁰ reported that the entropic gain of releasing a bound water molecule from the binding site of protein can be 7 cal/mol. K, corresponding to an energy gain of 2.1 kcal/mol at 300 K.⁶⁰ The entropic contribution to binding affinity is observed upon removing water molecules from the binding sites of protein molecules, and is an essential part which cannot be ignored in drug design.⁶¹

The position of water molecules in the binding sites can be used to design better inhibitors in which the principle lies in the fact that a substituent is added to the inhibitor that replaced a water molecule that bounded to kinase (design inhibitor that includes a structural water mimic).⁶²

An increase in ligand affinity can result if the contribution of substituent is greater than free energy cost which results from displacing solvent molecules. This is an easy process because the ligand already has paid the energy cost as translational and rotational entropy.⁶¹
3.7 Analysis of the binding mode of inhibitor (5)-PDK-1 complex

In this section we discussed the binding mode of inhibitor (5) with PDK-1 kinase complex (Figure 3.13(a) and (b)). Inhibitor (5) occupied allosteric site of the protein kinase PDK-1 called the PDK1 -interacting fragment (PIF)tide-binding site, or PIF pocket. This inhibitor was occupied PIF/Phosphate pocket which was determined by Lys115, Ile118, Ile119, Val124, Leu155 residues.

Inhibitor (5) binds to the inactive kinase conformation (DFG-out) in the PIF/Phosphate pocket of PDK-1 kinase, so this inhibitor considered as Type II (Deep pocket binder) inhibitor. It is worth noting that this is the first reported example of Type II (DFG-out) kinase inhibitor for AGC kinase.⁶³

Inhibitor (5) consists three molecular fragments: a hinge binding group, a linker, and a hydrophobic moiety.⁶³ The hydrophobic moiety interacts with the phosphate pocket through four strong hydrogen bond interactions. Three hydrogen bonds are formed between O34 atom and hydrogen atoms (HZ1, HZ2, and HZ3) of the amino group of Lys³⁶. The fourth strong hydrogen bond is formed between the carbonyl group of inhibitor and the amino group hydrogen atom of Asp¹⁴⁸ as shown in table 3.6.

This inhibitor interacts with the hinge region through Ser⁸⁵, and Ala⁸⁷. The first interaction is between H22 amino group hydrogen atom of inhibitor and the

carbonyl group of Ser⁸⁵. The second interaction is between carbonyl group of inhibitor and amino group hydrogen atom of backbone Ala⁸⁷ (Table 3.6).

It is worth noting that a strong intramolecular hydrogen bond interaction is present. This intramolecular interaction is between the carbonyl group and the amino group hydrogen atom (HO1). This type of interaction is like inhibitor (2) between hydrogen atom of amino group (HO2) and oxygen of carbonyl group (O14) as shown in table 3.6.

As previously stated: replacing real rings by pseudo rings to form pseudo sixmembered ring is a new and non-conventional strategy and the new classes of kinase inhibitors follow this approach.⁵⁶

In addition to the strong interactions, other weak interactions were formed between inhibitor (5) and the PDK-1 (Figure 3.13(c)), C-H....C=O interaction⁵⁴ between the carbonyl group of Phe¹⁴⁹ and the carbon hydrogen atom of inhibitor (3.0 Å distance), between the carbon hydrogen atom of Leu⁸⁴ and the carbonyl group of inhibitor (3.0 Å distance) and between the carbon hydrogen atom of Tyr⁸⁶ and the carbonyl group of inhibitor (3.0 Å distance) as shown in figure 3.14(c).





Figure 3.13: (a) Inhibitor (5)-PDK-1 complex, (b) graphical representation of inhibitor (5) and (c) other weak interactions between inhibitor (5) and kinase

Atom of	Atom of	Comment
inhibitor	protein/H ₂ O	
0	H-MP7 ²⁷⁹ (Intramolecular hydrogen bond)	

Table 3.6: Hydrogen bond analysis of inhibitor (5) in MP7-PDK-1complex

Atom of	Atom of	Comment
inhibitor	protein/H ₂ O	
H22	O-Ser ⁸⁵	

Table 3.6: Hydrogen bond analysis of inhibitor (5) in MP7-PDK-1complex

Atom of	Atom of	Comment
inhibitor	protein/H ₂ O	
O31	H-Ala ⁸⁷	ALA 87

 Table 3.6: Hydrogen bond analysis of inhibitor (5) in MP7-PDK-1complex

Atom of	Atom of	Comment
inhibitor	protein/H ₂ O	
0	H-Asp ¹⁴⁸	ASP 148

Table 3.6: Hydrogen bond analysis of inhibitor (5) in MP7-PDK-1complex

Atom of inhibitor	Atom of protein/H2O	Comment
0	H-Lys ³⁶	Lys 36

A significant difference between the classical ATP-competitive inhibitors and Type II (Deep pocket binder) inhibitors were firstly, α C-helix of PDK-1 kinase was distorted. The conformational change in this helix is due to displacing of Glu-130 residue from the active site⁶⁰, this disruption was observed in our study.⁶⁴ Secondly, there was a hydrogen bond interaction between Arg¹³¹ with PS⁴⁸ in the classical ATP-competitive inhibitors, but this is replaced by Arg¹³¹ with Glu¹³⁰ in the inactive conformation.⁶⁴ This is consistent with our results.

It was reported that the distortion of DFG motif is the most obvious in Type II (Deep pocket binder) inhibitor. Rotation about ϕ main chain torsion angle of Asp³⁸¹, as a result of this rotation, Phe³⁸² removed from ATP pocket and Asp³⁸¹ becomes to the back pocket⁶⁵.

This conformational change creates an inactive state of the kinase because the flipped-out phenylalanine blocks ATP-binding site.⁶⁵ This distortion was not observed in our study because Asp³⁸¹ and Phe³⁸² residues were not present in the original PDB files of inhibitor-protein complex.

3.8 Effect of Thermodynamic parameters on the proteininhibitor complexes

The unfavorable interactions between studied inhibitors (1-5) and PDK-1 is reflected in the negative value of entropic contribution (T Δ S). This is due to the release of the ordered H₂O molecules in addition to the conformational change, which is typically negative as the association of a ligand with its target results in the loss of conformational freedom for one or both molecules.⁶⁶

In other words, the negative entropic contribution resulted from "freezing out" of translational, rotational and internal degrees of freedom of the ligand on binding.⁶⁷As illustrates in figure 3.14, the coefficient of determination value is R^2 =0.12. This reflects that entropic contribution is not a driving force of binding affinity. Whereas, as appears in figure 3.15. The difference of enthalpy is considered as a driving force of binding free energy, due to the high value of coefficient of determination ($R^2 = 0.98$).



Figure 3.14: Relationship between the difference of Gibbs free energy (ΔG) and the entropic contribution (T. ΔS)



Figure 3.15: Relationship between the difference of Gibbs free energy (Δ G) and the difference of enthalpy (Δ H)

Positive contributions from both enthalpy and entropy are requirements for high affinity of binding. These two aspects of drug association should be optimized in a challenging and perplexing process because enthalpy optimization can frequently be offset by a loss in entropy. Maximizing the enthalpy contribution is difficult due to the formation of favorable H-bonds and van der Waal contacts and this is opposed by the cost of desolvation of incorrectly positioned polar moieties within a molecule.⁶⁶

The previous statement is clearly observed when we try to maximize the number of hydrogen bond interactions in inhibitor (4)-protein complex, conversely, we faced positive contribution of entropy.

The binding free energy of a ligand for its target is a function of enthalpic and entropic contributions as defined by the Gibbs free energy change. This can be parsed into individual contributions of intermolecular van der Waal attractive forces, H-bonding interactions, and repulsive forces like the hydrophobic effect that drive a ligand out of water and into the hydrophobic cavity of a protein.⁶⁶

The negative binding free energy (ΔG_{bind}) of all complexes reflects the favorable interaction between inhibitor-protein complexes in pure water except in the first complex where it gave positive value.

Table 3.7: Thermodynamic parameters	of the five prot	tein-inhibitor	complexes t	that
calculated at T= 300 K and P=	1 atm			

	Expe	erimental	Calculated					
Inhibitor	IC50 (µM)	Δ G exp (Kcal/mole)	ΔΕ _{ΜΜ}	$\Delta \mathbf{G}_{sol}$	$\Delta \mathbf{G}_{calc}$ (MM-GBSA) (Kcal/mole)	TΔS (Kcal/mole)	Δ H (Kcal/mole)	
	93	-5.5	-35.9	21.0	0.3 ± 1.6	-15.2 ± 3.2	-14.9	
63L SCH ₃ NH ₂ NH ₂	17	-6.5	-56.7	36.4	-1.6 ± 2.0	-18.7 ± 2.6	-20.3	
MOL	1.1	-8.1	-65.8	32.2	-21.0 ± 1.5	-12.6 ± 7.2	-33.6	
61Y	0.013	-10.8	-58.7	23.1	-15.0 ± 1.8	-20.6 ± 3.6	-35.6	
	-	_	-132.0	59.1	-52.3 ± 2.8	□-20.6± 3.6	-72.9	

3.9 Correlation between IC₅₀ and binding free energy

The acceptable IC₅₀ value for a ligand to possess a drug-like property is (1-10) nM. According to this parameter which is most critical in determining the drug candidate, we conclude that inhibitor (1), inhibitor (2), inhibitor (3), and inhibitor (4) do not possess a drug-like property (IC₅₀ = 93 μ M, 17 μ M, 1.1 μ M, 0.013 μ M), respectively, because they dissociate and do not stay bound to the enzyme.

These values are correlated with binding free energies of these complexes that calculated by MM-GBSA which are 0.3 K cal/mol, -1.6 Kcal/mol, -21.0 Kcal/mol, -15.0 Kcal/mol. We noted that IC₅₀ becomes lower, binding free energy become larger in negative sign, and become more druggable property.

IC₅₀ of Inhibitor (1) is equals 93 μ M, this inhibitor does not possess drug-like property, and this result agree with what we computed, ΔG_{bind} of this inhibitor with PDK-1 kinase is 0.3 Kcal/mol. The positive sign of this value resulted from the formation of only one hydrogen bond interaction.

But if we look about inhibitor (5), we noted that this inhibitor has high value of ΔG_{bind} equals -52.3 Kcal/mol, this high negative value resulted from the formation of six strong hydrogen bond interactions. with no water mediated was present.

But if we compared inhibitor (3) and inhibitor (4) we see the IC₅₀ values of inhibitor (3) and inhibitor (4) is 1.1 μ M and 0.013 μ M respectively. But the binding affinities of these inhibitors are -21.0 Kcal/mol and -15.0 Kcal/mol respectively. This is explained by the fact that the stronger the hydrogen bonds formed between the water molecule and the binding site, more favorable enthalpic contribution occur, and at the same time water molecules become less disordered and more highly restricted (less favorable entropic contribution).⁶⁵

In inhibitor (4)-protein complex there were three water molecules instated in inhibitor (3)-protein complex there were only two water molecules.

As appear in figure 3.16, the IC_{50} values are well correlated with the binding affinities that calculated by MM-GBSA. The coefficient of determination in this case equals 0.55.



Figure 3.16: Correlation between the IC₅₀ values and binding free energies that calculated by MM-GBSA

Whereas, the correlation between the IC_{50} and the binding free energy that calculated by MM-PBSA, is weaker (Figure 3.17). The coefficient of determination in this case equals 0.14.





It is worth noting, that the coefficient of determination is the same with correlation between IC_{50} values and the experimental binding free energies equals 0.55 (Figure 3.18).



Figure 3.18: Correlation between the IC₅₀ values and the experimental binding free energies

3.10 Energies calculated by MM-GBSA and contributing energies

All inhibitors studied in this work except inhibitor (3) share the most prominent binding contributions from the van der Waals (VDW) interactions. VDW values were -42.8 Kcal/mol, -30.7 Kcal/mol, -40.3 Kcal/mol, -76.4 Kcal/mol for inhibitor (1)-kinase complex, inhibitor (2)-kinase complex, inhibitor (4)-kinase complex, and inhibitor (5)-protein complex, respectively (Table 3.8, Figure 3.19).



Figure 3.19: Contributions of electrostatic energy and van der Waals energies to the difference of Gibbs free energy (ΔG) of different inhibitor-PDK-1 kinase complexes

Table 3.8: Total binding free energy (Δ Gtotal), van der Waals energy (VDW), electrostatic energy (ELE),solvation free energy (Δ Gsol), and binding free energy of inhibitor-protein complex (Δ Gbind) that

calculated by MM-GBSA. All energies are in unit kcal/mol

Inhibitor	$\Delta \mathbf{G}_{\text{total}}$	VDW	ELE	$\Delta \mathbf{G}_{\mathbf{sol}}$	∆G _{bind} (MM-GBSA)
	-15.0	-30.7	-5.3	21.0	0.3 ± 1.6
63L	-20.3	-25.7	-31.0	36.4	-1.6 ± 2.0
MOL H ₃ C	-33.6	-40.3	-25.5	32.2	-21.0 ± 1.5
61Y H ₃ C NH ₂ NH ₂	-35.6	-42.8	-15.9	23.1	-15.0 ± 1.8
MP7	-72.9	-76.4	-55.6	59.1	-52.3 ± 2.8

The binding free energy values are well correlated with van der Waals Energy calculated by molecular mechanics (VDW). The coefficient of determination value in this case equals 0.96 (Figure 3.20).



Figure 3.20: Relationship between the difference of Gibbs free energy (ΔG) and the Van der Waals energy

To the contrary, the correlation between the electrostatic energy and the binding free energy, is weaker (Figure 3.21). The coefficient of determination in this



Figure 3.231: Relationship between the difference of Gibbs free energy (ΔG) and the electrostatic energy

3.11 Classification of inhibitors studied

The structure of all inhibitor-protein complexes except inhibitor (5) show a phosphorylated T-loop and are therefore, assumed to be in an active state.¹⁷According to Traxler model, inhibitors (1-4) bind to the activated (phosphorylated) form of the protein kinase and occupy the ATP binding site with at least a formation of one hydrogen bond with the hinge region. This reflects the fact that these inhibitors belong to traditional pharmacophore model (type I) or classical ATP-competitive inhibitor.⁶⁸

It is worth noting that these inhibitors are reversible ATP-competitive because the type of interaction is mostly hydrogen bonding interaction and no irreversible covalent bond formation.⁶⁹

Inhibitor (5) binds to the inactive kinase conformation (DFG-out) in the PIF/Phosphate pocket of PDK-1 kinase, so this inhibitor is considered as Type II (Deep pocket binder) inhibitor. It is worth noting that this is the first reported example of Type II (DFG-out) kinase inhibitor for AGC kinase.⁶³

Another evidence proved that inhibitor (5) is considered as deep-pocket binder inhibitor Type (II) is the absence of water-mediated hydrogen bond interactions. The presence of water molecules in the binding sites is consider a feature that distinguish Type (I) from Type (II) inhibitors.

3.12 Analysis of the inhibitors according to Lipinski's Rule of five, Veber Rule and MDDR Rule

The fragment -based approach to calculate the polar surface area (PSA) descriptor, is a free software package. In this study, *Molinspirstion* was used to calculate PSA (with other useful molecular descriptors). *SMILES* files are required to process the values.⁷⁰

All inhibitors in this study agreed with the Lipinski's rule of five (ROF), except for inhibitor (2). The number of atoms that donate hydrogen atoms to form hydrogen bonds (HBD) was 6 which is higher than the acceptable value (Fig 3.22).

According to Veber's and MDDR Rules, all values for inhibitors in this study are consistent with these Rules except for inhibitor (2) which has NOR value of one (Fig 3.22). This value is lower than the acceptable value (NOR=3).





Inhibitor number

Table 3.9: Analysis of the inhibitors according to Lipinski's Rule of five, Veber and MDDR Rules

	Inhibitor	Molecular	A ⁷⁰	PSA ⁷⁰	Num-H	Num-H	Num-	Number	Ligand
		weight	logP	(Å ²)	acceptor	donar	of	of	efficiency
		(g/mol)			Atoms	Atoms	rings	Rotatable	
								Bond	
	620								
		214.23	1.84	66.60	5	2	3	2	0.33
	63L CH ₃ O NH ₂ NH ₂	215.24	-1.26	130.04	7	6	1	2	0.45
	MOL	267.34	2.81	80.49	5	3	3	4	0.45
1	61	305.35	2.00	109.17	7	4	4	3	0.47
0 =		516.50	3.70	108.99	8	3	5	8	0.32

In addition to the previous parameters that determine the drug-like properties, the ligand efficiency is an essential common metric to assess the drug-like quality of a compound⁷¹. This is estimated by relating binding free energy to the number of heavy atoms in a molecule (LE= $-\Delta G/ HA$).⁷² Therefore, the resulting ligand efficiency tends to be maximal for small molecules (e.g. fragments) and then steadily decreases as heavier atoms are added. The LE value for a small molecule that inhibit protein-protein interaction is a round 0.24, whereas LE is equal to 0.3 or higher is a desired value.

It is observed that all ligand efficiency values of our inhibitors were higher than 0.3. So we can conclude that all our inhibitors except inhibitor (2) have the drug-like properties when appling Lipinski's Rule of five, Veber Rule and MDDR Rule.

There is a good correlation between the difference of Gibbs free energy (ΔG) calculated by MM-GBSA and the molecular weight, the value of correlation coefficient of determination is R² = 0.94 (Figure 3.23).



Figure 3.23: Relationship between the difference of Gibbs free energy (Δ G) calculated by MM-GBSA and the molecular weight

Also a good correlation between the difference of Gibbs free energy (Δ G) calculated by MM-GBSA and the lipophilicity property of inhibitor (logP), the value of coefficient of determination is R² = 0.54 (Figure 3.24).



Figure 3.24: Relationship between the difference of Gibbs free energy (Δ G) calculated by MM-GBSA and log P

To the contrary, the correlation between the binding free energy (ΔG) calculated by MM-GBSA and the polar surace area (PSA), is weaker (Figure 3.25).



The coefficient of determination in this case equals 0.02.

Figure 3.25: Relationship between the difference of Gibbs free energy (Δ G) calculated by MM-GBSA and PSA

3.13 Potency and selectivity of Inhibitor (5)

Inhibitor (5) made six strong hydrogen bonds with the PIF/Phosphate pocket of PDK-1 kinase with no water mediated hydrogen bond interaction; this interaction reflects the high affinity ($\Delta G_{bind} = -52.3$ Kcal/mol), which indicates high potency of this inhibitor.

The high sequence similarity in the ATP binding pocket between different kinases is a major challenge for developing inhibitors that are specific for one or a small number of kinases.

Inhibitor (5) is consider as type II inhibitors, which occupied PIF/phosphate pocket of PDK-1 kinase. This allosteric site in protein kinase are especially used for developing more selective inhibitors⁷³. This offers a possibility to develop more compounds with higher selectivity more than in the case of classical ATP-competitive inhibitors.⁶⁴

Deep-pocket binder molecules, when used as single substrate, can be classified as substrate-selective PDK-1 inhibitors. When used in combination with ATP-competitive inhibitors they tend to suppress the activation of the downstream kinases.⁷³

4. CONCLUSION

Molecular dynamics simulation was employed to identify an exquisitely potent PDK-1 inhibitor 5 (1-(3,4-difluorobenzyl)-2-oxo-N-{(1R)-2-[(2- oxo-2,3 dihydro-1H-benzimidazol-5-yl)oxy]-1-phenylethyl}-1,2-dihydropyridine carboxamide) that uniquely binds to the inactive kinase conformation.

This inhibitor is tightly bound to PDK-1 through five strong hydrogen bonds with the PIF/Phosphate pocket of PDK-1 kinase with no water mediated hydrogen bond interactions. This interaction reflects the high affinity of drug to receptor $(\Delta G_{\text{bind}} = -52.3 \text{ Kcal/mol}).$

In contrast to compounds 1-4, which are classical ATP-competitive kinase inhibitors (DFG-in) which are 6-methoxy-2-(1H-pyrazol-5-yl)-1H-benzimidazole (inhibitor 1), 4-dicarboxylicacid diamide (inhibitor 2), 4-butyl-6-(1H-pyrrolo[2,3b]pyridin-3-yl)pyrimidin- 2-amine (inhibitor 3), 4-ethyl-6-[5-(1H-pyrazol-4-yl)-1Hpyrrolo[2,3-b]pyridin-3-yl]pyrimidin-2-amine (inhibitor 4).

MM-PBSA and MM-GBSA both methods are used in this work to calculate the binding free energies of PDK-1 kinase with four inhibitors. There is a good correlation between binding free energy which was calculated by MM-GBSA and experimental values of binding free energy which are derived from the experimental reported IC₅₀ values ($R^2 = 0.55$). To the contrary, the correlation between the binding free energy was calculated by MM-PBSA and experimental values of binding free energy which are derived from the experimental reported IC₅₀ values is weaker ($R^2 = 0.14$). This result agrees with some of the reports that MM-GBSA is considered a better approach than the MM-PBSA in calculating the binding free energies when metals are not involved.

According to IC₅₀ values we conclude that inhibitor (1), inhibitor (2), inhibitor (3), and inhibitor (4) do not possess a drug-like property (IC₅₀ = 93 μ M, 17 μ M, 1.1 μ M, 0.013 μ M), respectively, because they dissociate and do not stay bound to the enzyme.

And these values are correlated with binding free energies of these complexes that calculated by MM-GBSA which are 0.3 K cal/mol, -1.6 Kcal/mol, -21.0 Kcal/mol, -15.0 Kcal/mol, -52.3 Kcal/mol. We noted that IC₅₀ become lower, and binding free energy become larger in negative sign, and become more druggable property.

As PDK-1 is a well validated anticancer target, the final results reveal the binding modes between PDK-1 kinase and the five inhibitors which can be used in the future in drug design for cancer treatment. The position of water molecules in the binding sites of inhibitor (2)-kinase and inhibitor (3)-kinase complexes can be used to design better inhibitors in which the principle lies in the fact that a substituent is added to the ligand that displaces a bound water molecule based on the creation of new inhibitor that includes a structural water mimic.

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APPENDICES

APPENDIX A: SEP NON-STANDARD RESIDUE AND INHIBITOR FILES

Input file for force filed modification of SEP residue

File 1: SEP_leap.frcmod

From VanBeek et al. Biophys J. (2007) 92, 4168-4178	
MASS	
DOND	
BOND	
ANGLE	
OH-P-OH 45.000 109.500	
DIHE	
IMPROPER	
NONBON	

Input file for identification atom types and atom charges of inhibitor (1)

File 2: Inhibitor1.mol2

620								
26	28	1	0	0				
SMA	LL							
bcc								
@ <t]< td=""><td>RIPOS</td><td>S>A'</td><td>TON</td><td>Л</td><td></td><td></td><td></td><td></td></t]<>	RIPOS	S>A'	TON	Л				
1	C1	3	1.42	90	24.4900	5.9150 ca	1 620	-0.035200
2	C2	3	2.83	40	24.3570	5.9830 ca	1 620	0.031400
3	N3	3	1.17	70	25.7430	5.4030 na	1 620	-0.283300
4	C4	3	0.60	50	23.4550	6.3540 ca	1 620	-0.221000
5	N5	3	3.35	80	25.5140	5.4990 nc	1 620	-0.525100
б	C6	3	3.39	20	23.1590	6.4750 ca	1 620	-0.035000
7	C7	3	2.38	40	26.3250	5.1600 cd	1 620	0.468400
8	C8	3	1.16	50	22.2790	6.8370 ca	1 620	0.150100
Input file for force filed modification of inhibitor (1)

File 3: Inhibitor1.frcmod

remark goes	here			
MASS				
BOND				
ANGLE				
DIHE				
IMPROPER				
ca-ca-ca-na	1.1	180.0	2.0	Using default value
ca-ca-ca-nc	1.1	180.0	2.0	Using default value
ca-cd-na-hn	1.1	180.0	2.0	General improper torsional angle (2 general atom
types)				
ca-ca-ca-ha	1.1	180.0	2.0	General improper torsional angle (2 general atom
types)				
cd-na-cd-nc	1.1	180.0	2.0	Using default value
ca-ca-ca-os	1.1	180.0	2.0	Using default value
cc-cd-cd-na	1.1	180.0	2.0	Using default value
cc-cd-cc-ha	1.1	180.0	2.0	Using default value
cd-hn-na-nd	1.1	180.0	2.0	General improper torsional angle (2 general atom
types)				
cc-h4-cc-nd	1.1	180.0	2.0	Using default value
NONBON				

APPENDIX B: INPUT FILES FOR SIMULATION

Input file for minimization of water and ion molecules in inhibitor (1) -protein complex

File 1: min.in

```
Minimization of water
&cntrl
 imin=1,maxcyc=1000,ncyc=500,
 cut=10.0,ntb=1,
 ntc=2,ntf=2,
 ntpr=100,
 ntr=1, restraintmask=':1-282',
 restraint_wt=2.0
/
Hold protein and ligand fixed
10.0
RES 1-283
END
END
```

Input file for minimization of the whole complex

File 2: min_all.in

```
Minimization of the whole system
&cntrl
imin=1,maxcyc=1000,ncyc=500,
cut=10.0,ntb=1,
ntc=2,ntf=2,
ntpr=100,
ntr=0,
/
END
```

Input file for heating the inhibitor (1) -protein complex from 0K to 300K

File 3: eat.in

```
Heating from OK to 300K with weak restraints
&cntrl
imin=0,irest=0,ntx=1,
nstlim=25000,dt=0.002,
ntc=2,ntf=2,
cut=10.0, ntb=1,
ntpr=500, ntwx=500,
ntt=3, gamma_ln=2.0,
tempi=0.0, temp0=300.0, ig=-1,
ntr=1, restraintmask=':1-282',
restraint_wt=2.0,
nmropt=1
/
&wt TYPE='TEMPO', istep1=0, istep2=25000,
value1=0.1, value2=300.0, /
&wt TYPE='END' /
```

Input file for density equilibration of inhibitor (1) -protein complex

File 4: density.in

```
&cntrl

imin=0,irest=1,ntx=5,

nstlim=25000,dt=0.002,

ntc=2,ntf=2,

cut=10.0, ntb=2, ntp=1, taup=1.0,

ntpr=500, ntwx=500,

ntt=3, gamma_ln=2.0,

temp0=300.0, ig=-1,

ntr=1, restraintmask=':1-282',

restraint_wt=2.0,

/
```

Input file for unrestrained equilibration of inhibitor (1) -protein complex

File 5: equil.in

```
&cntrl

imin=0,irest=1,ntx=5,

nstlim=250000,dt=0.002,

ntc=2,ntf=2,

cut=10.0, ntb=2, ntp=1, taup=2.0,

ntpr=1000, ntwx=1000,

ntt=3, gamma_ln=2.0,

temp0=300.0, ig=-1,

/
```

Input file for unrestrained production of inhibitor (1) -protein complex

File 6: prod.in

```
&cntrl

imin=0,irest=1,ntx=5,

nstlim=250000,dt=0.002,

ntc=2,ntf=2,

cut=10.0, ntb=2, ntp=1, taup=2.0,

ntpr=5000, ntwx=5000,

ntt=3, gamma_ln=2.0,

temp0=300.0, ig=-1,

/
```

Input file for running MM-PBSA and MM-GBSA

File 7: mmpbsa.in

```
Input file for running PB and GB
&general
endframe=50, verbose=1,
# entropy=1,
/
&gb
igb=2, saltcon=0.100
/
&pb
istrng=0.100,
/
```

Input file for running entropy calculations using Nmode

File 8: mmpbsa_nm.in

```
Input file for running entropy calculations using NMode
&general
endframe=50, keep_files=2,
/
&nmode
nmstartframe=5, nmendframe=45,
nminterval=5, nmode_igb=1, nmode_istrng=0.1,
/
```

Input file for running mass-weighted RMSD measurements

File 9: mesure_equil_rmsd.ptraj

trajin equil.mdcrd

reference com_wat.inpcrd

rms reference out equil.rmsd @CA,C,N 0.1

Input file for running conversion from mdcrd file to binpos file

File 10: mdcrd_to_binpos.ptraj

trajin prod1.mdcrd

trajin prod2.mdcrd

trajin prod3.mdcrd

trajin prod4.mdcrd

trajout prod.binpos binpos

Input file for hydrogen bonding analysis of inhibitor (1) -protein complex

File 11: analyse_hbond.ptraj

trajin prod.binpos

hbond :1-274 out nhb.dat avgout avghb.dat

APPENDIX C: OUTPUT FILES

Output file of ΔG_{bind} for inhibitor (1)-protein complex that resulted from MM-GBSA

File 1: MM-GBSA for ΔG_{bind} protein-inhibitor (1) complex

GENERALIZED BORN:

Complex: Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS EEL EGB ESURF	-2387.1522 -19906.0747 -4475.4622 88.0450	18.1991 28.5435 20.5100 0.4429	2.5737 4.0367 2.9006 0.0626
G gas G solv	-22293.2270 -4387.4172	34.4401 20.4611	4.8706 2.8936
TOTAL	-26680.6441	32.3518	4.5752
Receptor: Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS EEL EGB ESURF	-2354.6887 -19888.1698 -4477.9034 89.4199	18.1946 28.4029 20.2762 0.4413	2.5731 4.0168 2.8675 0.0624
G gas G solv	-22242.8585 -4388.4835	34.5269 20.2199	4.8828 2.8595
TOTAL	-26631.3420	32.4518	4.5894
Ligand: Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS EEL EGB ESURF G gas G solv	-1.8136 -12.6183 -22.0954 2.1777 -14.4319 -19.9177	0.2344 0.5191 0.4232 0.0136 0.4936 0.4177	0.0331 0.0734 0.0598 0.0019 0.0698 0.0591
TOTAL	-34.3496	0.5901	0.0835
Differences (Complex Energy Component	- Receptor - Ligand): Average	Std. Dev.	Std. Err. of Mean
VDWAALS EEL EGB ESURF	-30.6500 -5.2866 24.5367 -3.5526	1.0904 1.6542 1.0683 0.0346	0.1542 0.2339 0.1511 0.0049
DELTA G gas DELTA G solv	-35.9366 20.9841	1.8382 1.0689	0.2600 0.1512
DELTA TOTAL	-14.9525	1.6290	0.2304

Output file of ΔG_{bind} for inhibitor (2)-protein complex that resulted from MM-PBSA

File 2: MM-PBSA for ΔG_{bind} protein-inhibitor (1) complex

POISSON BOLTZMANN:

Complex: Energy Component	Average	Std. Dev.	Std. Err.	of Mean
VDWAALS EEL EPB	-2387.1522 -19906.0747 -4075.2673	18.1991 28.5435 19.1133		2.5737 4.0367 2.7030
ENPOLAR EDISPER	2236.9806 -1309.5224	3.0541 2.5485		0.4319 0.3604
G gas G solv	-22293.2270 -3147.8090	34.4401 19.3236		4.8706 2.7328
TOTAL	-25441.0360	31.8761		4.5080
Receptor: Energy Component	Average	Std. Dev.	Std. Err.	of Mean
VDWAALS	-2354.6887	18.1946		2.5731
EEL	-19888.1698	28.4029		4.0168
EPB	-4083.8299	19.0226		2.6902
ENPOLAR	2231.3477	3.0596		0.4327
EDISPER	-1318.4636	2.5071		0.3546
G gas	-22242.8585	34.5269		4.8828
G solv	-3170.9459	19.3747		2.7400
TOTAL	-25413.8044	31.9662		4.5207
Ligand: Energy Component	Average	Std. Dev.	Std. Err.	of Mean
VDWAALS	-1.8136	0.2344		0.0331
EEL	-12.6183	0.5191		0.0734
EPB	-21.0535	0.3862		0.0546
ENPOLAR	25.0435	0.1278		0.0181
EDISPER	-27.6378	0.1729		0.0245
G gas	-14.4319	0.4936		0.0698
G solv	-23.6478	0.4276		0.0605
TOTAL	-38.0796	0.5793		0.0819
Differences (Complex	- Receptor - Ligand):			
Energy Component	Average	Std. Dev.	Std. Err.	of Mean
VDWAALS	-30.6500	1.0904		0.1542
EEL	-5.2866	1.6542		0.2339
EPB	29.6162	1.0069		0.1424
ENPOLAR	-19.4106	0.2120		0.0300
EDISPER	36.5790	0.3302		0.0467
DELTA G gas	-35.9366	1.8382		0.2600
DELTA G solv	46.7846	1.1997		0.1697
DELTA TOTAL	10.8480	1.6722		0.2365

Output file of ΔG_{bind} for inhibitor (2)-protein complex that resulted from MM-GBSA

File 3: MM-GBSA for ΔG_{bind} protein-inhibitor (2) complex generalized born:

Complex: Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS	-2382.0840	19.2872	2.7276
EEL	-19827.2949	40.0446	5.6632
EGB	-4408.0595	25.1368	3.5549
ESURF	87.6976	0.4609	0.0652
G gas	-22209.3789	35.4489	5.0132
G solv	-4320.3619	24.9850	3.5334
TOTAL	-26529.7408	27.4538	3.8825
Receptor: Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS	-2357.5888	19.7084	2.7872
EEL	-19954.9192	40.0431	5.6629
EGB	-4425.5261	25.1074	3.5507
ESURF	89.3861	0.4567	0.0646
G gas	-22312.5080	35.2631	4.9869
G solv	-4336.1399	24.9597	3.5298
TOTAL	-26648.6480	27.4911	3.8878
Ligand: Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS	1.2240	1.0341	0.1463
EEL	158.6175	2.1640	0.3060
EGB	-23.0458	0.9649	0.1365
ESURF	2.4317	0.0103	0.0015
G gas	159.8415	2.6065	0.3686
G solv	-20.6142	0.9607	0.1359
Differences (Complex Energy Component	- Receptor - Ligand): Average	Std. Dev.	Std. Err. of Mean
VDWAALS	-25.7192	2.1653	0.3062
EEL	-30.9932	3.1003	0.4384
EGB	40.5125	2.1040	0.2976
ESURF	-4.1202	0.0338	0.0048
DELTA G gas	-56.7124	2.9103	0.4116
DELTA G solv	36.3922	2.0901	0.2956
DELTA TOTAL	-20.3201	1.9515	0.2760

Output file of ΔG_{bind} for inhibitor (2)-protein complex that resulted from MM-PBSA

Complex: Energy Component Average Std. Dev. Std. Err. of Mean -2382.0840 19.2872 40.0446 VDWAALS 2.7276 5.6632 EEL -19827.2949 -4033.4629 EPB 22.8552 3.2322 3.0962 2.5382 ENPOLAR 2225.5414 0.4379 -1297.4909 EDISPER 0.3590 35.4489 G gas -22209.3789 5.0132 35.4489 22.5362 G solv -3105.4124 3.1871 TOTAL -25314.7913 26.9355 3.8092 Receptor: Energy Component Average Std. Dev. Std. Err. of Mean -2357.5888 19.7084 VDWAALS 2.7872 EEL EPB 40.0431 -19954.9192 5.6629 22.9975 3.0854 2.5773 -4053.8735 3.2523 2223.0370 ENPOLAR 0.4363 -1307.2618 EDISPER 0.3645 -22312.5080 -3138.0983 35.2631 22.6852 4.9869 G gas G solv 3.2082 3.8201 TOTAL -25450.6063 27.0122 Ligand: Energy Component Average Std. Dev. Std. Err. of Mean 1.22401.0341158.61752.1640 VDWAALS 0.1463 158.6175 0.3060 EEL EPB -21.0045 0.8966 0.1268 0.1046 ENPOLAR 21.9338 0.0148 -24.3504 EDISPER 0.0226 159.8415 2.6065 G gas 0.3686 -23.4211 2.0005 G solv 0.1273 2.5309 TOTAL 136.4204 0.3579 Differences (Complex - Receptor - Ligand): Energy Component Average Std. Dev. Std. Err. of Mean _____ VDWAALS -25.7192 2.1653 0.3062 -30.9932 3.1003 EEL 0.4384 EPB 0.3208 ENPOLAR 0.0301

File 4: MM-PBSA for ΔG_{bind} protein-inhibitor (2) complex

POISSON BOLTZMANN:

2.2681 0.2129 0.3239 -30.9952 41.4151 -19.4294 EDISPER 34.1212 0.0458 DELTA G gas -56.7124 2.9103 0.4116 DELTA G solv 56.1069 2.2933 0.3243 DELTA TOTAL -0.6054 1.9526 0.2761

Output file of ΔG_{bind} for inhibitor (3)-protein complex that resulted from MM-GBSA

File 5: MM-GBSA for ΔG_{bind} protein-inhibitor (3) complex

GENERALIZED BORN:

Complex: Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS EEL EGB ESURF	-2413.2870 -19820.4055 -4419.2772 83.6140	21.5813 30.7686 22.1522 0.5858	3.0521 4.3513 3.1328 0.0828
G gas G solv	-22233.6925 -4335.6633	34.3084 22.0179	4.8519 3.1138
TOTAL	-26569.3558	30.8426	4.3618
Receptor: Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS EEL EGB ESURF	-2371.3691 -19879.6261 -4435.8541 86.0346	21.3473 30.4154 21.8024 0.5906	3.0190 4.3014 3.0833 0.0835
G gas G solv	-22250.9952 -4349.8194	34.4450 21.6557	4.8713 3.0626
TOTAL	-26600.8147	31.0148	4.3861
Ligand: Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS EEL EGB ESURF G gas G solv	-1.6425 84.7296 -21.0097 2.9199 83.0871 -18.0898	0.7487 1.3193 0.6344 0.0130 1.5013 0.6318	0.1059 0.1866 0.0897 0.0018 0.2123 0.0893
TOTAL	64.9973	1.8088	0.2558
Differences (Complex Energy Component	- Receptor - Ligand): Average	Std. Dev.	Std. Err. of Mean
VDWAALS EEL EGB ESURF	-40.2754 -25.5090 37.5866 -5.3405	1.5773 2.5000 2.4295 0.0522	0.2231 0.3536 0.3436 0.0074
DELTA G gas DELTA G solv	-65.7844 32.2460	2.5661 2.4181	0.3629 0.3420
DELTA TOTAL	-33.5384	1.4634	0.2070

File 6: MM-PBSA for ΔG_{bind} protein-inhibitor (3) complex

POISSON BOLTZMANN:

Complex: Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS	-2413.2870	21.5813	3.0521
EEL	-19820.4055	30.7686	4.3513
EPB	-3993.3020	21.3042	3.0129
ENPOLAR	2204.6542	4.0279	0.5696
EDTSPER	-1269 9511	3 1058	0 4392
LUIJIEN	1209.9911	5.1050	0.4572
G gas	-22233.6925	34.3084	4.8519
G solv	-3058.5989	22.0081	3.1124
TOTAL	-25292.2914	30.6631	4.3364
Receptor:			
Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS	-2371.3691	21.3473	3.0190
EEL	-19879.6261	30.4154	4.3014
EPB	-4025.4893	21.1439	2.9902
ENPOLAR	2199.6821	4.0331	0.5704
EDISPER	-1287.5893	3.0903	0.4370
G gas	-22250,9952	34.4450	4.8713
G solv	-3113.3965	21.6740	3.0652
TOTAL	-25364.3917	31.0895	4.3967
Ligand:			
Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS	-1.6425	0.7487	0.1059
EEL	84.7296	1.3193	0.1866
EPB	-19.4857	0.6965	0.0985
ENPOLAR	32.7201	0.1509	0.0213
EDISPER	-35,1051	0.1973	0.0279
G gas	83.0871	1.5013	0.2123
G solv	-21.8707	0.7350	0.1040
TOTAL	61.2164	1.6242	0.2297
Differences (Comple	x - Receptor - Licond);		
Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS	-40,2754	1.5773	0.2231
FEL	-25,5090	2.5000	0.3536
FPB	51,6731	3 3971	0.1790
	-27 7/20	0 2755	0.4790
ENTSDED	52 7/22	0.2/33	0.0390
LUISPER	32.1433	0.5451	0.0485
DELTA G gas	-65.7844	2.5661	0.3629
DELTA G solv	76.6684	3.2720	0.4627

Output file of ΔG_{bind} for inhibitor (4)-protein complex that resulted from MM-GBSA

File 7: MM-GBSA for ΔG_{bind} protein-inhibitor (4) complex

GENERALIZED BORN:			
Complex: Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS EEL	-2353.3994 -19658.0219	17.5785 34.9942	2.4860 4.9489
EGB	-4319.8457	18,9849	2,6849
ESURF	81.1668	0.6750	0.0955
G gas	-22011.4213	37.1507	5.2539
G solv	-4238.6789	18.7732	2.6549
TOTAL	-26250.1002	33.7015	4.7661
Receptor:			
Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS	-2308.7545	17.7632	2.5121
EEL	-19710.5088	35.4969	5.0200
EGB	-4319.0185	19.2179	2.7178
ESURF	83.4388	0.0004	0.0942
G gas	-22019.2633	37.8085	5.3469
G solv	-4235.5797	18.9902	2.6856
TOTAL	-26254.8430	33.7533	4.7734
Ligand:			
Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS	-1.8345	0.6602	0.0934
EEL	68.3767	1.3554	0.1917
EGB	-29.0700	0.5917	0.0837
ESURF	2.8460	0.0161	0.0023
G gas	66.5422	1.4134	0.1999
G solv	-26.2240	0.5889	0.0833
TOTAL	40.3182	1.5645	0.2213
Differences (Complex -	Receptor - Ligand):		
Energy Component	Average	Std. Dev.	Std. Err. of Mean
VDWAALS	-42.8103	1.8014	0.2548
EEL	-15.8898	3.5751	0.5056
EGB	28.2428	3.3158	0.4689
ESURF	-5.1180	0.0605	0.0086
DELTA G gas	-58.7002	3.1095	0.4397
DELTA G solv	23.1248	3.2962	0.4662
DELTA TOTAL	-35.5753	1.7954	0.2539

File 8: MM-PBSA for ΔG_{bind} protein-inhibitor (4) complex

POISSON BOLTZMANN:

Complex: Energy Component	Average	Std. Dev.	Std. Err. of Mea	an
VDWAALS	-2353.3994	17.5785	2.4860	Э
EEL	-19658.0219	34.9942	4.9489	9
EPB	-3927.1281	17.4236	2.464	1
ENPOLAR	2158.3678	3.8127	0.5392	2
EDISPER	-1224.9080	3.1128	0.4402	2
G gas	-22011.4213	37.1507	5.253	9
G solv	-2993.6683	17.1491	2.425	2
TOTAL	-25005.0896	31.3041	4.427	1
Recentor:				
Energy Component	Average	Std. Dev.	Std. Err. of Mea	an
VDWAALS	-2308.7545	17.7632	2.512	1
EEL	-19710.5088	35,4969	5.0200	9
EPB	-3936.5249	17.6422	2,4950	9
ENPOLAR	2151.7913	3.8115	0.5390	- 9
EDISPER	-1240.8397	3.1114	0.4400	9
G gas	-22019.2633	37.8085	5.346	9
G solv	-3025.5734	17.3662	2.455	9
TOTAL	-25044.8367	31.2005	4.4124	4
Ligand:		1000 B	1000 a con 112100	
Energy Component	Average	Std. Dev.	Std. Err. of Mea	n
		· · · · · · · · · · · · · · · · · · ·		
VDWAALS	-1.8345	0.6602	0.0934	ł
EEL SOO	68.3767	1.3554	0.1917	
EPB	-20.9990	0.0359	0.0899	1
ENPOLAR	33.9443	0.1/51	0.0248	5
EDISPER	-38.3390	0.1905	0.0278	5
6 035	66 5422	1 4124	0 1000	
G solv	-31 3950	0 7515	0.1993	2
0 3000	-51.5950	0.7515	0.100.	,
τοται	35 1472	1 3986	0 1979	2
IUTAL	55.1472	1.5500	0.1570	·
Differences (Complex	· - Receptor - Ligand);			
Energy Component		Std Dev	Std Err of Me:	20
Lifergy component				
VDWAAL S	-42,8103	1.8014	0.254	B
FEL	-15,8898	3, 5751	0.5050	5
EPB	36.3964	3,3776	0.477	7
ENPOLAR	-27.3677	0.2681	0.0379	9
EDISPER	54,2713	0.3780	0.053	5
	5112125	0.0700	0.000.	-
DELTA G gas	-58.7002	3,1095	0.4397	7
DELTA G solv	63.3000	3.4647	0.4900	Э

File 9: MM-GBSA for ΔG_{bind} protein-inhibitor (5) complex

GENERALIZED BORN:

	Complex: Energy Component	Average	Std. Dev.	Std. Err. of Mean
	VDWAALS	-2334.6961	13.3627	1.8898
	EEL	-20111.3512	40.3633	5.7082
	EGB	-4065.6798	22.2102	3.1410
	ESURF	88.9084	0.4125	0.0583
	G gas	-22446.0473	41.1968	5.8261
	G solv	-3976.7714	22.1986	3.1394
	TOTAL	-26422.8187	31.1430	4.4043
	Receptor:			
	Energy Component	Average	Std. Dev.	Std. Err. of Mean
	VDWAALS	-2257.1401	12.9660	1.8337
	EEL	-20139.3788	39.9200	5.6455
	EGB	-4094.2921	22.2570	3.1476
	ESURF	94.0722	0.3976	0.0562
	G das	-22396.5189	41.7082	5,8984
	G solv	-4000.2199	22.2518	3.1469
	TOTAL	-26396.7388	31.7675	4.4926
	Ligand:			
	Energy Component	Average	Std. Dev.	Std. Err. of Mean
	VDWAALS	-1.1075	1.4022	0.1983
	EEL	83.6674	1.2794	0.1809
	EGB	-40.1660	0.7469	0.1056
	ESURF	4.4974	0.0174	0.0025
	Glas	82.5599	1,9406	0.2744
	G solv	-35.6687	0.7471	0.1056
	TOTAL	46.8913	1.9986	0.2826
Г)ifferences (Complex -	Receptor - Ligand):		
E	Energy Component	Average	Std. Dev. S	td. Err. of Mean
		-76 //05	2 2222	0 4573
	FI	- 55 6309	3 1700	0.4373
C	CR	68 7794	1 4712	0.4405
C		-9 6612	0 1114	0.2001
C	JUNE	-9.0012	0.1114	0.0158
D	ELTA G gas	-132.0884	3.3350	0.4716
D	DELTA G solv	59.1172	1.4823	0.2096
D	DELTA TOTAL	-72.9712	2.7883	0.3943

Output file of ΔG_{bind} for inhibitor (5)-protein complex that resulted from MM-PBSA

File 10: MM-PBSA for ΔG_{bind} protein-inhibitor (5) complex

	POISSON BOLTZMANN:			
	Complex:	Aug 5200	std Dov	Std Err of Moor
	Energy component	Average	Std. Dev.	Std. Err. or mean
	VDWAALS	-2334 6961	13 3627	1 8808
	FEI	-20111 3512	40 3633	5 7082
	EDR	- 3542 1323	21 0261	2 9735
		2228 7802	1 9792	0 2656
		1252 4595	1.0702	0.2030
	EDISPER	-1252.4505	1.7487	0.2475
	6 0.35	- 22446 0473	41 1968	5 8261
		-2555 9106	20 6993	2 9259
	0 3000	-2333.0100	20.0885	2:9230
	TOTAL	-25001.8578	32.5969	4.6099
r	Pacantor:			
F	Energy Component	Aver 300	std Dev	Std Err of Mean
	inergy component	Average		Std. Err. or hear
1	DWAALS	-2257.1401	12,9660	1.8337
F	FI	-20139 3788	39,9200	5.6455
Ē	PB	- 3573, 4939	20,9491	2,9626
F		2235,8833	1.8308	0.2589
F	DISPER	-1281,1244	1.7811	0.2519
1		1201,1244	1.7011	0.2517
0	aas	-22396.5189	41,7082	5,8984
0	G solv	-2618.7350	20,6246	2,9168
1	3.3004	201017550	20.0240	2.9100
٦	TOTAL	-25015.2538	33.3003	4.7094
L	igand:			
E	Energy Component	Average	Std. Dev.	Std. Err. of Mean
١	/DWAALS	-1.1075	1.4022	0.1983
E	EEL	83.6674	1.2794	0.1809
E	EPB	-37.6917	0.6818	0.0964
E	ENPOLAR	54.4222	0.1989	0.0281
E	EDISPER	-53.3586	0.2648	0.0375
C	i gas	82.5599	1.9406	0.2744
C	G solv	-36.6281	0.7588	0.1073
٦	TOTAL	45.9319	1.9257	0.2723
	ifferences (Complex	Decenter Licend)		
	component -	Receptor - Ligand)	. std Dov s	td Err of Moon
C	nergy component	Average	Std. Dev. 3	stu. Err. or mean
V	DWAALS	-76.4485	3, 2333	0.4573
F	FL	-55.6398	3,1700	0.4483
F	PB	69.0533	1,7267	0.2442
F	NPOLAR	-51,5253	0.2645	0.0374
F	DISPER	82.0245	0.6233	0.0881
-			0.0233	010001
P	ELTA G gas	-132.0884	3,3350	0.4716
D	ELTA G solv	99.5525	2.0608	0.2914
1		ಎನ್.ಎ.ಕ.ಕ.ಕ.ಕ.ಕ.		
-				

Output file of ΔS for inhibitor (1)-protein complex that resulted from Nmode

File 11: Entropic contribution of protein-inhibitor (1) complex

Complex:				
Entropy Term	Average	Std. Dev.	Std. Err.	of Mean
Translational Rotational Vibrational Total	16.9773 17.5628 3217.1317 3251.6718	0.0000 0.0068 3.2815 3.2827		0.0000 0.0023 1.0938 1.0942
Receptor: Entropy Term	Average	Std. Dev.	Std. Err.	of Mean
Translational Rotational Vibrational Total	16.9713 17.5549 3199.0826 3233.6089	0.0000 0.0017 3.2836 3.2845		0.0000 0.0006 1.0945 1.0948
Ligand: Entropy Term	Average	Std. Dev.	Std. Err.	of Mean
Translational Rotational Vibrational Total	12.5107 9.6839 11.0817 33.2764	0.0000 0.0000 0.4783 0.4783		0.0000 0.0000 0.1594 0.1594
Differences (Complex Entropy Term	<pre>c - Receptor - Ligand):</pre>	Std. Dev.	Std. Err.	of Mean
Translational Rotational Vibrational	-12.5047 -9.6760 6.9673	0.0000 0.0062 3.2004		0.0000 0.0021 1.0668
DELTA S total=	-15.2135	3.2025		1.0675

ENTROPY RESULTS (HARMONIC APPROXIMATION) CALCULATED WITH NMODE:

Output file of ΔS for inhibitor (2)-protein complex that resulted from Nmode

File 12: Entropic contribution of protein-inhibitor (2) complex

	Anomic Arthornation, calco			
Complex: Entropy Term	Average	Std. Dev.	Std. Err.	of Mean
Translational Rotational Vibrational Total	16.9773 17.5751 3221.8426 3256.3949	0.0000 0.0023 3.0359 3.0350		0.0000 0.0008 1.0120 1.0117
Receptor: Entropy Term	Average	Std. Dev.	Std. Err.	of Mean
Translational Rotational Vibrational Total	16.9713 17.5723 3205.1449 3239.6885	0.0000 0.0012 2.6761 2.6765		0.0000 0.0004 0.8920 0.8922
Ligand: Entropy Term	Average	Std. Dev.	Std. Err.	of Mean
Translational Rotational Vibrational Total	12.5148 9.6022 13.3378 35.4548	0.0000 0.0000 0.4639 0.4639		0.0000 0.0000 0.1546 0.1546
Differences (Comple Entropy Term	x - Receptor - Ligand): Average	Std. Dev.	Std. Err.	of Mean
Translational Rotational Vibrational	-12.5089 -9.5994 3.3599	0.0000 0.0027 2.6447		0.0000 0.0009 0.8816
DELTA S total=	-18.7484	2.6433		0.8811

ENTROPY RESULTS (HARMONIC APPROXIMATION) CALCULATED WITH NMODE:

Output file of ΔS for inhibitor (3)-protein complex that resulted from Nmode

File 13: Entropic contribution of protein-inhibitor (3) complex

ENTROPY RESULTS	(HARMONIC APPROXIMATION)	CALCULATED WITH NMODE	:
Complex: Entropy Term	Average	Std. Dev.	Std. Err. of Mean
Translational Rotational Vibrational Total	16.9734 17.5666 3205.8049 3240.3448	0.0000 0.0027 4.7201 4.7214	0.0000 0.0009 1.5734 1.5738
Receptor: Entropy Term	Average	Std. Dev.	Std. Err. of Mean
Translational Rotational Vibrational Total	16.9659 17.5490 3176.7016 3211.2167	0.0000 0.0113 9.3158 9.3266	0.0000 0.0038 3.1053 3.1089
Ligand: Entropy Term	Average	Std. Dev.	Std. Err. of Mean
Translational Rotational Vibrational Total	12.7072 10.2694 18.7394 41.7160	0.0000 0.0022 0.0090 0.0068	0.0000 0.0007 0.0030 0.0023
Differences (Con Entropy Term	mplex - Receptor - Ligand) Average	: Std. Dev. S	Std. Err. of Mean
Translational Rotational Vibrational	-12.6997 -10.2518 10.3639	0.0000 0.0114 7.2080	0.0000 0.0038 2.4027
DELTA S total=	-12.5878	7.2183	2.4061

Output file of ΔS for inhibitor (4)-protein complex that resulted from Nmode

File 14: Entropic contribution of protein-inhibitor (4) complex

ENTROPY RESULTS (HAI	RMONIC APPROXIMATION) CALCO	LATED WITH NMU	DE:	
Complex: Entropy Term	Average	Std. Dev.	Std. Err.	of Mean
Translational Rotational Vibrational Total	16.9561 17.5128 3132.5212 3166.9900	0.0000 0.0017 4.7875 4.7872		0.0000 0.0006 1.5958 1.5957
Receptor: Entropy Term	Average	Std. Dev.	Std. Err.	of Mean
Translational Rotational Vibrational Total	16.9474 17.5062 3108.9020 3143.3558	0.0000 0.0027 7.1749 7.1748		0.0000 0.0009 2.3916 2.3916
Ligand: Entropy Term	Average	Std. Dev.	Std. Err.	of Mean
Translational Rotational Vibrational Total	12.8252 10.4617 20.9163 44.2035	0.0000 0.0001 0.0024 0.0026		0.0000 0.0000 0.0008 0.0009
Differences (Complex Entropy Term	x - Receptor - Ligand): Average	Std. Dev.	Std. Err.	of Mean
Translational Rotational Vibrational	-12.8166 -10.4552 2.7028	0.0000 0.0026 3.6238		0.0000 0.0009 1.2079
DELTA S total=	-20.5692	3.6237		1.2079

ENTROPY RESULTS (HARMONIC APPROXIMATION) CALCULATED WITH NMODE.

Building a library file for SEP residue

The connectivity information in the pdb file was deleted. Using *xLeap*, atoms are

bonded together manually. First, the SEP non-standard residue was loaded (Fig 1)

Xleap SEP = loadpdb SEP.pdb

S 🕒 🗉 XLEaP: Universe Editor
File Edit Verbosity
<pre>Welcome to LEaP! (no leaprc in search path) > SEP = loadpdb SEP.pdb Loading PDB file: ./SEP.pdb Unknown residue: SEP number: 0 type: Terminal/last relaxing end constraints to try for a dbase match -no luck Created a new atom named: N within residue: .R<sep 241=""> Created a new atom named: CA within residue: .R<sep 241=""> Created a new atom named: CB within residue: .R<sep 241=""> Created a new atom named: CG within residue: .R<sep 241=""> Created a new atom named: CW within residue: .R<sep 241=""> Created a new atom named: CW within residue: .R<sep 241=""> Created a new atom named: CW within residue: .R<sep 241=""> Created a new atom named: OW within residue: .R<sep 241=""> Created a new atom named: OW within residue: .R<sep 241=""> Created a new atom named: OW within residue: .R<sep 241=""> Created a new atom named: O1P within residue: .R<sep 241=""> Created a new atom named: O2P within residue: .R<sep 241=""> Created a new atom named: O3P within residue: .R<sep 241=""> Created a new atom named: O3P within residue: .R<sep 241=""> Created a new atom named: O3P within residue: .R<sep 241=""> Created a new atom named: O3P within residue: .R<sep 241=""> Created a new atom named: O3P within residue: .R<sep 241=""> Created a new atom named: O3P within residue: .R<sep 241=""> Created a new atom named: O3P within residue: .R<sep 241=""> Created a new atom named: O3P within residue: .R<sep 241=""> Created a new atom named: O3P within residue: .R<sep 241=""> Created a new atom named: O3P within residue: .R<sep 241=""> Created a new atom named: O3P within residue: .R<sep 241=""> total atoms in file: 10 The file contained 10 atoms not in residue templates >.</sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></sep></pre>



All atoms of SEP residue were added to the new *UNIT*. It is important to check that *xLeap* "created" the correct atoms. The easiest way is to check for the total number of atoms in the file which should be 10.

The pdb file containing atom coordinates for SEP residue was loaded. *XLeap* does not have necessary SEP parameters and connectivity data, so this information was entered manually.

The SEP residue in *xLeap* was edited (Fig 2) by the command:



🛞 🖨 🗉 XLEaP: Unit editor: SEP	S 🖨 🗊 XLEaP: Unit editor: SEP		
Unit Edit Display	Unit Edit Display		
-Manipulation ◇Select ◇Twist ◇Move ◇Erase ◇Draw	- Manipulation ◇Select ◇Twist ◇Move ◇Erase ◇Draw		
(a) 0 N S P +other elements	(b) O N S P +other elements		
	H12 H3 H3 H7 H14 H17 H14 H17 H18 H2 H2 H2 H2 H2 H2 H2 H2 H2 H2 H2 H2 H2		

Figure 2: *XLeap* window shows (a) atoms of SEP residue before connecting between them, (b) SEP non-standard residue after connection

Parameters of each atom in SEP residue are given by parm99.dat file which

lies *in \$AMBERHOME/dat/leap/parm* table given in *xLeap*. The atom types of all

SEP atoms were identified by blue table (Table 1).

Table 1: Blue table is given by XLeap shows all of atom types in SEP non-standard residue

	- 04 1 4 1						
XLEaP: Edit selected atoms: SEP							
Table Opera	Table Operations						
Table Editor: The Table Editor: Tab	Table Editor: The table has no errors and 0 warnings. Table Editor: Table has been 'saved' back to the program.						
NAME	NAME TYPE CHARGE ELEMENT unused PERT.name PERT.type DELTA.charge						
N	NT	0,000000	N				0,000000
CA	СТ	0.000000	C				0.000000
CB	СТ	0.000000	C				0.000000
OG	OS	0.000000	0				0.000000
C	CT	0.000000	C				0.000000
0	OS	0.000000	0				0.000000
Р	Р	0.000000	Р				0.000000
01P	02	0.000000	0				0.000000
02P	OH	0.000000	0				0.000000
03P	OH	0.000000	0				0.000000
H12	H	0.000000	H				0.000000
H13	H	0.000000	H				0.000000
H14	H1	0.000000	H				0.000000
H15	H1	0.000000	H				0.000000
H16	H1	0.000000	H				0.000000
H17	H1	0.000000	H				0.000000
H18	H1	0.000000	H				0.000000
H21	HO	0.000000	H				0.000000
H22	НО	0.000000	Н				0,000000

The O2P and O3P atoms are oxygen atoms in hydroxyl group, so the type of this oxygen is OH in the blue table. But the O1P is oxygen atom in phosphate group, so the type of this atom is O2 according to PARM99.dat¹⁷

H12 and H13 are connected with nitrogen atom, so these atoms were assigned with an appropriate atom type which is H. HC and H1 were assigned for H in aliphatic bond to Carbon without electron withdrawing group and H in aliphatic bond to Carbon with one electron withdrawing group, respectively.¹⁷

Then library file of SEP was saved which will enable *xLeap* to recognize this residue in the future. This step is very essential to prevent the repetition of all of the previous steps each time. This was done by the following command.

```
Saveoff SEP SEP_leap.lib
Savepdb SEP SEP_leap.pdb
```

The missing bonds and angle parameters of SEP residue were identified by *xLeap*. This was achieved by using the following commands

```
xleap -s -f $AMBERHOME/dat/leap/cmd/oldff/leaprc.ff99SB
loadoff SEP_leap.lib
check SEP
```

S 🗢 🗉 XLEaP: Universe Editor
File Edit Verbosity
<pre>Sourcing: /home/mazen/amber16/dat/leap/cmd/oldff/leaprc.ff99SB Log file: ./leap.log Loading parameters: /home/mazen/amber16/dat/leap/parm/parm99.dat Reading title: PARM99 for DNA,RNA,AA, organic molecules, TIP3P wat. Polariz.& LP incl.02/04/99 Loading parameters: /home/mazen/amber16/dat/leap/parm/frcmod.ff99SB Reading force field modification type file (fromod) Reading title: Modification/update of parm99.dat (Hornak & Simmerling) Loading library: /home/mazen/amber16/dat/leap/lib/all_nucleic94.lib Loading library: /home/mazen/amber16/dat/leap/lib/all_amino94.lib Loading library: /home/mazen/amber16/dat/leap/lib/all_amino194.lib Loading library: /home/mazen/amber16/dat/leap/lib/all_amino194.lib Loading library: /home/mazen/amber16/dat/leap/lib/all_amino194.lib Loading library: /home/mazen/amber16/dat/leap/lib/solvents.lib Loading library: /home/mazen/amber16/dat/leap/lib/solvents.lib Loading library: /home/mazen/amber16/dat/leap/lib/solvents.lib > loading library: /sEP.lib Loading library: /sEP.lib Checking 'SEP' Checking for bond parameters. Checking for bond parameters. Checking for angle parameters. Checking for angle parameters. Chucking for angle parameters. Could not find angle parameters. Unit is OK.</pre>

Figure 3: *XLeap* window shows the missing parameters of non-standard residue (SEP)

An *frcmod* file s required to provide all the bonds, angles and dihedral parameters that are not present in the standard FF99SB force field. The only missing parameter in SEP residue is the OH-P-OH angle parameter (Fig 3). So SEP_leap.frcmod file (see Appendix A) was created to define the missing OH-P-OH angle parameter. Then the *frcmod* file of SEP was loaded using this command

```
loadamberparams SEP_leap.frcmod
```

Finally, the SEP residue was successfully built using *xLeap*, by check the residue (UNIT is OK) and by saving it as *prmtop* and *inpcrd* file (Fig 4).

🔊 🚍 🔲 🛛 XLEaP: Universe Editor

File Edit Verbosity

```
Loading parameters: /home/mazen/amber16/dat/leap/parm/parm99.dat

Reading title:

PARM99 for DNA,RNA,AA, organic molecules, TIP3P wat. Polariz.& LP incl.02/04/99

Loading parameters: /home/mazen/amber16/dat/leap/parm/frcmod.ff99SB

Reading force field modification type file (frcmod)

Reading title:

Modification/update of parm99.dat (Hornak & Simmerling)

Loading library: /home/mazen/amber16/dat/leap/lib/all_nucleic94.lib

Loading library: /home/mazen/amber16/dat/leap/lib/all_amino94.lib

Loading library: /home/mazen/amber16/dat/leap/lib/all_aminor94.lib

Loading library: /home/mazen/amber16/dat/leap/lib/all_aminor94.lib

Loading library: /home/mazen/amber16/dat/leap/lib/all_aminor94.lib

Loading library: /home/mazen/amber16/dat/leap/lib/solvents.lib

> loadoff SEP_leap.lib

Loading library: ./SEP_leap.lib

> loadoff SEP_leap.lib

Loading parameters: ./SEP_leap.frcmod

Loading parameters: ./SEP_leap.frcmod

Reading force field modification type file (frcmod)

Reading for bond parameters.

Checking for bond parameters.

Checking for bond parameters.

Checking for angle parameters.

Checking for angle parameters.

Checking for angle parameters.
```

Figure 4: *XLeap* window shows that SEP residue succussefully built using *xLeap*

Creating AMBER input files

The *inpcrd* and *Prmtop* files are the coordinate files and molecular topology/parameter, respectively. These files are necessary for running molecular dynamics simulation of protein-ligand complexes using *Sander*.

Antechamber is designed to be used with the "general AMBER force field (*GAFF*)", and was successfully used in the production of *frcmod* files and *mol2* files of inhibitors. *GAFF* force field covers most pharmaceutical molecules and is

compatible with *AMBER* force fields. *GAFF* is a complete force field and covers all the organic molecules that contain C, N, O, S, P, H, F, Cl, Br and I.⁷⁴

The hydrogenated 6-methoxy-2-(1H-pyrazol-5-yl)-1H-benzimidazole (inhibitor 1) coordinates were done using *Pymol*, then *antechamber* command was used to create the "*mol2*" file using the following command in terminal:

```
antechamber -i inhibitor1.pdb -fi pdb -o inhibitor1.mol2 -fo mol2 -
c bcc -s 2
```

This command line produced a number of files in CAPITALS. These files are used by antechamber and are not required here. These files are considered as intermediate files, but *mol2* file of inhibitor1 (see Appendix A) is the most important one because it reveals the definition of our inhibitor (1) residue, including all of the charges and atom types.

To specify any missing parameters (bonds, angles, dihedral angles) before we can create our *prmtop* and *inpcrd* files in *Leap*, we run the *parmchk* command in terminal to test if all the parameters we require are available.

```
parmchk -i inhibitor1.mol2 -f mol2 -o inhibitor1.frcmod
```

Running this command produced a file called *inhibitor1.frcmod* (see Appendix A). This is a parameter file that can be loaded into *xLeap* in order to add missing parameters.

Subsequently, *xLeap* was used to form *prmtop* and *inpcrd* files using the

following command was typed in terminal in order to open *xLeap*:

xleap -s -f \$AMBERHOME/dat/leap/cmd/leaprc.ff99SB

This command line starts *xleap* and loads the configuration files needed for

AMBER FF99SB force field as shown in Figure 5.

```
> SXLEAP: Universe Editor
File Edit Verbosity

Velcome to LEAP!
Sourcing: /home/admins/Desktop/amber16/dat/leap/cmd/leaprc.ff99SB
Log file: ./leap.log
Loading parameters: /home/admins/Desktop/amber16/dat/leap/parm/parm99.dat
Reading title:
PARM99 for DNA.RNA.AA, organic molecules, TIP3P wat. Polariz.6 LP incl.02/04/99
Loading parameters: /home/admins/Desktop/amber16/dat/leap/parm/frcmod.ff99SB
Reading force field modification type file (frcmod)
Reading library: /home/admins/Desktop/amber16/dat/leap/lib/all_muno94.lib
Loading library: /home/admins/Desktop/amber16/dat/leap/lib/all_mino94.lib
Loading library: /home/admins/Desktop/amber16/dat/leap/lib/all_amino94.lib
Loading library: /home/admins/Desktop/amber16/dat/leap/lib/all_amino194.lib
Loading library: /home/admins/Desktop/amber16/dat/leap/lib/solvents.lib
>.
```

To ensure that *xLeap* has the *GAFF* force field, it is loaded into *xLeap* by

using command line:

Source leaprc.gaff

xLeap looks like this:



Figure 6: XLeap window shows preparing to load the protein-inhibitor (1) complex X-ray structure

Now inhibitor (1) unit (inhibitor1.mol2) is loaded:

```
620 = loadmol2 inhibitor1.mol2
```

loadamberparams inhibitor1.frcmod

The library file for inhibitor (1) was created, as well as the prmtop and inpcrd

files using the command lines:

```
saveoff 620 inhibitor1.lib
saveamberparm 620 inhibitor1.prmtop inhibitor1.inpcrd
```

Inhibitor (1) can be seen (Fig 7 (a)) using edit command:

edit 620


Figure 7: XLeap window showing the graphical representation of (a) inhibitor (1), (b) inhibitor (2), (c) inhibitor (3), (d) inhibitor (4), and (e) inhibitor (5), respectively

Now *xLeap* was ready to load the protein_inhibitor1.pdb file without having any problems. protein_inhibitor (1).pdb file was loaded into xleap after setting it as a new unit called "com" by writing the following commands:

```
Xleap -f $AMBERHOME/dat/leap/cmd/leaprc.ff99SB
Source leaprc.gaff
Loadoff SEP_leap.lib
Loadamberparams SEP_leap.frcmod
620 = loadmol2 620.mol2
Loadamberparams 620.frcmod
Com = loadpdb protein_inhibitor1_dry.pdb
Edit com
```

The *xLeap* window shows the graphical representation of protein_inhibitor1_dry.pdb (Figure 8).



Figure 8: *Xleap* window showing the graphical representation of protein_inhibitor1_dry.pdb

```
set default PBRadii mbondi2
saveamberparm com protein_inhibitor1_dry.prmtop
protein_inhibitor1_dry.inpcrd
saveamberparm 620 inhibitor1.prmtop inhibitor1.inpcrd
```

 $\Delta G^{0}_{nonpolar}$ was calculated using the default parameters $\Upsilon = 0.00500$ kcal/Å² and $\beta = 0.0000$ kcal/mol this was achieved using the mbondi2 radius, because this method is affective in the calculation of non-polar solvation energy. The system was neutralized by adding counter ions:

charge com

Thus, two Cl- atoms were added to neutralize the protein_inhibitor1 complex. This task was done using the following order:

addions com Cl- 0

This command line causes a columbic potential on a grid of 1Å resolution and then puts the counter ions simultaneously at the points of lowest/greatest electrostatic potential (Figure 9).

XLEaP: Universe Editor File Edit Verbosity 🛛 From VanBeek et al. Biophys J. (2007) 92, 4168-4178 > loadoff SEP.lib Loading library: ./SEP.lib > com = loadpdb 5hng.pdb Loading PDB file: ./Shng.pdb One sided connection. Residue: default_name missing connect0 atom. One sided connection. Residue: default_name missing connect1 atom. Lotal atoms in file: 2314 Leap added 2335 missing atoms according to residue templates: 2335 H / lone pairs > charge com 2.000000 2.000000 Total unperturbed charge: Total perturbed charge: > addions com Cl- 0 2 Cl- ions required to neutralize. Adding 2 counter ions to "com" using 1A grid Grid extends from solute vdw + 2.47 to 8.57 Resolution: 1 grid build: 0 sec 1.00 Angstrom. (no solvent present) Calculating grid charges charges: 14 sec Placed Cl- in com at (51.08, 42.64, -5.67). Placed Cl- in com at (32.08, 48.64, 3.33). Done adding ions. > ^

Figure 9: *XLeap* editor shows neutralization of the protein_inhibitor1_dry.pdb complex by addition of chlorine ions

Finally, the system was solvated using the following command to add a periodic rectangular box of TIP3P within a distance from the surface of the box to the closest atom of the solute was set to 10 Å in x, y, and z directions (Figure 2.10).⁷⁵

solvatebox com TIP3PBOX 10.0

The prmtop and inpcrd files for the solvated system were saved using the

Following commands:

```
saveamberparm com protein_inhibitor1_wat.prmtop
protein_inhibitor1_wat.inpcrd
```

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Figure 10: *Xleap* window showing the graphical representation of protein_inhibitor1_wat.pdb