

HOMOLOGY MODEL, DOCKING ANALYSIS AND MOLECULAR DYNAMICS SIMULATION OF CANNABINOID CB2 RECEPTOR

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Manuscript received: July 2019

Abstract

Using cannabinoid CB1 as structural template, the 3D model of CB2 receptor was established with homology model method, and refined with molecular dynamics method. The docking of well-known reference antagonist compounds with CB2 was studied according to the model. Then the structure and components of active site in CB2 were investigated by aromatic interactions, hydrogen bond interactions and binding free energy analysis.

Rezumat

Folosind receptorul canabinoid CB1 ca model structural, a fost stabilită structura 3D a receptorului CB2 prin metoda modelului de omologie și definită prin intermediul dinamicii moleculare. Amestecul de compuși antagoniști de referință ai CB2 a fost studiat conform modelului de andocare. Ulterior, structura și componentele situsului activ din CB2 au fost cercetate prin intermediul interacțiunilor aromatice, legăturilor de hidrogen și analiza energiei libere de legare.

Keywords: CB2 receptor, homology model, molecular dynamics

Introduction

Cannabinoid receptors are involved in a variety of physiological processes such as appetite, mood, pain-sensation and memory. The potential therapeutic applications targeted cannabinoid receptors include management of glaucoma, attenuation of nausea and vomiting, suppression of muscle spasticity and therapeutic effects of analgesia [1, 2]. Cannabinoid receptors were abbreviated CB and numbered by a subscript in the order of their discovery. Now two cannabinoid receptor types were identified unequivocally, named CB1 and CB2. Although they share a certain degree of structural homology, there are differences in their tissue distribution and their signalling mechanisms. CB1 was located primarily in the central and peripheral nervous system. The antagonists or inverse agonists of CB1 have potential therapeutic application, such as treatment schizophrenia [3] and appetite suppressants [4]. The side effects of this therapeutic application are involved in dysphoria, sedation, euphoria and alterations in cognition and memory [5]. CB2 was located mainly in immune cells including tonsils, thymus, T lymphocytes, B lymphocytes, monocytes, macrophages, polymorphonuclear cells and natural killer (NK) cells. CB2 plays roles in immune regulation and inhibiting the release of cytokines. It was showed that CB2 is down-regulated at protein and mRNA levels during B-cell differentiation [6]. Recent studies showed that CB1 and CB2 were

overexpressed on tumour cells in various types of cancers, and therefore could be used as novel targets for cancer [7].

Both cannabinoid receptors belong to the G-protein-coupled receptors (GPCRs) [8]. The characteristics of GPCRs are seven hydrophobic transmembrane helices (TMH) and are integral membrane proteins. Therefore, the structure of cannabinoid receptors is very important for their function and for drug design. The human CB1 receptor, which was performed the point mutation T210A and replaced the third intracellular loop (ICL3) with the thermostable *Pyrococcus abyssi* glycogen synthase (PGS) domain, was determined at high resolution by X-ray crystallography [9]. This 3D structure of CB1 has allowed modelling CB2 with greater reliability.

Classical cannabinoids are tricyclic terpenoid derivatives such as D⁹-tetrahydrocannabinol bearing a benzopyran moiety. There are some cannabimimetic including tricyclic (such as CP55244) and bicyclic (such as CP55940) analogues which lack the pyran rings. Diarylpyrazole such as SR144528 is another type of cannabinoid analogues that is a selective and potent CB2 antagonist. SR144528 displays sub-nanomolar affinity (K_i = 0.6 nM) for both the rat spleen and cloned human CB2 receptors [10]. Other cannabinergics are the amino-alkylindoles that are potential non-steroidal anti-inflammatory agents. AM-630 as aminoalkylindoles

Results and Discussion

Homology model

Both cannabinoid receptor types contain seven hydrophobic TMH belonging to the large family of GPCRs

[25]. CB2 shared 46% of identity with CB1. The sequential alignment of CB1 and CB2, produced by CLUSTALW program, was showed in Figure 2.

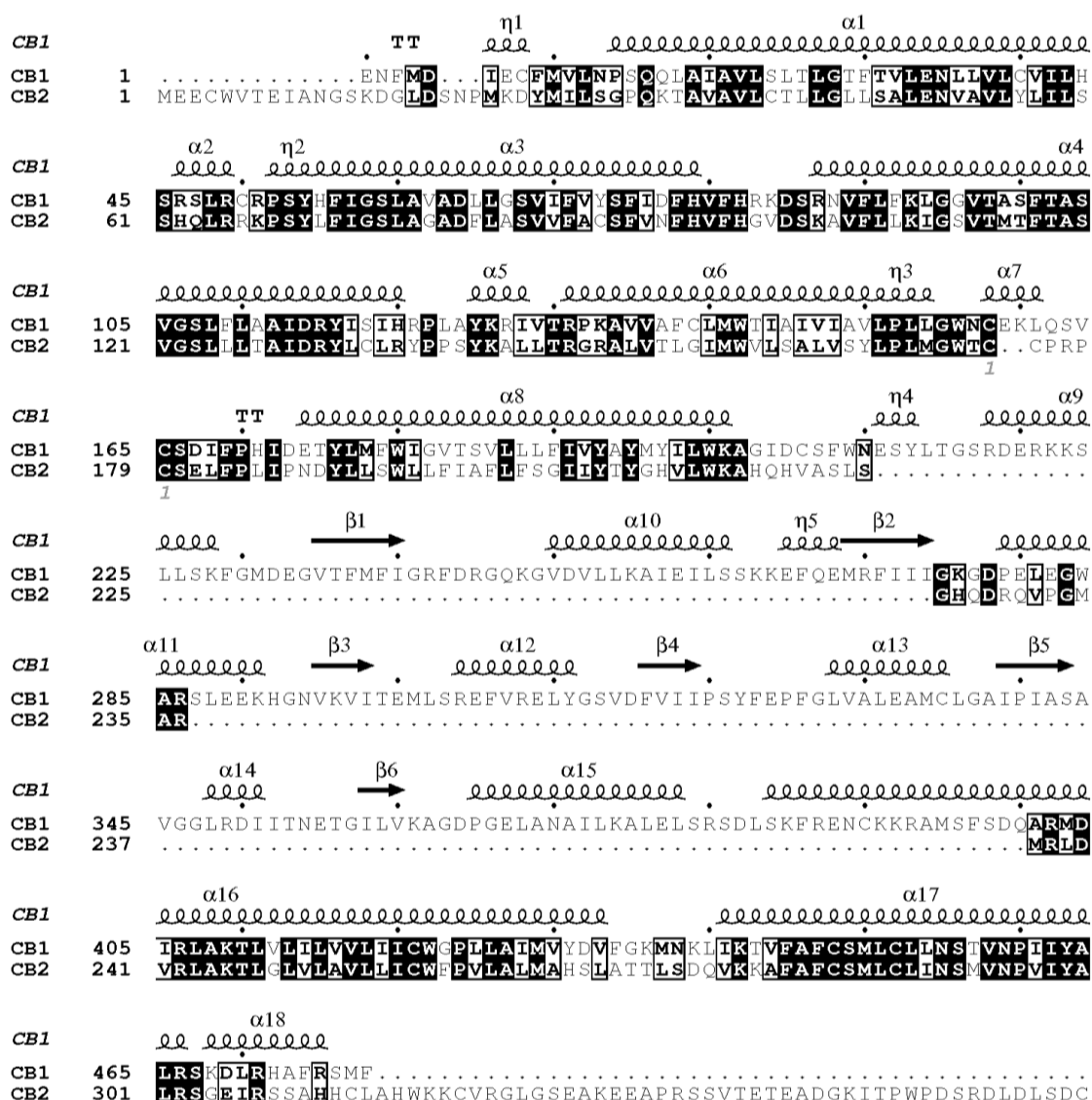


Figure 2.

Sequence alignment of 5U09 with CB2 sequence. The conserved patterns are boxed.

Most of the characteristic residues of CB2 are conserved in CB1. Compared to other transmembrane region, TM5 shows a rich aromatic region in CB1 and CB2, more remarkable in CB2. The most important differences are located in extracellular loop II (EL2), the N-terminal, C-terminal of TM7 and the C-terminal between CB1 and CB2 receptors. CB1 and CB2 possess a long N-terminal, and did not obtain any significant hit by searching in PSI-BLAST [26], and did not present in template structure. A region of rich proline was located in EL2 of CB2, while a region (RxAFRS) was located in C-terminal of CB1 which is well conserved in cation channel receptors (e.g. vanilloid receptor VR1) [27]. In CB1 crystal structure, the conformation

of EL2 is constrained by the presence of an intra-loop disulphide bond (Cys158-Cys165), also in CB2 there are two cysteines conserved (Cys174-Cys179) that could form a disulphide bridge. The sequence identity percentage of the TMH is 48%, 68%, 62%, 36%, 36%, 59% and 77% for TM1, TM2, TM3, TM4, TM5, TM6 and TM7 respectively. Except TM1, TM4 and TM5, the identity percentage between CB1 TMH and CB2 TMH is around 60%. This is consistent with the substrates selectivity in these regions.

A homology model of human CB2 receptor was established with human CB1 (PDB ID: 5U09) as template (Figure 3). The 3D-model was confirmed by ERRAT and PROCHECK. Ramachandran plot of the model

exhibited that 99.1% of residues are in “most favoured regions” and “additional allowed regions”, and only two residues are in “disallowed regions”, which means a good stereochemistry for over 99% of the residues. An overall quality factor of 94.9% was produced by ERRAT which is a program for verifying protein structures and compares to reliable high-resolution structures. It means that the calculated error values of 94.9% of all residues are below the 95% rejection limit (Figure 4). The transmembrane helices of model were superposed with the CB1 template in order to identify the orientation of the CB2 helices. The root-mean-square (RMS) value is 0.35 Å. So it is highly consistent between the transmembrane helices. The 3D model of CB2 constructed here is agreed with the studies of Martinelli *et al.* [28].

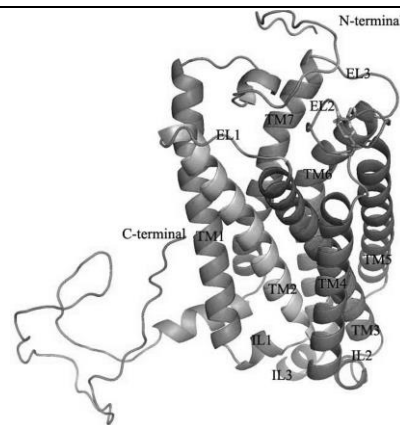


Figure 3.

The model of CB2 receptor drawn as cartoon diagram.
The disulphide bond is shown as sticks models

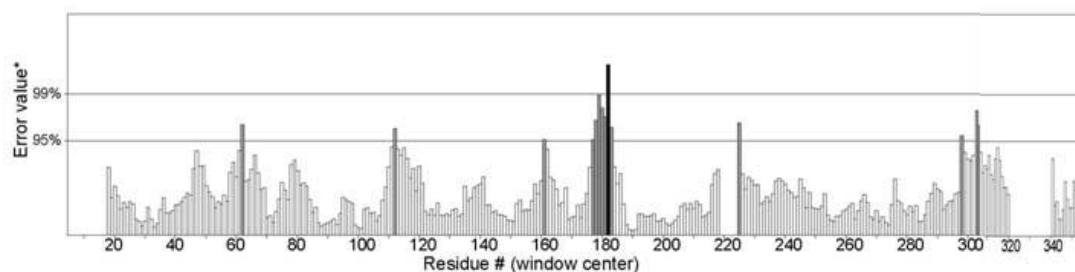


Figure 4.

The error values of each residue in CB2 model produced by ERRAT

The CB2 model conformation did not change drastically during the MD simulation of 10 ns. The RMSD values for CB2 backbone stayed within 1 nm range (Figure 5). In the root mean square fluctuation (RMSF) plot, residues with the largest fluctuations are highlighted during the simulation (Figure 6). We can conclude that the increase of the average RMSD was involved in fluctuations of flexible loops, N-terminal and C-terminal. By contrast, TMHs exhibited the lowest fluctuations and appeared stable during the MD simulation.

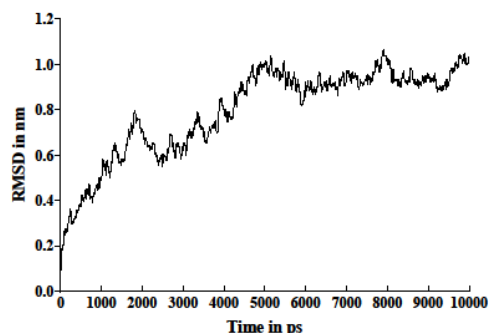


Figure 5.

CB2 backbone root mean squared deviation (RMSD) graph

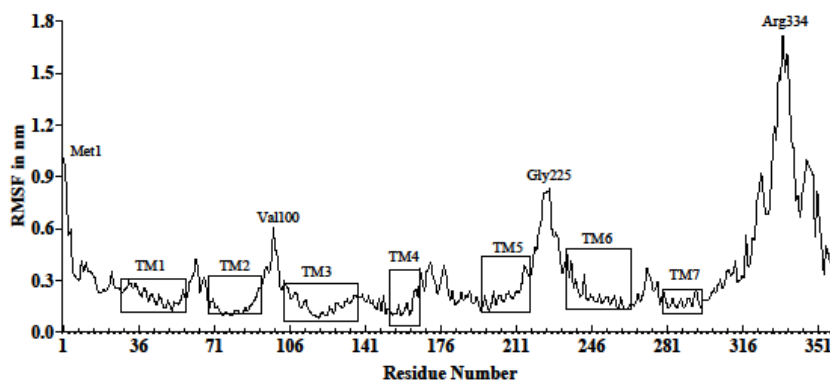


Figure 6.

Residue-specific RMSF during the MD simulation

Cannabinoid ligands docking

The best CB2 model was used with the fewest restraints violations and lowest value of the pdf from 20 models obtained with MODELLER. Docking was carried out on two selective CB2 antagonists to investigate the characteristics of CB2 binding site (Figure 7). Generally, the interactions between ligand and protein are mediated by specific aromatic interactions and hydrogen bond interactions [29]. So, aromatic interactions and hydrogen bond interactions were used as the major criterion for analysis. This binding site was computed by ligand protein contact (LPC) server [30]. Table I illustrated the key LPC data predicted for ligands as well as the k_i values and the binding free energy. Docking results exhibited that the CB2 active site composed of the TM2-TM3-E2-TM5-TM6-TM7 region. The hydrogen bond interactions and aromatic interactions were predicted

by LPC for the docking. The docking studies are consistent with the following binding free energy results.

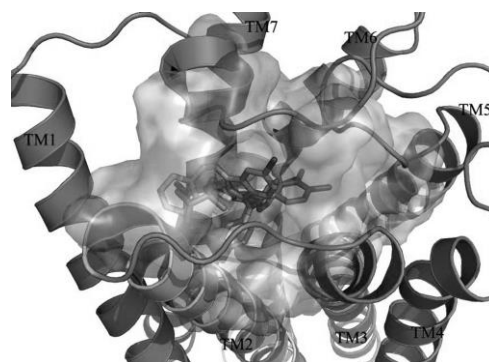


Figure 7.

Binding mode of AM630 and SR144528 with CB2. The compounds are represented as sticks.

Table I

The docking results of AM630 and SR144528 with CB2

Ligand	Hydrogen bond	Aromatic–aromatic contact	binding free energy (kJ/mol)	K_i (nM)
AM630	O(CO)-Phe87	Tyr25, Phe87, Phe106, Phe183, Tyr258, Phe281	-174.8	32.1 [11]
SR144528	N(1)-Ser285 N(2)-Ser285	Phe87, Phe183, Tyr258, Phe281	-185.7	0.6 [10]

Binding free energy analysis

The binding free energy of AM630 is -174.8 kJ/mol predicted by G_MMPBSA, composed of electrostatic energy -20.8 kJ/mol, van der Waals energy -242.8 kJ/mol, SASA energy -24.1 kJ/mol and polar solvation energy 112.8 kJ/mol, while the predicted value of SR144528 is -185.7 kJ/mol, composed of electrostatic energy -19.2 kJ/mol, van der Waals energy -259.6 kJ/mol, SASA energy -27.1 kJ/mol and polar solvation energy 120.2 kJ/mol. Generally the more potent ligands show, the lower binding free energy protein-ligand

complexes display. But there is not the linear relationship between biological activity and binding free energy. So, the most significance is the intermolecular van der Waals contribution. The result is in agreement with the fact that the large hydrophobic binding surface was produced between compounds AM630 and SR144528 with CB2. Following the work of Gohlke *et al.* [31], the decomposition of binding free energy, called per-residue decomposition, was performed using *g_mmpbsa* tool in order to further study the ligand-protein interactions (Figure 8).

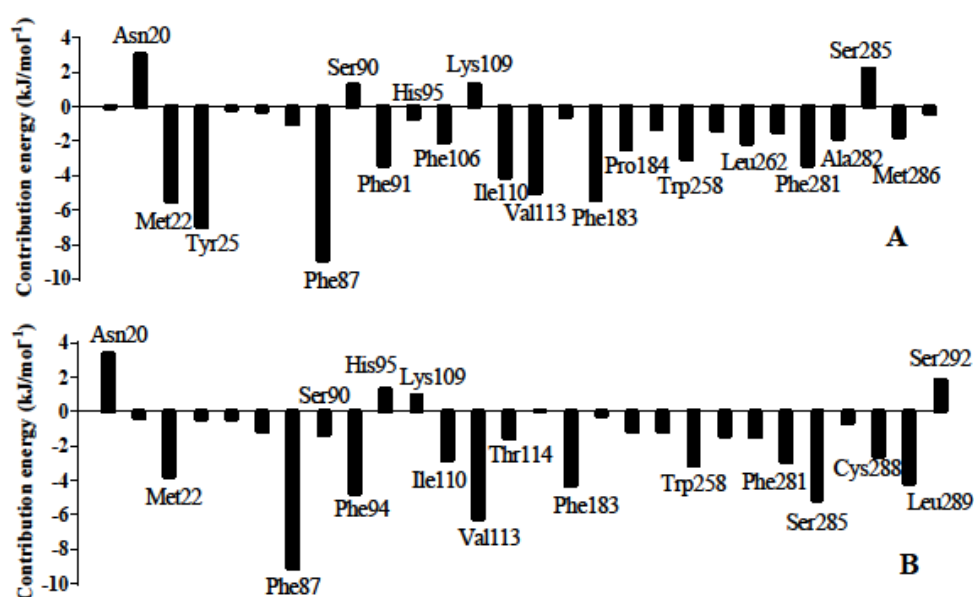


Figure 8.

Energy contribution of each residue to the binding of AM630 (A) and SR144528 (B) with CB2 active site

The binding free energy decomposition analysis exhibits that the common contributions are residues Met22, Phe87, Ile110, Val113, Phe183, Trp258, Phe281 with the binding for compounds AM630 and SR144528. The docking and binding free energy analysis is similar to what Xie and co-workers found that residues, val113, phe183 and phe281, are important for ligand binding to CB2 [32]. Residues Ser90 and Ser285 are unfavourable for binding AM630, while are favourable for binding SR144528. On the contrary, residue His95 is in dis-favour for binding SR144528 and in favour for binding AM630.

Conclusions

The 3D model of the CB2 was built on the basis of the highest resolution structural template of CB1. The interactions of the protein–ligand complexes are investigated by molecular docking and MD simulations, then analysed based on aromatic interactions, hydrogen bond interactions and binding free energy analysis. The studies showed that the binding process between ligand and CB2 is governed by hydrogen bond interactions and aromatic interactions. Because the cannabinoid receptors are important targets, the results reported here are very important for the searching of new selective and potent cannabinoid ligands. The design of novel cannabinoid ligands is in progress.

Acknowledgement

The project was supported by Jiangxi Province Science Foundation (20171BAB205104).

Conflict of interest

The authors declare no conflict of interest.

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