Molecular Modeling-Based Evaluation of hTLR10 and Identification of Potential Ligands in Toll-Like Receptor Signaling

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Abstract

Toll-like receptors (TLRs) are pattern recognition receptors that recognize pathogens based on distinct molecular signatures. The human (h)TLR1, 2, 6 and 10 belong to the hTLR1 subfamilies, which are localized in the extracellular regions and activated in response to diverse ligand molecules. Due to the unavailability of the hTLR10 crystal structure, the understanding of its homo and heterodimerization with hTLR2 and hTLR1 and the ligand responsible for its activation is limited. To improve our understanding of the TLR10 receptor-ligand interaction, we used homology modeling to construct a three-dimensional (3D) structure of hTLR10 and refined the model through molecular dynamics (MD) simulations. We utilized the optimized structures for the molecular docking in order to identify the potential site of interactions between the homo and heterodimer (hTLR10/2 and hTLR10/1). The docked complexes were then used for interaction with ligands (Pam3CSK4 and PamCysPamSK4) using MOE-Dock and ASEDock. Our docking studies have shown the binding orientations of hTLR10 heterodimer to be similar with other TLR2 family members. However, the binding orientation of hTLR10 homodimer is different from the heterodimer due to the presence of negative charged surfaces at the LRR11-14, thereby providing a specific cavity for ligand binding. Moreover, the multiple protein-ligand docking approach revealed that Pam3CSK4 might be the ligand for the hTLR10/2 complex and PamCysPamSK4, a di-acylated peptide, might activate hTLR10/1 hetero and hTLR10 homodimer. Therefore, the current modeled complexes can be a useful tool for further experimental studies on TLR biology.

Introduction

Toll-like receptors (TLRs) are evolutionarily conserved membrane-bound pattern recognition receptors (PRRs) that recognize a broad spectrum of microbial components such as lipopeptides and non-self nucleic acids [1]. TLRs play a central role in innate immunity and are required for the development of adaptive immune responses. TLRs are type I transmembrane glycoproteins that consist of an extracellular domain/ectodomain (ECD), a single transmembrane spanning segment and a globular cytoplasmic Toll/interleukin (IL)-1 receptor (TIR) domain [2,3]. The ECDs encompass 19–25 tandem copies of a motif known as the leucine-rich repeat (LRR), whose primary function appears to be ligand recognition and the formation of protein-protein interactions with its partners [4,5,6]. The cytoplasmatic TIR domain recruits the adaptor protein that is essential for the downstream signaling. To date, 10 hTLRs (human TLRs) and 12 mTLRs (mouse TLRs) have been identified, each of which responds to specific microbial products [2,7,8]. TLRs are categorized into six major families: TLR1, TLR3, TLR4, TLR5, TLR7 and TLR11. The classification is mainly based on amino acid similarities and the ligand properties [9]. The present study mainly focused on the TLR1 family, which includes TLR1, TLR2, TLR6 and TLR10.

Although TLR10 has been identified in humans and its equivalent homologs can be found in other mammals, it is absent from mice due to a retrovirus insertion. TLR10 is predominantly expressed in immune cell-rich tissues, such as the small intestine, stomach, thymus, peripheral blood lymphocytes, lymphnodes and tonsils [10,11]. Co-immunoprecipitation studies have shown the self association of TLR10, as well as TLR1 and TLR2 via the extracellular domain of these receptors, which plays a pivotal role in inducing type I interferon in plasmacytoid dendritic cells (PDC) [12]. The ligand specificity has been elucidated for most of the TLRs. Specifically, TLR2 and TLR4 recognize Gram-positive and Gram-negative bacterial cell wall products, respectively, while TLR5 recognizes a structural epitope of bacterial flagellin and TLR3, 7, 8 and 9 have been demonstrated to recognize different forms of microbial-derived nucleic acids [2,13]. However, the ligand specificity of TLR10 remains elusive. The present study was conducted to identify the hTLR10 self dimerization region as well as its partners and possible ligands to activate these complexes.

Computational modeling has become an essential tool in guiding and enabling rational decisions with respect to hypothesis driven biological research. In the absence of an experimentally determined structure, homology modeling could provide a rational opportunity to obtain a reasonable 3D structure. The goal of...
homology modeling is to model or predict the structural coordinates of a query protein based on the known structure of a sequence homology (template). The 3D structure of a protein provides important information for understanding its biochemical function and interaction properties in molecular detail. Here, we used a combination of homology modeling and MD simulation to construct a detailed 3D model of hTLR10. This model was then used to investigate the nature of the three different binding modes with hTLR1 and hTLR2, which provided the complex formation of the hTLR10/1, hTLR10/2 hetero and hTLR10 homodimer. These complex structures were then studied to analyze the intermolecular contacts and to identify the specific residual level of interactions between the two proteins. The results presented in this study demonstrated that the binding orientations are similar among all TLRs; however, the residual interactions with their partners are specific. The hTLR10 complexes were then used for protein-ligand docking to identify the potential ligands that activate the TLR10 signaling. Finally, we have provided possible ligands for the hTLR10 complex activation.

Methods

Homology modeling

The primary sequence of hTLR10 (accession number: AAQ58667) and crystal structure coordinates of hTLR1 (PDB ID: 2Z7X-B) were loaded into the Molecular Operating Environment (MOE). The primary structure of hTLR1 and hTLR10 were aligned and carefully checked to avoid deletions or insertions in each LRR and then corrected manually based on the LRR motif identified by TollML [14,15]. A series of 10 hTLR10 models were independently constructed with MOE by using the Boltzmann-weighted randomized procedure. There was no difference in the number and organization of secondary structural elements and no significant main chain deviation among the 10 models developed. However, the model with the highest packing score was selected for full energy minimization (MOE packing score is -2.5478). The selected final model was subjected to MD simulation for structural refinement [16]. The glycosylation sites present in the final model were predicted by using the NetOglyc server [17].

Molecular dynamics simulation

The MD simulations were conducted using the GROMACS 3.3.3 software [18]. The individual structures of the modeled hTLR10, 1 and 2 were placed into a cubic box maintaining 10 Å between the box edges and the protein surface. The resulting system was then solvated with the simple point charge (SPC) water molecule and then minimized by using steepest descent method with GROMOS96 43a1 force field. Periodic boundary conditions were applied in all directions and the system was neutralized by adding appropriate counter ions (Na+ or Cl-). A twin range cutoff was used for long-range interaction: 9 Å for van der Waals interactions and 14 Å for electrostatic interactions. All bond lengths were constrained with the LINCS algorithm [19]. The SETTLE algorithm was applied to constrain the geometry of water molecules [20]. The energy minimized system was subjected to 100 ps equilibration and subsequently used in the 2 ns production with a time step of 1 femtosecond (fs) at constant temperature (300 K), pressure (1 atm) and number of particles, without any position restraints. In each case, the final conformation obtained at the end of the simulation was further refined by energy minimization for molecular docking. The refined model was validated using the PROCHECK program [21] and the Verify 3-dimensional (3D) server [22,23].

Molecular docking

The three widely accepted rigid-body protein-protein docking programs, GRAMM-X [24], ZDOCK [25] and RosettaDock [26] were used to predict and assess the interactions between hTLR10/1, hTLR10/2 and hTLR10 homodimer complexes. Each docking method returned the 100 most probable predictions out of thousands of candidates based on geometry, hydrophobicity and electrostatic complementarity of the molecular surface. We thus performed restrained pairwise docking for hTLR10 heterodimer complex based on the comparative studies results. Our comparative studies clearly indicated that C-terminal region of LRR11–14 plays a crucial role in dimerization. When we used our comparative studies results as a constraint for homodimer, we obtained the complex similar to heterodimer. However, the active site was positioned far away from the ligand binding region and hence, it was not appropriate to conduct further studies. Consequently, the residues located near the hTLR10 active site were considered as a constraint in our docking studies. The buried surface interaction area of dimer models were calculated using the PROTORP server (protein-protein interface analysis) [27].

Electrostatic potential calculation

The electrostatic potential surface of the hTLR10, 2 and 1 were calculated using the PyMOL APBS (http://apbs.sourceforge.net) tools. The molecular surfaces and ribbons of the figures were built and colored with the PyMOL visualizing tool [28].

Computational alanine scanning to predict the binding energy Hot-spots in the complex interface

The hTLR10/1, hTLR10/2 hetero and hTLR10 homodimer were submitted to the Robetta Alanine scanning server (http://roberta.bakerlab.org/alscansubmit.jsp) to predict the energetically important amino acid residues in its interface. The alanine scanning server calculates the effects of alanine mutations on the binding free energy of a protein-protein complex by using a simple free energy function. The program replaces each of the interface residues with alanine residue and computes the effect of this mutation on the binding free energy of the complex. This analysis serves as the identification of important residues that are crucial for hTLR10 homo and heterodimer formation.

Construction and geometry optimization of hTLR10 ligands

Two ligands, Pam3CSK4 and PamCysPamSK4, have been predicted to activate the hTLR10 complexes. The crystal structure of hTLR1/2/Pam3CSK4 complex is known, and we took the Pam3CSK4 site for docking. In case of PamCysPamSK4, we modified the Pam3CSK4 at C2 position. Partial charges were added to protonated and unprotonated molecules using the Merck Molecular Force Field 94X (MFF94X), which is suitable for small molecules [29]. All structures were energy minimized using the conjugated gradient/truncated Newton optimization algorithm with a convergence criterion of 0.05 kcal/mol, ε = 1.

Binding site selection and exploration

The Site Finder module of MOE 2008.10 was used to identify the possible substrate binding pockets within the newly generated 3D structure of the hTLR10/1, hTLR10/2 hetero and hTLR10 homodimer. Hydrophobic or hydrophilic alpha spheres served as probes denoting zones of tight atom packing. The alpha spheres were used as centroids for the creation of dummy atoms to define the potential sites throughout the docking process.
hTLR10 complex-ligand docking

Protein-ligand docking was conducted using MOE-Dock 2008.10 and ASEDock. The docking procedure has been previously described [30], and the same procedure was followed for hTLR10 complexes and its ligands.

Validation of the hTLR10 ligand docking process via hTLR1/2/Pam3CSK4 docking

Two docking methods, MOE-Dock 2008.10 and ASEDock, were used to assess the validity of the hTLR10 complex-ligand docking predictions by calculating possible bound conformations of hTLR1/2/Pam3CSK4 complexes. The crystal structure of hTLR1/2 was retrieved from the PDB and prepared for docking, and partial charges and hydrogen atoms were added. The obtained docking poses (hTLR1/2/Pam3CSK4) were then compared with the original crystal structure.

Results and Discussion

Molecular model of hTLR10 and evaluation

We built the 3D structure of hTLR10 using the crystal structure of hTLR1, which serves as a suitable template (PDB ID 2Z7X-B) [31]. Generally, TLR ectodomains are composed of a repeated number of LRR, which vary in numbers among the known TLRs [32]. Recent modeling studies of endosomal TLRs (hTLR7, 8, 9) have shown that researchers used multiple templates to derive an optimum alignment for all endosomal LRRs [33]. Conversely, we used a single template (2Z7X-B) because it has 19 LRR, which are aligned well with the 19 LRR in the target hTLR10. The primary sequence alignment of hTLR10 and its template was used to construct the final model (Figure 1). The identity and similarity between the target and template were 40.74% and 59.26%, respectively. Additionally, the sequence identity of each LRR between the target/template is listed in Table 1. Direct comparison of the modeled hTLR10 with its template showed that the structural differences observed in the loop regions of the LRR6, LRR14, LRR15, LRR19 and the rest of the regions were similar to that of the template with RMSD values of 1.14 Å. These structural differences occur away from the ligand binding sites (Figure S1). Moreover, when we compared hTLR10 with the available crystal structure of hTLR2 (in the same family) [31], we found that the variation in loop regions in LRR1, LRR2, LRR3, LRR4, LRR5, LRR7, LRR8, LRR11, and the rest of the regions was similar to that of hTLR2 with RMSD values of 1.34 Å. The final 3D structure of hTLR10 showed large characteristic “horseshoe-like” structures with concave surfaces formed by 19 parallel β-strands, whereas the convex surface contains more diverse secondary structural elements, including a variety of different length loops, 12 α-helices and three 3_10 helices. Like all TLRs, hTLR10 also contains the N-linked glycosylation consensus sites. The glycans were exemplified to be non-functional for ligand binding [25,28–31]. These unusual regions of hTLR1 and 2 possess hydrophobic leucine, which can be replaced by other hydrophobic amino acids such as isoleucine, valine and phenylalanine. The consensus asparagine (N) in the LRR motif plays an important role in the formation of the hydrogen bonds with the carbonyl backbone of the neighboring strands [5,35]. The hTLR10 N-terminal domain was composed of 1–4 LRR modules, with each LRR module length being around 24 residues and the structurally important asparagine ladder being conserved. Conversely, the central domain is composed of 5–11 LRR modules and differs considerably from the standard structure, with the lengths of their LRR modules ranging from 20–30 amino acids and their β-sheet conformations deviating significantly from those of the standard LRRs. The absence of an asparagine ladder in the central domain leads to the formation of an unusual β-sheet conformation. However, we observed such unusual regions in hTLR1 and 2. These unusual regions of hTLR1 and 2 possess hydrophobic pockets, which have been reported to play an essential role in ligand recognition [31]. This analysis shows that residues situated in the ligand binding hydrophobic pockets in hTLR2 and 1 are conserved in hTLR10 (Figure 3 shaded by cyan). Hence, we believe that this region in hTLR10 might play a role in ligand recognition. Finally, the C-terminal domain of hTLR10 contains 12–19 LRR modules and the asparagine ladder is conserved in all LRRs, with each LRR module ranging in length from 20–30
amino acids. Bell et al. predicted that there were insertions that occur at the position of LRR12 in hTLR2, LRR4 in hTLR10, and LRR12 in hTLR1 [32]. However, our structure based analysis of hTLR10 along with the crystal structures of hTLR1 and hTLR2 did not reveal such an insertion in these three receptors.

### Receptor dimerization interface

The signaling mechanism of all TLRs is likely involved in dimerization of the ectodomains [36]. However, this can be achieved in various ways by using different receptors and stimulation. It should be noted that the ligand induced

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**Figure 1. Sequence alignment used to build hTLR10, based on the hTLR1 template retrieved by MOE.** Gray blocks: level of sequence similarity. Tallest blocks: identical residues at that position. Intermediate blocks: non-identical residues that are relatively conservative with respect to their properties. Small blocks: residues sharing mild conservation with respect to structure or function. The absence of a block indicates no appreciable structure/function conservation. Gaps in one sequence relative to the other are indicated by dashes. 2Z7X-B denotes hTLR1. The nineteen LRR are highlighted and contrasted by varying the colors. doi:10.1371/journal.pone.0012713.g001
dimerization takes place in the TLR signaling, and a few crystal structures have already been solved based on this idea [31,37,38]. However, this cannot be conducted computationally; hence, we initially focused on the dimerization mechanism of hTLR10 with hTLR2 and 1 using protein-protein docking. The procedure of protein-protein docking is highly computationally oriented. The reliability of docking results strongly depends on the quality of the docking methods. To verify the prediction confidence of the TLR-TLR interaction of these three methods, GRAMM-X, ZDOCK and RosettaDock, we restrainedly inputted hTLR1 and hTLR2, for which the heterodimeric crystal structure is known, as test cases [31]. The native dimerization structure of hTLR2/1 was found in the top 100 solutions and it was ranked 23rd by GRAMM-X, 1st by RosettaDock and 80th by Z-DOCK. These results demonstrate that our docking protocol was reliable; therefore, we used it in the subsequent hTLR10/2, hTLR10/1 hetero and hTLR10 homo-dimer docking calculations.

We conducted restrained rigid-body docking of the hTLR10 homo and heterodimer. Each docking returned the 100 most probable models from unbound monomer components. Thus, each complex received a total of 300 candidate models separated into three sets. Some models from the same set had similar conformations, whereas most differed considerably from one another. There were some shared models (intersection) across both sets for each complex. These shared models were considered as more confident solutions than others. The optimal docking solution was selected for each complex from the 300 candidates based on the following criteria: (i) models that do not exist in the intersection of the three resulting sets were excluded; (ii) only those models in which the dimerization geometry was supported by the results obtained from our comparative studies were included. The comparative studies clearly indicated that the dimerization region is located at the border of the central region followed by the C-terminal end (Figure 3 represented in green color shade). Additionally, we considered the active site region to be located on the adjacent sides in the dimer interface. This two-step filtering led to a unique solution. The ZDOCK/GRAMM-X/RosettaDock ranking and the buried surface interaction area of all optimal models are provided in Table 2.

Table 1. Sequence identities (%) of target-template of each LRR pairs.

| LRRs  | hTLR10 | LRRs  | hTLR10 |
|-------|--------|-------|--------|
| 1     | 45.83  | 11    | 24.14  |
| 2     | 62.5   | 12    | 40.91  |
| 3     | 38.1   | 13    | 41.67  |
| 4     | 68     | 14    | 36     |
| 5     | 47.83  | 15    | 37.5   |
| 6     | 26.09  | 16    | 45.45  |
| 7     | 41.67  | 17    | 43.48  |
| 8     | 23.33  | 18    | 36.36  |
| 9     | 44.44  | 19    | 20.83  |
| 10    | 50     | Avg   | 40.74  |

Note: LRRs, Leucine rich repeats; hTLR10, human Toll like receptor10; Avg, average value.

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Figure 3. Structure-based sequence alignments of hTLR10, 2 and 1. The hTLR10, 2 and 1 sequences are aligned based on their structures. Conserved leucines and asparagine ladder are written in red and green, respectively. The positions of β-strands are shown above the consensus patterns.

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Table 2. Ranking and interaction area of the selected docking models.

| Complexes      | GRAMM-X | Z-DOCK | RosettaDock | Interaction Area (Å²) |
|----------------|---------|--------|-------------|----------------------|
| hTLR10/hTLR2   | 33      | 26     | 1           | 945.58               |
| hTLR10/hTLR1   | 45      | 73     | 8           | 809.63               |
| hTLR10/hTLR10  | 81      | 59     | 13          | 858.74               |

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hTLR10-hTLR1 complex

The buried interface surface area between the hTLR10 and the hTLR10 is 809.63 Å² from hTLR10 and 798.12 Å² from hTLR1 (Figure 5A&B). There are 15 residues from hTLR10 that makes contact (ΔASA>1 Å²) with 10 residues from hTLR1. The non-covalent interactions such as hydrophobic and ionic interactions across the hTLR10 and the hTLR1 are listed in Figure 5C. Five hydrogen bonds are present in the interface of the hTLR10/1 complex (Figure 5D). The NH₂ group of R337 forms two hydrogen bonds with carboxyl group of E385 and E363, respectively. The carboxamide group of Q383 forms a hydrogen bond with the carboxyl group of E385 of hTLR10. The side chain of K383 forms two hydrogen bonds with the backbone oxygen group of L359 and the backbone oxygen group of N357. Single apostrophes are used for the hTLR10 residues and double apostrophes for the hTLR2 to differentiate them from the ligand throughout the article.

hTLR10-hTLR10 dimer

The hTLR10 homodimer docking orientation is the same, but the dimerization region is different from the hTLR10 heterodimer complexes. The hTLR10 homodimerization region includes LRR11-17 in comparison to LRR11-14 in the heterodimer complex. Such types of dimerization have been conspicuously noted in TLR4 homodimer. The hTLR10 homodimer docked complex resembles a typical ‘m’ shaped heterodimer, with two N-termini extending outwards in opposite directions and LRRCT modules converging at the center (Figure 6A&B). However, the hTLR10 homodimer possesses more hydrophobic and hydrogen bond interactions than its heterodimer complex. The buried surface area across the hTLR10 homodimer complex is 858.74 Å². There are 13 and 14 residues from hTLR10 chain D and E* that contribute to the above interface. Figure 6C shows the hydrophobic and ionic interaction between the homodimer of hTLR10/10*. Eight hydrogen bonds are present at the interface of the hTLR10 homodimer, which includes the carbonyl backbone of N407 and Y430 that forms two hydrogen bonds with the side chain of K432. Similarly, the side chains of K432, Q454, N362 and K383 form hydrogen bonds with the side chains of N404, N423, K356 and E358, respectively. The backbone oxygen group of K409 forms hydrogen bonds with the side chain of Y403 (Figure 6D). Single apostrophes are used for the hTLR10 residues and asterisks for the other hTLR10 monomer to differentiate them from the ligand throughout the article.
The docking results revealed that the amino acid residues across the interface of the hTLR10 homo and heterodimer vary depending on their partners, which clearly demonstrates that hTLR10 binds specifically with its partners. The predicted docked complexes were cross checked by the electrostatic potential of the individual structure. The surface electrostatic calculation from hTLR10 revealed that the highly negative charged surface along with a few positive patches were located at the central region, followed by the C-terminal (LRR11-14) (represented by the green color in Figure 5), whereas hTLR2 and 1 contain highly basic amino acids (Figure 2B). The hTLR10 negatively charged surface (LRR11-14) would bind with hTLR2 and 1 (LRR11-14) based on their respective opposite charges. However, in the case of hTLR10, homodimer formation occurs between LRR11-17 of each monomer. These LRR regions of each monomer have eight negative and four positive charges where only a few residues are involved in the complementary interaction has shown in Figure S3.

Virtual Ala scanning
An approximate estimation of the individual contributions of the amino acid residues involved in the interaction was obtained by computational methods. We used the Rosetta interface computational mutagenesis approach [39], which is similar in principle to the experimental Ala-scanning mutagenesis procedure, to estimate the change in the binding free energy (ΔΔGbind) when each residue at the interface of the hTLR10/2, hTLR10/1 and hTLR10/10* complex is mutated to Ala. We utilized a cut-off of ΔΔGbind of 1.0 kcal/mol to qualitatively identify hot-spot residues that are essential for the interactions. Correctly identified essential residues have predicted and observed (ΔΔGbind) values greater than or equal to 1 kcal/mol; whereas correctly identified neutral residues have predicted and observed (ΔΔG bind) values less than 1 kcal/mol.

To verify the prediction confidence of alanine-scanning mutagenesis, we input the hTLR2/1 crystal structure as a test case. The program predicted a few residues across the complex interface. Of those, P315 in TLR1 has a value of 2.65. A mutation in this residue leads to loss of the hTLR1/2 signaling [40]. This shows that P315 is essential for the dimerization of hTLR1/2 complex. Therefore, we believe that our protocol is reliable and we subsequently used it to evaluate the docked complexes. The alanine scanning experiments showed that two residues (T361 and K383) from hTLR10 and two
residues (K347 and Y376) from hTLR2 were crucial for the formation of hTLR10/2 complex (Table 3), emphasizing their contribution to the complex stability. Conversely, evaluation of the hTLR10/1 complex revealed that two residues from hTLR10 (K383 and E385) and two residues from hTLR1 (R337 and Q383) were indispensable for hTLR10/1 complex formation. Finally, the dimer complex showed that one residue from hTLR10 (K383) and one residue from hTLR10* (K356) were essential for homodimer complex formation. The rest of the residues at both complex interfaces had DDGbind values >0.1 kcal/mol, which suggests that their mutation to Ala does not have any functional consequences on the formation of complexes.

Possible ligand recognition for hTLR10 based on the cavity volume

Recent studies have revealed the binding of the tri-acylated (Pam3CSK4) and di-acylated lipopeptide (Pam2CSK4) to hTLR2/1 and mTLR2/6 complex, respectively [31, 41].

![Figure 6. Docking model of the hTLR10/10* complex.](image)

**Table 3. Results of the predicted contributions of residues through virtual alanine scanning.**

|          | TLR10/2 |          | TLR10/1 |          | TLR10/10* |
|----------|---------|----------|---------|----------|-----------|
|          | PDB Chain | ΔΔGbind | PDB Chain | ΔΔGbind | PDB Chain | ΔΔGbind |
| Y323     | A       | 2.04     | Q383     | B       | 1.25      | K383     | E       | 1.35     |
| K347     | A       | 2.72     | R337     | B       | 1.70      | T430     | E       | 3.56     |
| K383     | A       | 0.92     | K383     | D       | 1.04      | K383     | D       | 1.32     |
| T361     | D       | 1.23     | E385     | D       | 1.01      | K432     | D       | 2.68     |
| K383     | D       | 0.93     |          |         |           |          |         |          |

Note: Chain A, hTLR2; Chain B, hTLR1; Chain D, hTLR10; Chain E, hTLR10*; PDB, position of mutated residues in the pdb file; ΔΔGbind, predicted change in binding free energy upon alanine mutation.

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demonstrated that hTLR2 possesses the di-acylated lipid binding pocket with a volume of \( \sim 1200 \text{ Å}^3 \) and that hTLR1 has a single acylated lipid binding pocket with a volume of \( 400 \text{ Å}^3 \). However, the cavity volume of our hTLR10 model was \( 508 \text{ Å}^3 \). This pocket size diversity is mandatory for discrimination of structurally analogous lipopeptides. Unlike other LRR proteins, the ligand binding cavity is located in the convex region, which lies in between LRR9–12 [31]. The domain-exchange experiments showed the importance of these regions (LRR9–12) in TLR1 and 6, which plays a crucial role in discrimination of a wide variety of ligands.

In addition to the cavity volume, we compared the hydrophobic amino acids lining the lipid binding pocket of hTLR10 (LRR9–12) with the template hTLR1 (LRR9–12). The results showed that out of 18 hydrophobic residues, nine were identical and the rest were replaced by other hydrophobic residues in hTLR10 (Figure 7C). Previous studies have shown the importance of M338 and L360 in TLR1, which play a crucial role in ligand recognition. Substitution of these residues in TLR1 by F343 and F365 in TLR6 results in blocking the ligand binding cavity due to the bulky side chain. However, these important residues are identical in hTLR10 (M338 and L360) and could play a significant role in recognizing a single acyl chain of lipopeptide (Figure 7D).

The results of the present study clearly demonstrate that hTLR10 might be able to accommodate single acyl chains from a tri or di-lipopeptide ligands. This provides some information that may enable identification of the ligand responsible for the activation of the hTLR10/2, hTLR10/1, and hTLR10/10 complexes. Previous studies have shown that hTLR2 can accommodate two ester bound lipid chains [31] and hTLR10 was found to possess amide bound lipid binding chains in the current study, which indicates that the tri-acylated (Pam3CSK4) lipopeptides act as ligand molecules. Additionally, the hTLR10 and hTLR1 central domain (LRR9–12) each possess a single acyl binding cavity, which indicates that neither Pam3CSK4, PamCysPamSK4 nor Macrophage-activating lipopeptide-2 (MALP-2) are responsible for the activation of hTLR10 homo and hTLR10/1 heterodimer complexes. So, we ruled out the possibility of the Pam3CSK4 and MALP-2 due to its nature. These ligands possess two ester bound lipid chains that run parallel to each other and it is not possible to accommodate both hTLR10 homo and hTLR10/1 heterodimer complexes due to the lack of a binding cavity. Whereas, the PamCysPamSK4 structure is different from the above di-acyl peptides, which have acyl chains located at the C1 and C3 atoms of glycerol (Figure S4). We have hypothesized that these two acyl chains can be accommodated by hTLR1 and hTLR10 respectively. We have tested our hypothesis by the docking experiments.

Protein-ligand docking validation of known TLR1/2/ Pam3CSK4 complexes

To evaluate our docking simulation, the crystal structure of hTLR1/2/Pam3CSK4 was downloaded from the PDB and used to conduct two different docking calculations [31]. The dominant clusters from the MOE-Dock and ASEDock docking simulations were found to have the same binding orientation when compared with the original crystal structure. The similarity between the present docked poses (ASEDock yielded bound conformations with the lowest RMSD (0.14), followed by MOE-Dock (0.38)) and the crystal structures shows that our docking protocol was able to reproduce the near native hTLR1/2/Pam3CSK4 complex. Therefore, we consider that our protocol to be trustworthy and we used it in the subsequent docking calculations.

Figure 7. Structural comparison of ligand binding sites. (A) Structural comparison of the lipid-binding pockets of hTLR1 (Blue) and hTLR10 [12]. (B) Superimposition of the TLR6 (Green), TLR10 [12] and TLR1 (Blue): The residues responsible for the channel-blocking F343 and F365 in TLR6 are shown as green sticks and the corresponding residues M338 and L360 of TLR10 and M338+ and L360+ of TLR1 that are not involved in blocking are shown as red and blue sticks, respectively.

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hTLR10 complexes and their ligand interaction

The hypothesized (PamCysPamSK4, Pam-CSK4) ligands were docked into the active site of hTLR10/1, hTLR10/2 and hTLR10/10. A close up view of the PamCSK4 in the hTLR10/2 binding sites revealed an extensive spatial overlap of the predicted best poses, despite the variety of docking methods employed (Figure 8A). The bound conformation of the ligand present in the hTLR10/2 shows that the two ester-bound lipid chains of Pam3CSK4 are inserted in hTLR2, through a crevice formed between the LRR11 and LRR12 loops and interact with its internal hydrophobic pocket, whereas the remaining amide-bound lipid chains are inserted into hTLR10 through the gap between the LRR11 and LRR12 loops and interact with its internal hydrophobic pocket. The conserved cysteiny1 group and glycerol backbone of the lipopeptide is located in the narrow opening formed, where the hTLR10 and hTLR2 pockets join. Their positions are partially fixed by five hydrogen bonds: between the backbone nitrogen of F349 and a carbonyl oxygen of the lipopeptide, between the backbone oxygen of N294 and the backbone nitrogen of K4 in the lipopeptide, between the backbone oxygen of Y313 and the backbone nitrogen of K5 in the lipopeptide, between the carboxamide group of Q316 and the amide oxygen in the lipopeptide and between the Q315 side chain and the carbonyl backbone of K5 in the lipopeptide. The docked complex of hTLR10/1/PamCysPamSK4 shows a slight diverse binding mode in comparison with the TLR2/1/PamCSK4 and mTLR2/6/PamCSK4 complex.

It is important to note that the ester bound lipid chain attached to the glycerol is longer than the amide bound lipid chain. The cavity volume comparison clearly indicates that hTLR10 (518 Å³) is slightly higher than hTLR1 (400 Å³). Hence, the ester bound lipid chain cannot accommodate in hTLR1. As expected, the docking results showed that ester-bound lipid chains are recognized through the binding pocket of hTLR10 and that amide-bound chains are recognized through the binding pocket of hTLR1 (Figure 8B). There are four hydrogen bonds that are formed in the hTLR10/1 binding pocket: between the carboxylic group of D288 and the NH2 group of K4 in the lipopeptide, between the hydroxyl group of Y320 and the amide oxygen in the lipopeptide, between the carbonyl backbone of G313 and the backbone nitrogen of K3 in the lipopeptide and between the carboxamide of Q316 and the amide oxygen in the lipopeptide. Single apostrophes are used for the hTLR10 residues and a plus symbol for the hTLR1 monomer to differentiate them from the ligand.

In case of the hTLR10 homodimer, docking with PamCysPamSK4 revealed that there is a slight difference in the orientation of the ligand recognition site as well as the residual contribution in the binding pocket. The LRR11 and LRR12 loops of both hTLR10 chains are located in the center of the dimerization interface and provide key hydrophobic residues to enable recognition of the lipid chains of PamCysPamSK4 (Figure 8C). The hydrogen bonds present at the interface of hTLR10/10 with PamCysPamSK4 are as follows: between the carboxylic group of Q516 and the amide oxygen in the lipopeptide, between the backbone oxygen of D288 and K4 in the lipopeptide, between the backbone oxygen of Y313 and the backbone nitrogen of K3 in the lipopeptide, between the carboxamide of Q316 and the amide oxygen in the lipopeptide and between the carboxamide of Q315 and K5 in the lipopeptide. The docking studies clearly indicate that LRR9-12 present in the TLR1 family members plays a crucial role in ligand recognition, and that the hydrophobic residues present in that region are important for recognizing the acyl chain. These results are in agreement with the domain-swap experiments [31]. Furthermore, the possible ligands predicted in this study are in agreement with those of a previous study conducted by Hasan et al., who reported that the ligands for TLR1 family such as tri- and di-acylated lipopeptides might be the same for hTLR10 [12].

Taken together, we elucidated the 3D structure of hTLR10 to show its homo and heterodimerization with hTLR1 and hTLR2. Based on these models, we also suggest three possible receptor dimerization schemes that require different minimum ligand sizes. These complexes provided the specific active sites, which enabled us to predict the possible ligands necessary for the TLR10 activation. We also proposed that PamCysPamSK4 and Pam-CSK4 might be possible ligands for the activation of TLR10 signaling pathways. Our models provide a structural framework for interpreting experimental data and allow predictions of the TLR signal transduction process. The presented modeling approach can be extended to other repetitive protein domains.

Supporting Information

**Figure S1** The backbone superimposition of hTLR10 with hTLR2 and hTLR1 are shown as red, cyan and blue, respectively. Found at: doi:10.1371/journal.pone.0012713.s001 (1.75 MB TIF)

**Figure S2** The Ramachandran plot of refined hTLR10, 2 and 1. Found at: doi:10.1371/journal.pone.0012713.s002 (1.85 MB TIF)

**Figure S3** Electrostatic potential on the molecular surface of hTLR10, 2 and 1. The LRR9-17 patches of hTLR10, 2 and 1 are circled. Red and blue indicates negative and positive potential, respectively. The surface potential was calculated and displayed using the PyMOL ABPS tool. Found at: doi:10.1371/journal.pone.0012713.s003 (2.64 MB TIF)

**Figure S4** Chemical structure of lipopeptides used for docking studies. Found at: doi:10.1371/journal.pone.0012713.s004 (0.76 MB TIF)

**Author Contributions**

Conceived and designed the experiments: RGG BM SC. Performed the experiments: RGG BM. Analyzed the data: RGG BM. Contributed reagents/materials/analysis tools: GL SC. Wrote the paper: RGG BM SC.
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