



## Fragment evolution for GPCRs: the role of secondary binding sites in optimization†

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**We developed a docking-based fragment evolution approach that extends orthosteric fragments towards a less conserved secondary binding pocket of GPCRs. Evaluating 13 000 extensions for the  $\beta_1$ - and  $\beta_2$ -adrenergic receptors we synthesized and tested 112 bitopic molecules. Our results confirmed the positive contribution of the secondary binding pocket to both potency and selectivity optimizations.**

With more than 30% of approved drugs, G protein-coupled receptors (GPCRs) belong to one of the major drug target classes.<sup>1</sup> Endogenous GPCR ligands are typically binding in the conserved orthosteric binding pocket (OBP) available from the extracellular side of the cell membrane, and it is a highly tractable pocket for synthetic compounds. Fragment-based approaches, however, might explore both orthosteric and potential allosteric sites and therefore provide easy-to-develop starting points for subsequent fragment-to-lead optimizations. Thanks to the increasing number of available X-ray structures,<sup>2</sup> medicinal chemistry teams can implement structure-based strategies for these optimizations. Here we present a general structure-based protocol for the evolution of aminergic GPCR fragment hits exemplified on the  $\beta_1$ - and  $\beta_2$ -adrenergic receptors.

We have previously investigated the ligand space of the  $\beta_2$ -adrenergic receptor ( $\beta_2$ AR) with two of our approaches for structure-based FBLD. First, we have grown fragments emerging from an *in vitro* screen with our *in silico* structure-aware tool-box PINGUI.<sup>3</sup> Second, utilizing our database of easily synthesizable molecules SCUBIDOO,<sup>4</sup> we implemented a matrix-based strategy

for potency optimization.<sup>5</sup> Here, we merge these two strategies and apply them to a selectivity-driven task. Since fragments usually bind to the strongest hot spots, typically located in the orthosteric pocket, we started from fragments bound in the OBP of both the  $\beta_1$ -adrenergic receptor ( $\beta_1$ AR) and  $\beta_2$ AR with approximately equal affinity and attempted to grow each of them such that the resulting molecule would prefer either the  $\beta_1$ AR or the  $\beta_2$ AR. Using every building block in multiple molecules, this approach leads to “molecular matrices” and facilitates direct comparisons of optimized fragments.<sup>5,6</sup> As we worked on two receptors, there are two such matrices, one for the  $\beta_1$ AR and one for the  $\beta_2$ AR. Our concept was based on the expectation that the less conserved nature of the allosteric secondary binding pockets (SBPs) available in the extracellular vestibule of G protein-coupled receptors (GPCRs) would result in selective bitopic (*i.e.* occupying both OBP and SBP) compounds.<sup>7</sup> In addition to the potentially improved selectivity, the increased number of interactions anticipates improvements in the binding affinity of the generated bitopic compounds. Using two sets of fragments specifically designed for the  $\beta_1$  and  $\beta_2$  SBPs, our further goal was identifying the role of the SPB in both potency and selectivity optimizations. Comparative analysis of the corresponding optimization matrices revealed that SBP-binding moieties mostly improved the binding affinity and to a smaller extent also played a part in compound selectivity. Furthermore, mapping the corresponding SBPs with a limited number of fragments provided detailed information on the activity landscape of the targets and hints about the main selectivity drivers for both of the receptors.

The orthosteric binding pocket of  $\beta_1$ AR and  $\beta_2$ AR is embedded in the transmembrane intrahelical space and binds their endogenous ligands, *i.e.* adrenaline and noradrenaline. This site is highly similar between both receptors: only one residue differs in the OBP, F359<sup>7,35</sup> in the  $\beta_1$ AR is changed to Y308<sup>7,35</sup> in the  $\beta_2$ AR (Fig. S1, ESI†). Both receptors harbor an allosteric secondary binding pocket (SBP) located in the extracellular vestibule<sup>9</sup> containing three divergent residues: V360<sup>7,36</sup>, D356<sup>7,32</sup> and I118<sup>2,64</sup> in the  $\beta_1$ AR are replaced with I309<sup>7,36</sup>, K305<sup>7,32</sup> and H93<sup>2,64</sup>, respectively, in the  $\beta_2$ AR. The last two have drastic physicochemical property differences which makes the

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SBP a sound target area for exploring its role in the selectivity between the two receptors.<sup>8</sup>

First, we selected the core OBP binding fragment set. Docking 387 707 primary amine fragments (*cf.* ESI,† Methods) to the OBP of both the  $\beta_1$ AR and  $\beta_2$ AR was done using FRED.<sup>9</sup> This yielded 238 fragments which were then carefully visually inspected for their binding modes and interactions formed with the receptors (*e.g.* direct interactions with D113<sup>3,32</sup> and N312<sup>7,39</sup> were mandatory). This evaluation resulted in 32 fragments that were subjected to synthetic feasibility and availability checks. Finally, 21 OBP fragments were ordered and tested experimentally (Table S1 and Fig. S2, S4, ESI†).

In this study we combined the structure-based evolution of the core fragments with the power of SAR-driven optimization using the concept of ligand efficiency (LE =  $pK_D$ /number of non-hydrogen atoms in the molecule).<sup>10</sup> Out of the 21 fragments tested, 13 (62%) showed a ligand efficiency (LE) above 0.4, which is a remarkably high fraction (see ESI,† Table SI\_assays.csv). The five fragments (A1–A5) with the best LE (Fig. S3, ESI†) and no significant preference for either target were selected for further growing. Interestingly, some compounds (*e.g.* A11 and A12) showed a more noticeable fold selectivity of 9.8 and 9.3, respectively, suggesting that the OBP is involved in ligand selectivity. Curiously, none of the assayed fragments showed higher affinity for the  $\beta_1$ AR than for the  $\beta_2$ AR.

For the selection of suitable fragments with affinity towards the secondary binding site, we first defined surrogates. Considering that reductive alkylation was selected as a robust reaction for growing the orthosteric fragments, our surrogates were derived from aldehyde building blocks by replacing the aldehyde moiety with a primary amine group. This procedure allows us to mimic the interactions of the final product during the initial docking calculation in which the surrogate is placed inside the SBP. The 7893 surrogates (*cf.* ESI,† Methods) were individually docked to the SBP of both receptors. For each receptor, the top 1000 surrogates according to FRED score were kept after ensuring that they did not overlap with the predicted core fragment position in the OBP. As we intended to explore the receptor-specific affinities and the selectivity, surrogates ranked favorably in both receptors were removed, yielding a set of 598 preferred surrogates for each receptor (*i.e.* 402 surrogates were found in both receptors in the top 1000 ranked molecules).

For each OBP fragment to be grown, the corresponding derivative products were generated by reacting it *in silico* with each of the 598 surrogates selected in the previous step using the specified chemistry of reductive alkylation. Docking-derived binding modes and receptor interactions were carefully inspected visually, furthermore ensuring pose fidelity, *i.e.* that the binding modes of the SBP surrogate and the core OBP fragment did overlay consistently with the pose of the corresponding product. This analysis left 13 and 16 derivatives for the  $\beta_1$ AR and  $\beta_2$ AR, respectively. All the key steps of the computational workflow are presented in Fig. 1.

Next, synthesis of bitopic compounds was realized by the robust reductive alkylation reaction (Fig. S15, ESI†), which shows broad substrate and functional group tolerance and is only little sensitive to reaction parameters.<sup>11</sup> Moreover, reductive alkylation results in a protonable amine in the molecule, which is a precondition for

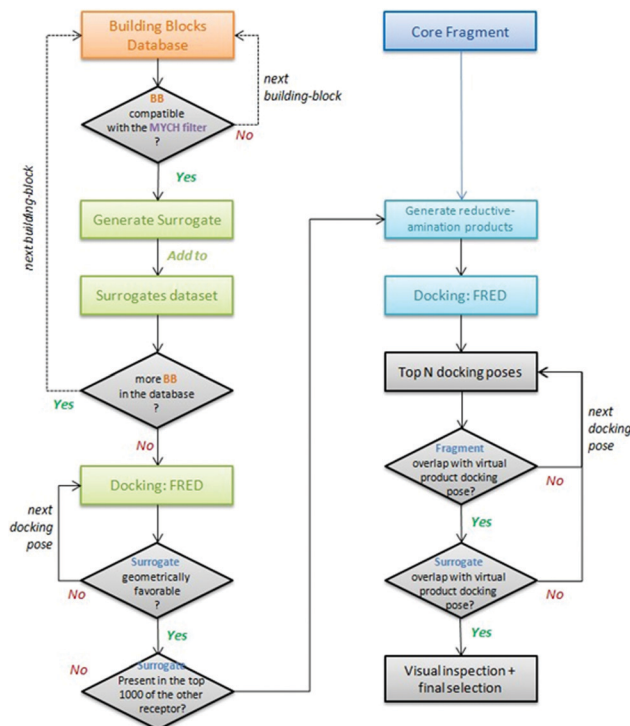
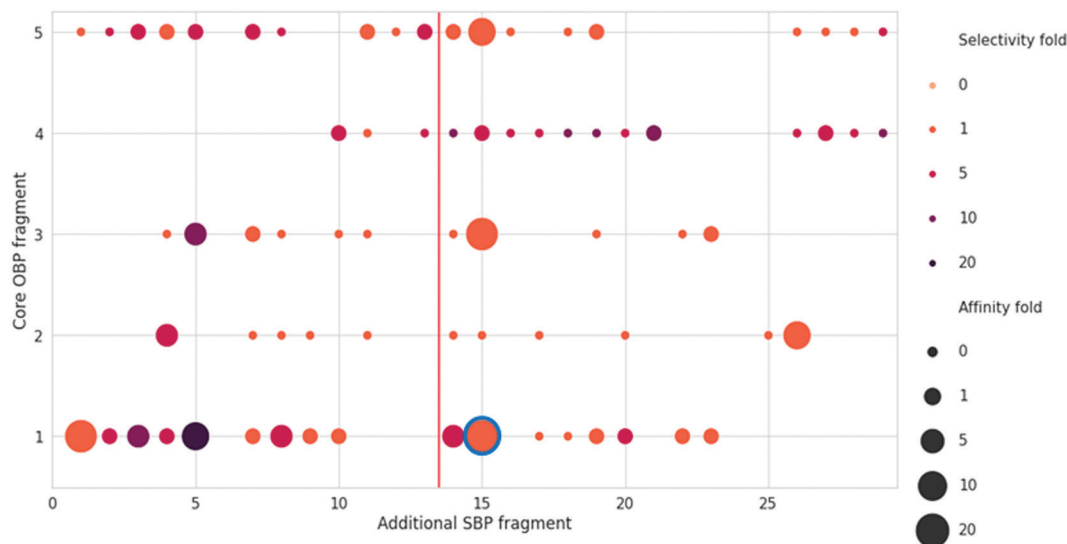


Fig. 1 Workflow depicting all required steps to select the building blocks and build the final molecular matrices.

binding to the  $\beta$ ARs. One hundred twelve out of the 145 designed bitopic products (Fig. S5–S12, ESI†) were synthesized successfully (62% and 93% of compounds designed for  $\beta_1$ AR and  $\beta_2$ AR, respectively) and were submitted to biological assays.

All 112 synthesized molecules were assessed for affinity in the <sup>3</sup>H-CGP12177 whole cell binding assay (Fig. 2 and Fig. S5–S9 and SI\_assays.csv for summary data, ESI†). Fig. 2 suggests that the core compounds behaved differently when combined with the SBP fragments. For example, core fragment A4 retained affinity for both receptors when the  $\beta_2$ -SBP bitopic compounds were examined, however when the same core was combined with the  $\beta_1$ -SBP fragment extensions, a loss in affinity was observed in many cases (Fig. 2). When looking at both matrices together, only one out of the 29 SBP fragments (3%) did not yield any active compound against its designed target receptor (B24), while 18 SBP fragments (62%) showed at least one product improving the initial affinity of its core OBP fragment (Fig. 2 and Fig. S13, S14, ESI†). Overall, the accuracy for the  $\beta_1$ -SBP bitopic compounds (*i.e.* compounds with improved affinity against  $\beta_1$ AR) was 34%. It was 23% for  $\beta_2$ -SBP bitopic compounds. In the  $\beta_2$ AR-specific matrix (Fig. S10, ESI†), when compared to their respective core OBP fragments, 15 out of the 60 (25%) synthesized compounds showed an increase in affinity against the  $\beta_2$ AR and 35 compounds (58%) showed a decrease in affinity against the  $\beta_1$ AR. Compound A1B15 ( $K_D$  = 35 nM) displayed the highest affinity gain of 69-fold of any compound. The addition of certain SBP-fragments resulted in a lower affinity for all cores *e.g.* B16, B17 and B18. In contrast, for the  $\beta_1$ AR matrix (Fig. S11, ESI†),





**Fig. 2** Bubble plot representation of the fold selectivity and the fold affinity improvement ratio for the core OBP fragments and the SBP fragments in the  $\beta_1$ AR. Ratio below 1 indicates a property deterioration while a ratio above 1 indicates a property gain. The vertical red line separates the  $\beta_1$ AR matrix (*i.e.* compounds designed for higher  $\beta_1$ -affinity/selectivity) (left) from those designed for the  $\beta_2$ AR matrix (right).

addition of **B1** to core **A1** caused the largest increase in affinity – an increase of 9.1 and 27-fold over the core compound **A1** at the  $\beta_1$ AR and  $\beta_2$ AR, respectively. For the  $\beta_2$ AR matrix, it was the addition of fragment **B15** that resulted in the greatest increase of affinity compared to the core alone. Thus, the addition of **B15** increased the affinity of core **A1** by 27- and 69-fold (molecule **A1B15**), 20- and 21-fold (**A3B15**) and 13- and 11-fold (**A5B15**) at the  $\beta_1$ AR and  $\beta_2$ AR, respectively. The molecule with the highest overall affinity of 7 nM at the  $\beta_2$ AR was **A4B27**, however with only marginal increases of the affinity at the  $\beta_1$ - and  $\beta_2$ AR (2.5- and 5-fold, respectively) over the core fragment.

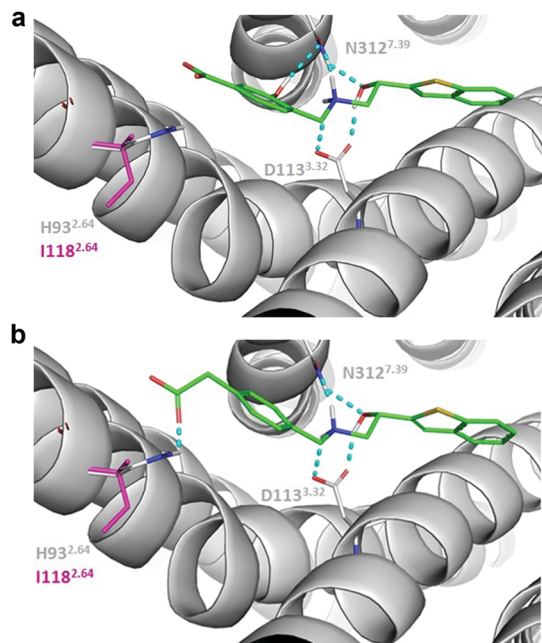
In the  $\beta_2$ AR-specific matrix, compound **A4B14** showed the most pronounced selectivity for the  $\beta_2$ AR (14-fold). Curiously, it was one of the compounds designed for the  $\beta_1$ AR, **A1B05**, that showed the highest absolute  $\beta_2$ AR-selectivity overall (*i.e.* from either matrix). Compound **A4** yielded the series with highest  $\beta_2$  preference (maximum fold selectivity = 14.1, mean fold selectivity = 8.2). This can be explained by the fact that **A4** was the core OBP fragment with the highest initial fold selectivity (4.5). Interestingly, compound **B21**, which contains a carboxyl moiety (*i.e.* predicted to interact with K305<sup>7.32</sup> or H93<sup>2.64</sup> in the  $\beta_2$ AR), yielded three products with appreciable selectivity (fold selectivity > 10). These outcomes indicate that some degree of selectivity can be achieved *via* the SBP by targeting K305<sup>7.32</sup> or H93<sup>2.64</sup>, but that the initial OBP contribution is vital for the total achievable selectivity.

While it has to be admitted that the overall selectivity trends are moderate, the analysis of several pairs still provides illuminating details about the molecular origins of affinity and selectivity between the  $\beta_1$ AR and  $\beta_2$ AR. Compounds **A1B14** and **A1B17** (Fig. S18, ESI<sup>†</sup>) share the same building block in the OBP, and both are mildly  $\beta_2$ -selective. The building blocks in the SBP, **B14** and **B17**, differ by the heteroatoms in the five-membered ring (triazole *vs.* thiophene) and the distance

between the ring and the carboxyl group (one carbon *vs.* direct attachment). The shorter fragment yields the less potent compounds in both receptors, indicating that the interaction entertained by the carboxyl group lies further away from the OBP. This is likely to be close to position 2.64, which is a His in the  $\beta_2$ AR, thus also explaining the slight preference of the compound for this receptor. Comparative analysis of the corresponding matrix revealed that we have basically two situations to consider. In the first one, small changes of the OBP-binding fragment influenced the affinity of the bitopic compounds (compound **A1B15** *versus* **A2B15**). Here we hypothesize that the changed binding mode of the orthosteric fragment points the growing vector to a different part of the SBP. In the other case, changes of the SBP-binding fragment position the identical OBP-binding fragment differently (compound **A1B14** *versus* **A1B17**), which in turn influences the affinity of the bitopic compounds. This highlights the complexity of fragment growing procedures, where subtle changes, such as a slight increase in steric constraints, might have a tremendous impact on the resulting binding mode. This observation also emphasizes the utility of the matrix approach in FBDD optimization programs.

In terms of selectivity, an instructive pair is **A1B18** and **A1B21** (Fig. 3, Fig. S19, ESI<sup>†</sup>). In the docking poses, the hydroxy group on the phenyl ring of **B18** interacts with the conserved Tyr<sup>7.43</sup>, while the carboxyl group can't quite reach to position 2.64 and thus presumably interacts with a water molecule. This explains the relatively similar affinities in the  $\beta_1$ AR and  $\beta_2$ AR. When **B21** is installed, there are two changes: first, the hydroxy group is missing, second, there is an additional carbon between phenyl ring and carboxyl group. This, similar to what was observed for building blocks **B14** and **B17**, pushes the carboxyl group more towards residue 2.64. The resulting favorable interaction with the positively charged His at this position in the  $\beta_2$ AR seems to compensate for the loss of the interaction





**Fig. 3** Binding mode prediction of (a) **A1B18**, (b) **A1B21** in the  $\beta_1$ AR (residues in magenta) and the  $\beta_2$ AR (residues in light grey) binding sites. Ligands are shown with green sticks and polar contacts are represented with cyan dashed lines.

with Tyr<sup>7.43</sup>. Conversely, given that residue 2.64 is an Ile in the  $\beta_1$ AR, the resulting proximity between a charged moiety and an apolar side chains results in complete loss of binding of **A1B21** to this receptor. These results together with other pairs identified in the  $\beta_1$ AR and  $\beta_2$ AR matrices suggest that the selectivity of bitopic compounds is influenced by both of the pockets *i.e.* a more complicated picture than the one implicated previously.<sup>7,12,13</sup> Our analysis revealed that the optimization of selectivity needs the identification of a suitable combination of OBP- and SBP-binding fragments. The matrix-based methodology suggested here can contribute to effectively finding such combinations.

Although we explored SAR and selectivity trends effectively, we achieved mild selectivity for our  $\beta_2$ AR designs at best. What could have been the reasons for this limited selectivity? The primary reason is certainly that the two receptors have very similar binding pockets. The OBP is virtually the same except for the Phe to Tyr change at position 7.35, and substantial exploitable differences occur only at the far end of the SBP. A second reason might be that we used a homology model of the human  $\beta_1$ AR, as the X-ray structures available at the start of this project were exclusively from turkey. Last, but not least, selectivity in this receptor system might not only be due to direct ligand–receptor interactions. Different water networks, polarization effects and subtle conformational changes of the receptor upon ligand binding, to name just a few, might play substantial roles. However, these phenomena are clearly beyond our approach, which used docking to rigid receptors.

Our study yielded insights into structure–activity and structure–selectivity relationships and demonstrated the power of

using optimization matrices as a design concept that ensures efficient resource usage while maximizing the information obtained from experiments. Paired with high hit rates, such methods can support hit expansion in the lead discovery projects. Investigating the impact of the orthosteric and secondary binding pockets on the binding affinity and selectivity, we found that the optimization is influenced by both sites. While binding affinity was less sensitive to the SBP interactions, our data suggests that the optimization of selectivity needs the identification of bitopic compounds constructed from suitable OBP and SBP pairs. This goal can be achieved efficiently by the optimization matrix approach used here. Providing detailed SAR information, this strategy maps both of the binding pockets effectively and suggests viable growing vectors for optimizing OBP-bound fragment starting points.

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## Conflicts of interest

There are no conflicts to declare.

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