MECHANISMS OF MODULATION OF
NICOTINIC ACETYLCHOLINE RECEPTORS

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MECHANISMS OF MODULATION OF
NICOTINIC ACETYLCHELINE RECEPTORS

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Abstract

Inappropriate expression of nicotinic acetylcholine receptors in the central nervous system is associated with nicotine addiction, Alzheimer’s disease, Parkinson’s disease and other disorders. Modulators (drugs) have the potential to restore circuit properties that arise from inappropriate expression of nicotinic receptor’s. Compounds that interact with allosteric sites have a distinct advantage over agonists and partial agonists, in that, they retain normal activation patterns by allowing binding of the endogenous ligand. Positive allosteric modulators boost the receptors ability to respond to agonist. Studies of these modulators constitute a first step toward the identification and development of better compounds that minimize signaling errors at cholinergic synapses. We have used single molecule methods to investigate the action of a novel positive allosteric modulator, desformylflustrabromine (dFBr), on nicotinic receptors. Our studies were focused on the α4β2 subtype of nicotinic receptors in the brain. These receptors exist in two forms with low sensitivity (α43β22) or, alternatively, high sensitivity (α42β23) to agonist. Our experiments allowed us to develop detailed gating models for high and low sensitivity receptors, as well as gain new insights regarding the mechanisms that underlie potentiation by allosteric modulators. We found that dFBr potentiates low sensitivity receptors by destabilizing desensitized states of the receptor. In contrast, potentiation of high sensitivity receptors arises from a synchronization of openings following an application of agonist due to an increase in the opening rate. Based on our results we now have a better understanding of the advantages of dFBr on high and low sensitivity receptors.
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Chapter 1: Overview and Research Aims

1.1 Background

1.1.1 General Overview of Synaptic Transmission

At chemical synapses, invasion of the axon terminal by the action potential results in opening of voltage gated calcium channels and calcium influx. Elevated levels of calcium ions trigger fusion of synaptic vesicles that contain neurotransmitter with the presynaptic membrane. Neurotransmitter is released into the synaptic cleft and diffuses a short distance (~30 nm) where it binds to postsynaptic receptors. Binding of neurotransmitter biases the receptor to open, allowing cations to enter the post synaptic cell. If a threshold is reached in the membrane potential, then an action potential is produced. Activation of excitatory receptors leads to cation influx resulting in depolarization of the postsynaptic membrane (Albuquerque et al., 2009).

Neurotransmitter receptors affect signals that are sent from one neuron to another and therefore harbor an opportunity for intervention by drugs. Inappropriate expression of nicotinic receptors in cholinergic systems is associated with diseases and disorders such as nicotine addiction and Alzheimer’s disease. Nicotine addiction is a complex disorder, but upregulation and consequent overexpression of nicotinic receptors in the brain is a primary contributing factor in the reinforcing nature of nicotine addiction (Fenster et al., 1999; Buisson and Bertrand, 2001). While nicotine addiction is thought to occur by an overexpression of receptors, the loss of cholinergic receptors is likely to contribute to cognitive decline in Alzheimer’s disease and other
neurodegenerative disorders. Alzheimer's disease emerges when there is underexpression of nicotinic receptors in the brain, with cell loss and loss of receptors being correlated to a loss of cognitive abilities in patients with the disease (Nordberg and Winblad, 1986). If there are too few receptors on postsynaptic cells, then signals are too weak, and neurons are unable to effectively communicate. If too many receptors are present and are overexpressed, then the resulting signals can become too strong. In scenarios where circuits are not working properly due to over- or underexpression of nicotinic receptors, then a strategy is needed to alter receptor responsiveness.

Two competing strategies for correcting errors in cholinergic signaling include the use of partial agonists or positive allosteric modulators. While partial agonists (Coe et al., 2005) are one strategy for modulation of nicotinic receptors, they are an improper choice due to competition with acetylcholine for binding. Continuous exposure of receptors to agonist and partial agonists can produce receptor desensitization, resulting in receptors being unavailable for further activation. Alternatively, a compound that interacts with allosteric sites can be utilized for correcting cholinergic signals, allowing retention of the normal activation pattern of the endogenous agonist. Positive allosteric modulators are compounds that are unable to or produce little receptor activation on their own, and instead boost the receptors ability to respond to agonist. These compounds have a potential advantage in not only preventing drug induced desensitization, but also by restoring normal levels of acetylcholine activity, yielding better functional outcomes compared to partial agonists.
Given that inappropriate expression $\alpha 4\beta 2$ receptors are associated with a variety of CNS disorders an effective therapeutic intervention would require a library of allosteric modulators with unique functional properties. Selective positive allosteric modulators are important in the development of compounds with specific functional properties. Studying the mechanism of action for positive allosteric modulators helps cultivate an understanding of the effects a drug has on a cholinergic synapse. Establishing a model constitutes a first step toward the development and identification of better compounds that can correct errors in cholinergic signaling. In this thesis we investigated the action of a particular modulator called dFBr. A long term goal of these studies is to facilitate the directed development of new and better compounds.

1.1.2 Characteristics of $\alpha 4\beta 2$ Receptors

In mammals, neuronal nAChRs are transmembrane proteins with a pentameric arrangement of subunits arranged around a central pore. Neuronal nAChRs are formed from a collection of nine $\alpha$ ($\alpha 2-\alpha 7,\alpha 9,\alpha 10$), and three $\beta$ ($\beta 2-\beta 4$) subunits (Changeux, 2010). The major nAChRs within the brain are homomeric receptor composed of $\alpha 7$ subunits and heteromeric receptors composed of $\alpha 4$ and $\beta 2$ subunits (Khiroug et al., 2004; Moore and McCarthy, 1995). The heteromeric receptors can assemble in alternate forms alternate forms, low sensitivity (LS) receptors consisting of three alpha and two beta subunits ($\alpha 4_3\beta 2_2$) that bind ACh and nicotine with low affinity (EC$_{50}$ ~100 $\mu$M), and high sensitivity (HS) receptors consisting of two alpha and three beta subunits ($\alpha 4_2\beta 2_3$) that bind ACh and nicotine with high-affinity (EC$_{50}$ ~1$\mu$M) (Nelson et al., 2003;
Khiroug et al., 2004; Buisson et al., 1996). High sensitivity receptors require a relatively low concentration of agonist in order to produce a half maximal response, (low EC$_{50}$). Low sensitivity receptors require higher concentrations in order to elicit a half maximal response. Isolated $\alpha_{4}\beta_{2}$ receptors display two distinct conductance's, $\sim$20 pS for the HS subtype and $\sim$30 pS for the LS subtype (Charnet et al., 1992; Buisson et al., 1996; Li and Steinbach, 2010).

Most work on the physiological properties of $\alpha_{4}\beta_{2}$ nAChRs has been done in oocytes and HEK 293 cells. When equivalent amounts of $\alpha$4 and $\beta$2 cDNAs are injected into oocytes, both HS and LS receptors are expressed in equivalent amounts, whereas in HEK293 cells the receptors predominantly assemble as the LS subtype, comprising 80% of the receptor population (Harpsøe et al., 2011). Using HEK293 cells is an advantage, in that, it allows us to compare our single-channel data to whole-cell data that has been previously obtained from studies utilizing this cell line. This allows us to determine whether our models are supported by whole-cell data.

nAChRs binds agonists such as the endogenous agonist ACh or exogenous agonist nicotine (Changeux, 2010). In the absence of agonist, nAChRs are in a closed, nonconductive state. Binding of ACh leads to channel opening and cation influx, however, in the continued presence of agonist receptors become unresponsive. This phenomenon is called desensitization. If agonist is removed when receptors are mostly open, then receptors have the chance to close (deactivate) allowing them to be available to immediately reopen upon exposure to agonist. In contrast, when receptors are exposed to extended periods of elevated levels of agonist then receptors are not
immediately available for reactivation due to being desensitized. Once desensitized receptors require time to return to a resting state.

Nicotinic receptors play interesting and unusual roles within functional circuits, with some being activated quickly, while others are slower. In classical transmission the concentration profile is fast, in this situation it is unlikely that nAChRs will desensitize. If the concentration profile is fast then receptors will open for a short time, ACh will unbind and the receptors will close. In volume transmission model, agonist could remain elevated for extended periods. In this scenario diffusion of acetylcholine through extracellular fluid would activate extrasynaptic nAChRs (Agnati et al., 1995). Delaying transmitter removal gives receptors the opportunity to enter desensitized states.

While the precise functional roles of nAChRs in cholinergic transmission are unclear, reduction in the total number of available nicotinic receptors has been linked to a wide variety of neurodegenerative disorders and diseases including: schizophrenia, Parkinson’s disease, autism, Alzheimer’s disease, and PTSD (Breese et al., 2000; Terry et al., 2015; Dome et al., 2010; Hajas and Rogers, 2010; Dani and Bertrand, 2007). Alzheimer’s is a catastrophic disease accompanied by a loss of cholinergic cells within the brain, and reduced expression levels of α4 and α7 containing nAChRs (Martin-Ruiz et al., 1999; Guan et al., 2000; Whitehouse and Au, 1986; Marutle et al., 1999).
1.1.3 *Diseases Associated with Cholinergic Transmission*

1.1.3.1 *Alzheimer’s Disease*

Alzheimer’s disease (AD) is a neurodegenerative disorder with patients eventually showing a loss in cognition and control of daily activity over time. β-amyloid (Aβ) is an important factor in the initiation and promotion of AD, and soluble Aβ42 oligomers are currently thought to be the primary cause of the disorder. Investigations into the role of Aβ in AD patients show significant reductions in the number of nAChRs (37 to 57%) (Marutle et al., 1999). The degree of high affinity nicotinic agonist binding in the temporal cortex is correlated to the severity of AD, with a deficiency of high affinity nAChR binding sites linked to a loss of cognitive abilities (Wevers et al., 1999; Martin-Ruiz et al., 1999; Guan et al., 2000; Whitehouse and Au, 1986; Nordberg and Winblad, 1986).

A possible treatment option for alleviating the symptoms in AD patients is to elevate the concentration of ACh within the brain by using compounds that inhibit acetylcholinesterase (ChE) (Christner et al., 2003; Paterson and Nordberg, 2000). Galantamine is a ChE inhibitor that also interacts directly with nAChRs. While ChEs like galantamine improve cognitive function, the numerous peripheral and central nervous system side effects commonly associated with ChE compounds outweighs their benefits (Christner et al., 2003). These side effects coupled with ChEs being unable to prevent or reverse the progression of AD, means other options may be more suitable to address nAChR dysfunction in the disease. Additional treatment options include the development of selective nAChR agonists, partial agonists or positive allosteric
modulators (PAMs). The agonist nicotine has been shown to be beneficial for many neurodegenerative disorders including AD, but the addictive properties of nicotine make this compound unsuitable as a treatment option (Levin and Rezvani, 2000; Rusted and Warburton, 1992; Jones et al., 1992). The partial agonist varenicline has recently been shown to disrupt the action between fibrillary Aβ and α7 nAChRs, due to a possible overlap in binding sites (Ni et al., 2013). While there is interest in agonists and partial agonists as a potential treatment option for individuals with AD, these choices offer little in the area of selectivity (Pandya and Yakel, 2011). Compounds such as PAMs are better suited for boosting nAChR function in disorders such as AD due to their ability to interact with a site distinct from ACh, thus potentially yielding better functional outcomes. Desformylflustrabromine is a PAM for the α4β2 receptors (Sala et al., 2005) has promise as a future therapeutic option in the treatment of AD. Along with the compounds observed potentiation on α4β2 receptors, it has an additional unique property of being able to interfere with the inhibitory action of Aβ1-42 with α4β2 receptors (Pandya and Yakel, 2011).

1.1.3.2 Nicotine Addiction

Tobacco use and nicotine addiction are serious risk factors in the development of cancer, heart and respiratory diseases. Over 6 million deaths each year are a result of tobacco use, 5 million of those deaths are the result of direct tobacco use and over 600,000 deaths due to secondhand smoke (World Health Organization, 2015). The primary compound responsible for strong addiction to smoking is nicotine.
Nicotine is an effective nicotinic receptor agonist that displays relatively high affinity for nAChRs, exhibiting an EC₅₀ between 2-14 μM (Buisson et al., 1996; Buisson and Bertrand, 2001; Paradiso and Steinbach, 2003). Nicotine addiction is thought to be the result of profound desensitization, triggering upregulation of α4β2 receptors. In α4β2 nAChRs ACh and nicotine are similar with regards to how fast desensitization develops, but receptors stay desensitized for longer in the presence of nicotine (Paradiso and Steinbach, 2003). nAChRs are able to increase their expression (upregulation) when chronically exposed to nicotine, the receptor exhibiting the greatest upregulation being α4β2. The underlying mechanism for upregulation is not agreed upon, however, it is universally thought that upregulation of high affinity receptors is a main contributing factor in the reinforcing nature in nicotine addiction (Staley, 2006; Buisson and Bertrand, 2001; Fenster et al., 1999).

Nicotine can influence the mesocorticolimbic dopamine system (MDS), a dopaminergic pathway that leads from the ventral tegmental area (VTA) to the nucleus accumbens (NAcc) and limbic systems, the primary reward system of the brain. Figure 1.1 is a visual guide to interactions of some neurons in the VTA and where various subtypes of nAChRs are presumed to be located. Nicotine’s influence leads to increased levels of extracellular dopamine in the ventral striatum and pre-frontal cortex. Although the MDS is composed of many sections, nAChR’s expressed on dopaminergic cells and inputs into these cells contribute to the rewarding effects of nicotine (Dome et al., 2010; Mansvelder et al., 2002, 2007). Nicotine administration into the VTA results in an increase in dopamine release into the NAcc. This increase is due to the activation of α4β2 receptors on dopaminergic and GABAergic cells, and α7 nAChRs on
glutamatergic nerve terminals. Exposing $\alpha 4\beta 2$ receptors on the GABAergic and
dopaminergic terminals to nicotine causes receptors to undergo rapid desensitization
(Dome et al., 2010; Mansvelder and McGehee, 2000). While $\alpha 4\beta 2$ receptors are
heavily affected by nicotine, the concentration is not high enough to desensitize $\alpha 7$
nAChRs on glutamatergic inputs into the VTA. Thus, nicotine can activate the
glutamatergic nerve endings, further triggering the dopaminergic cells of the VTA.
Dopaminergic neurons are under tonic inhibitory control by GABA$_A$ and removal of this
inhibition leads to increased activity of dopaminergic neurons, which in turn leads to an
increase in dopamine release in the NAcc (Dome et al., 2010; Mansvelder et al., 2007;
Mansvelder and McGehee, 2000; Mansvelder et al., 2002).

The primary contributor for the pleasurable effects of nicotine is the
desensitization of $\alpha 4\beta 2$ receptors, leading to disinhibition of dopaminergic neurons
within the VTA. Once nicotine dependence has developed it requires nicotine
reinforcement and is difficult to quit. The only smoking cessation treatments currently
available are varenicline and nicotine replacement therapies. Other compounds have
the potential to be useful, such as cytisine, bupropion, NS9283 and dFBr (Mohamed et
al., 2015). While varenicline and cytisine have been found to be as effective as nicotine
replacement therapies for nicotine addiction, very little research has been done on the
effectiveness of PAMs. Whether these compounds will disrupt nicotine specific
desensitization and aid in smoking cessation remains to be seen, with more research in
this area being required.
The benefit of PAMs is that they occupy binding sites distinct from that of acetylcholine, thereby altering the receptors responsiveness to the endogenous ligand. PAMs have the ability to increase the probability of channel opening, while potentially decreasing the agonists potential for receptor desensitization (Williams et al., 2011; Pandya and Yakel, 2011; Hajas and Rogers, 2010). PAMs demonstrate a wide variety of modulatory actions on α4β2 receptors and can potentiate maximal amplitude responses differently. NS206 and dFBr are PAMs that affect both forms of α4β2 nAChRs. However, both potentiate max amplitude responses in a different manner. NS206 is thought to increase efficacy by an increase of the mean open time of the receptor for both stoichiometries, whereas our single channel analysis of dFBr found that it potentiation occurs by destabilizing desensitized states for the LS subtype, and increasing mean open time for the HS subtype (Olsen et al., 2013). PAMs cannot produce receptor activation on their own and can affect ligand gated ion channel function in a couple of ways, by either influencing channel conductance or by altering channel gating. (Williams et al., 2011; Pandya and Yakel, 2011; Hajas and Rogers, 2010).

A PAM can increase the apparent affinity (potency) on the receptor for agonist, while simultaneously not increasing the maximal response amplitude (efficacy). A modulator can be a PAM, having an effect on one of those properties while simultaneously not affecting the other. Several PAM’s of α4β2 nAChRs currently exist.
and affect the channel in a variety of ways. Some are selective for one subtype of α4β2, while others affect both subtypes differently.

1.1.4.1 NS9283

Preclinical studies have shown that NS9283 shows promise in the treatment of cognitive disorders (Timmermann et al., 2012; Lee et al., 2011). NS9283 is a PAM that selectively increases ACh potency for the LS subtype of nAChRs. It is thought to exert potentiation predominantly through reduction in the rate of deactivation of the receptor with the mechanism of action bearing resemblance to that of benzodiazepines at the GABAA receptor. With ultra-short applications of high concentrations of (1 mM) ACh, current decay is substantially slower with NS9283 (Grupe et al., 2013). The PAM does not have an effect on desensitization produced by ACh and was incapable of reactivating desensitized receptors (Grupe et al., 2013). The mechanism of NS9283 is thought to not involve a decrease in macroscopic desensitization, but rather via slowing the rate at which the receptor deactivates. However, we have shown for simulations involving dFBr that the deactivation time course of LS receptors is not a simple reflection of channel closing, but also involves entry into desensitized states. Therefore, the mechanism of modulation for NS9283 could involve the destabilization of a desensitized state, resulting in what appears to be a slowing of deactivation. The presence of NS9283 in the synaptic cleft would prolong the agonist-bound receptor state and ultimately result in an increase in charge transfer. Other PAMs, like dFBr, exhibit potentiation via an increase in mean open time, as well as decreasing ACh-induced macroscopic desensitization (Sala et al., 2005; Weltzin and Schulte, 2010).
Desformylflustrabromine (dFBr)

Desformylflustrabromine is an indolic alkaloid isolated from the marine bryozoan Flustra foliacea and was identified as a selective PAM of α4β2 nAChRs (Sala et al., 2005). dFBr has the ability to potentiate ACh-induced responses at concentration less than 10 μM and inhibit responses at concentrations greater than 10 μM. In the presence of dFBr, dose response curves show a decrease in the EC$_{50}$ and an increase in the maximal response curve. dFBr exhibits an EC$_{50}$ of 120 nM and IC$_{50}$ of 150 μM, and the observed inhibition invoked by dFBr is thought to be due to the displacement of ACh binding (Kim et al., 2007). Rebound currents on washout of ACh/dFBr were seen, indicating the presence of open channel block. Rebound currents occur when the open channel blocker is released while agonist is still bound to the receptor (Kim et al., 2007).

Studies on the therapeutic potential of dFBr are still in their early stages, but a rat self-administration study showed that dFBr has qualities that would make it an ideal candidate for smoking cessation treatments. Pretreatment of dFBr, as well as treatment while rats were engaged in self administration, reduced long term nicotine self-administration behavior. It was also found that dFBr was present in cerebral spinal fluid at one third the concentration of that within the bloodstream of administered rats. This provides evidence that dFBr crosses the blood brain barrier, but at one-third the concentration found in the bloodstream (Liu, 2013). Effective transit across the blood brain barrier is important if a compound is to be considered for smoking cessation therapies.
The in vitro effects of dFBr have been previously studied, but the mode by which dFBr selectively potentiates α4β2 nAChRs remains to be determined. The existing single-channel information on the effects of dFBr are limited and the information that is available briefly demonstrates that dFBr increases the channel-opening probability. It is speculated that the effects of dFBr occur because of an increase in the rate constants designated for channel opening and closing (Sala et al., 2005). A complete understanding of the mechanism of modulation is unavailable for the individual receptor subtypes. Gaining an understanding of the mechanism would improve our understanding of the potential effects dFBr has on a cholinergic synapse as well as aid in the development of future allosteric modulators. The selectivity that dFBr demonstrates makes it a valuable asset for further investigation involving modulation of α4β2 receptors.
1.1.5 Modeling

1.1.5.1 Classical Mechanisms

The earliest functional models for nAChR kinetics were made by del Castillo and Katz in 1957 and described the similarity between enzyme-substrate and receptor ligand interactions (Figure 1.2A) (Castillo and Katz, 1957). This model was further elaborated on by Katz and Theslaff in 1957 to account for desensitization and led to a cyclical like scheme (Figure 1.2B) (Katz and Thesleff, 1957). This cyclical scheme was expanded to include two open states, and is one of the more common mechanisms utilized for analyzing kinetic data (Edelstein et al., 1996).

In 1965, Monod-Wyman-Changeux (MWC) put forth an allosteric transitions theory in order to explain cooperative ligand binding observed in oligomeric proteins. This approach has had a large effect on the understanding of the cooperative interactions between hemoglobin and allosteric enzymes and has further been extended to neuronal transmembrane receptors. Allosteric theory has several components to it: (i) allosteric proteins are oligomers where the individual subunits are equivalent; (ii) each subunit contains a site for a ligand to bind to it; (iii) subunit conformations are restricted based on the interactions with the subunits around them; (iv) allosteric oligomers have access to two conformational states, O and C; (v) binding of ligand to one site alters the affinity of additional sites for ligand; (vi) molecular symmetry is conserved when an oligomer goes from one conformational state to another (Monod et al., 1965). Typical MWC model is shown in Figure 1.2C. Typical nAChR mechanisms, as shown in Figure 1.3A, give adequate fits with the fit often being improved by adding
an additional shut state following the open state, like a desensitized state. Our nicotinic HS and LS receptor mechanisms use the MWC model as a base.

1.1.5.2 Models Established for the HS and LS α4β2 Receptors

Determining the mechanisms of modulation for α4β2 nAChRs in the presence of dFBr is important for enhancing our understanding of the potential impact allosteric modulators could have on a cholinergic synapse. In this thesis we utilized single molecule methods to investigate the mechanism by which dFBr potentiates nAChRs. We provide the first understanding of the mechanism for the two α4β2 nicotinic receptor subtypes. We have developed full gating models that provide a fairly accurate representation of function for both forms of the receptor, LS (Figure 1.3B) and HS (Figure 1.3C). These models were used to perform simulations and determine the effects dFBr would have under a variety of agonist concentration profiles. The mechanisms of modulation are determinant on how the receptor subtypes are expressed within a system, with receptors changing their agonist concentration profile when they are modulated.
1.2 Research Aims

The focus of my research is establishing gating models for the HS and LS $\alpha 4\beta 2$ receptors and determining the mechanisms of modulation of the positive allosteric modulator dFBr. These goals were accomplished in two aims.

1.2.1 Aim 1: Establishment of Gating Models for the HS and LS $\alpha 4\beta 2$ Receptors

Before a modulators mechanism can be determined on a single channel level a baseline needs to be established. This is accomplished by determining gating models in agonist alone for each of the receptor subtypes. In Chapter 2, a 7 state gating model is established for the low sensitivity $\alpha 4\beta 2$ nAChRs under low concentrations of agonist ($\leq 1 \mu M$ ACh). The 7 state model contained two ligand binding reactions $C_1 \rightarrow C_2$ and $C_2 \rightarrow C_3$, two open states; a moniliganded state off of $C_2$ (O1) and a dililiganded state off of $C_3$ (O2), as well as two desensitized states, one off of $C_3$ (D1) and O2 (D2). In Chapter 3, the high sensitivity $\alpha 4\beta 2$ nAChRs required an 8 state model using data with concentrations ranging from 0.1 $\mu M$ to 10 $\mu M$ ACh for construction. The 8 state model contained two ligand binding reactions $C_1 \rightarrow C_2$ and $C_2 \rightarrow C_3$, mono and dililiganded interconnected open states (O1 off C2 and O2 off of C3), and three desensitized states with two off of O2 (D1 and D2) and one off of C3 (D3).
1.2.2 Aim 2: Modulation of α4β2 nAChRs by dFBr in Low Concentrations of Agonist

Once baseline models have been established in the absence of modulator then the mechanism of modulation by dFBr for the LS and HS receptors can be elucidated. Analysis of the mechanisms for the LS and HS receptors revealed two distinct mechanisms of modulation for dFBr on these receptors. In the latter half of Chapter 2 the mechanism of modulation of dFBr on LS α4β2 receptors are discussed in detail, with dFBr exhibiting potentiation by destabilizing desensitized states. Chapter 2 also contains a discussion on slow solution exchange in oocytes and its contribution in peak responses. Some of the discrepancy between our data and oocyte data is due to the slow response profile in oocytes. In Chapter 3 the mechanism of modulation of dFBr on HS receptors is determined, where it is found that dFBr’s effects are dependent upon the concentration of agonist. At low concentrations of agonist dFBr increases the mono-liganded open state lifetime, while at high concentrations an increase in the probability of entering the diliganded open state from the diliganded closed state is seen. Chapter 3 also contains a short discussion on the presence of HEPES and the effects the buffer may have on the HS data.

1.3 Hypothesis

1.3.1 dFBr Can Modulate α4β2 Receptors in Low Concentrations of Agonist

This hypothesis is based on the previous studies done by Weltzin et al and utilizes single-channel techniques (Weltzin and Schulte, 2010). This hypothesis addresses whether dFBr potentiates at low concentrations of agonist for the high and
low sensitivity α4β2 nAChRs. We express human α4β2 receptors in a mammalian cell line and use patch-clamp methods to record acetylcholine-evoked responses of HS or LS receptors in isolated patches of membrane. Unlike other methods, the single-channel methodology allows us to limit receptor heterogeneity as well as subject channels to a low concentrations of agonist. Utilizing this technique, we can gather long steady state recordings and develop a basic gating model for both stoichiometric forms of the receptor, which will be further used to determine dFBr’s mechanism of modulation.

1.3.2 dFBr Modulates by Destabilizing Desensitized States for α4β2 Receptors.

This hypothesis is an extension of the previous hypothesis and relies on the establishment of a basic gating model. The models generated will help determine the mechanism by which dFBr potentiates the high and low-sensitivity stoichiometric forms of the receptors. It is thought that the primary mode of potentiation exhibited by dFBr is through an increase in the open probability of the receptor. This can be achieved a few different ways, including: an increase in the rate of entry into the open state, leading to an increased probability of opening, or decreasing the rates out of the open states leading to an increase in mean open lifetime. However, an increase in open probability is not the only mechanism by which a potentiator can modulate a receptor. Our hypothesis is that dFBr’s primary mode of action is by the destabilization of one or more desensitized states, leading to potentiation. The development of a model and elucidating the mechanism of modulation will help further the characterization and therapeutic benefits of dFBr, as well as future α4β2 PAMs.
Figure 1.1: Interactions between VTA, NAcc and the Prefrontal Cortex. VTA: Ventral Tegmental Area, NAcc: Nucleus accumbens, PFC: prefrontal cortex, Glu: glutamatergic terminal, DA: Dopaminergic, GABA: GABAergic. Nicotinic receptors found on these terminals are labeled, with lines for the specific area the nicotinic receptor is located.
Figure 1.2: Classical Mechanisms. (A) Mechanism proposed by del Castillo and Katz (1957) to describe receptor-ligand interactions. (B) More recent model developed by Katz and Thesleff (1957) and modified to include two ligand binding sites as well as desensitized states. (C) A Monod-Wyman-Changeux (MWC) mechanism. A is agonist, R is receptor in the closed conformation, D is a receptor in the desensitized state, and O is receptor in the open conformation.
Figure 1.3: Complex Mechanisms. (A) A common example of a mechanism used to describe the nicotinic receptor with two ligand binding steps and mono and di-liganded openings. (B) Mechanism used to fit LS $\alpha 4\beta 2$ nAChR data. (C) Mechanism used to fit the HS $\alpha 4\beta 2$ nAChR data.
1.3 References


Chapter 2: Destabilization of Desensitized States of Low Sensitivity Nicotinic Acetylcholine Receptors

Abstract

A description of mechanisms that underlie modulation of nAChRs by allosteric ligands is critical for understanding the effects of modulation on cholinergic signaling in the CNS. We used single-channel methods to investigate the mechanism of potentiation of the human low-sensitivity, α4β2 nicotinic acetylcholine receptor by the allosteric ligand desformylflustrabromine (dFBr). Markov model analysis of steady-state, low concentration data from receptors expressed in HEK293 cells yielded a 7 state model with unique rate constants for naive and modulated receptors. The model consisted of two sequential binding reactions, mono- and diliganded open states, and two desensitized states, D1 and D2, accessible from the diliganded closed state and diliganded open state, respectively. We found that dFBr decreased the mean lifetimes of D1 (77 ms to 7.6 ms) and D2 (1.7 s to 490 ms). In simulations of macroscopic responses to concentration jumps, the EC50 of naive receptors was 74 μM for peak currents, with a maximal P_{open} of 0.9, and profound desensitization (steady-state P_{open} < 0.05 for all ACh concentrations) that developed rapidly (τ_{fast} = 70 ms) at high ACh concentration. dFBr increased the apparent affinity (EC50 = 18 μM) of receptors and potentiated steady-state, desensitized responses (~ 5-fold). To investigate the contribution of destabilization of desensitized states to modulation, we examined state occupancy probabilities for both models when driven by various profiles of ACh.
concentration. Potentiation of desensitized responses arose from a lower probability of occupancy of D2, whereas destabilization of D1 slowed the initial, fast rate of macroscopic desensitization. Recovery from desensitization was substantially faster with dFBr and correlated with decreased lifetimes of D1 and D2. Finally, we found that desensitization of LS receptors plays an important role in hastening the time course of deactivation following a brief pulse of agonist concentration\(^1\)

\(^1\) Demmerly, A., Edmonds, B.W. 2015. Destabilization of desensitized states underlies modulation of (α4)3(β2)2 receptors by desformylflustrabromine. J. Gen. Physiol.
2.1 Introduction

A major class of nicotinic receptors in brain is formed from $\alpha_4$ and $\beta_2$ subunits that assemble as pentamers with alternate stoichiometries, a high-sensitivity (HS; $\alpha_4\beta_2\beta_3$) receptor with an EC$_{50}$ of $\sim$ 1 $\mu$M and a low-sensitivity (LS; $\alpha_4\beta_3\beta_2$) receptor with an EC$_{50}$ of $\sim$ 75 $\mu$M (Nelson et al., 2003; Moroni et al., 2006). $\alpha_4\beta_2$ receptors are targets for compounds that bind allosteric sites and modulate responses to ACh. Compounds that potentiate and show selectivity for $\alpha_4\beta_2$ receptors may mitigate signaling deficiencies in Alzheimer’s disease (Warpman and Nordberg, 1995; Martin-Ruiz et al., 1999) and other cognitive disorders associated with insufficient receptor activation or expression. As an alternative to allosteric ligands, partial agonists (Sullivan et al., 1997; Coe et al., 2005) or inhibitors of acetylcholinesterase (Friedman, 2004) have been used as strategies to activate nAChRs; however, those options offer little opportunity for selectivity (Pandya and Yakel, 2011). Moreover, tonic occupation of orthosteric sites produces desensitization, resulting in a profound reduction in responsiveness to ACh (Paradiso and Steinbach, 2003). If cholinergic inputs are at least partially intact, allosteric ligands may yield better functional outcomes.

Novel, $\alpha_4\beta_2$-selective allosteric ligands are of particular interest as leads for development of additional compounds with distinct functional properties. NS9283 is an allosteric modulator with selectivity for the LS $\alpha_4\beta_2$ receptor subtype (Timmerman et al, 2012). In whole-cell recordings, NS9283 produces a marked increase in the potency of ACh with no apparent change in efficacy.
In response to brief, 1 ms applications of ACh, NS9283 prolongs the time course of deactivation ($\tau$ increased from 18 ms to 65 ms); however, the time course of desensitization is unaffected, and desensitized receptors are not reactivated by application of the modulator (Grupe et al., 2013).

Desformylflustrabromine is a metabolite originally isolated from *Flustra foliacea* that potentiates a mixture (HS and LS) of $\alpha4\beta2$ receptors, but not $\alpha3\beta4$ or homomeric $\alpha7$ receptors (Sala et al., 2005). The reported mechanism of potentiation is different from NS9283. dFBr increases the potency and efficacy of ACh (Kim et al., 2007), yielding up to 3-fold potentiation of peak current at high agonist concentration. The observation that dFBr activates desensitized receptors in whole-cell recordings led to the hypothesis that potentiation arises from stabilization of open states relative to desensitized states (Weltzin and Schulte, 2010); however, this would require that for steps into high concentrations of agonist, a significant fraction of the receptors is desensitized at the time of peak current. In a study of nonstationary noise of $\alpha4\beta2$ receptors expressed in HEK293 cells, the maximal open probability of LS receptors was reported to be large (> 0.8) (Li and Steinbach, 2010), which is at least inconsistent with marked potentiation of LS receptors for any mechanism other than recruitment of silent receptors (Fenster et al., 1999; Zhang and Steinbach, 2003; Li and Steinbach, 2010).
In this study, we used a single-channel approach to gain a better understanding of LS receptor gating and the mechanism of potentiation by dFBBr. Gating models developed from low concentration data exhibited characteristics typical of whole-cell responses previously reported for LS receptors (Nelson et al., 2003; Li and Steinbach, 2010; Grupe et al., 2013). We found evidence for two desensitized states, both of which were destabilized by dFBBr. Simulations of macroscopic responses showed that dFBBr produced marked potentiation of the steady-state response and accelerated recovery from desensitization. At saturating agonist, the peak open probability of naïve receptors was 0.9, and peak response was not potentiated.

2.2 Materials and Methods

2.2.1 Cell Techniques

HEK293 cells were maintained in Eagle's Minimum Essential Medium (MEM; Sigma Aldrich) supplemented with 1% GlutaMAX (Gibco), 10% bovine growth serum (HyClone, Thermo Scientific), penicillin (200 U/mL; Cellgro), and streptomycin (200 μg ml⁻¹; Cellgro), pH adjusted to 7.4. Cells were incubated at 37°C in a humidified atmosphere of 5% CO₂, passaged weekly and plated onto 35mm tissue culture dishes (Becton Dickinson) two days prior to transfection. Human α4 and β2 cDNAs [provided in pcDNA3.1/Hygro vectors (Invitrogen) by Dr. Marvin Schulte, University of the Sciences in Philadelphia] were purified using a Plasmid Midi Kit (Qiagen) and transiently transfected (SuperFect
Transfection Reagent, Qiagen) into HEK293 cells. Recordings were made approximately 2 days following transfection.

2.2.2 Electrophysiology Techniques

Desformylflustrabromine hydrochloride was obtained from Tocris. All other reagents, including acetylcholine chloride, were obtained from Sigma Aldrich. To minimize receptor rundown, patch-clamp recordings were made in the cell-attached configuration (Hamill et al., 1981). The membrane potential was nominally zeroed with a high K⁺ extracellular solution containing (in mM): 142 KCl, 5.4 NaCl, 2 CaCl₂, 1.7 MgCl₂, 5 HEPES, and 10 glucose, pH adjusted to 7.4. Pipette solutions were made by addition of either 1 μM ACh or, alternatively 1 μM ACh and 1 μM dFBr to (in mM) 5.4 KCl, 142 NaCl, 2 CaCl₂, 1.7 MgCl₂, and 5 HEPES, pH adjusted to 7.4, on the day of the experiment. All recordings were made at room temperature (~ 22°C).

Recording pipettes (thick-walled borosilicate glass; Warner Instruments) were pulled to a resistance of 10-14 MΩ using a PMP-102 micropipette puller (MicroData Instrument, Inc.). To reduce pipette capacitance, tips were coated with Sylgard (Dow Corning). Prior to recording, the pipettes were front- and then back-filled with the pipette solution containing ACh with or without dFBr. Cells were viewed with a Nikon FN1 Eclipse compound microscope, and patches were made (minimum seal resistance 5 GΩ) and then voltage-clamped at an applied potential +80 mV with a HEKA EPC-10 Double amplifier using PATCHMASTER.
software. Data were low-pass filtered at 5 kHz (4-pole Bessel, -3dB) and
digitized to the computer at 20 kHz.

Amplitude distributions (Figure 3.1) were fitted with the sum of 2 Gaussian
components. Dose-response data for peak responses in Figure 3.4 were fitted
with the Hill equation:

\[ y = \frac{y_{\text{max}}}{1 + (EC_{50}/A)^n}, \quad (1) \]

where \( A \) is agonist concentration, \( EC_{50} \) is the concentration of agonist at which
\( y = y_{\text{max}}/2 \), and \( n \) is the Hill coefficient. Values for simulated steady-state
responses were obtained at 15 s following a simulated jump into ACh.

2.2.3 Model Fitting

Idealized records for patches that appeared to contain 1 channel were
constructed using the segmental \( k \)-means method (Qin, 2004), and fitted rate
constants were obtained for models using a maximum likelihood algorithm (MIL)
in QuB (Qin et al., 1996, 1997). Imposed dead times of 200 \( \mu \)s provided the
most reasonable lifetimes for state C3 as well as the fit of the probability density
function (thick line in distributions). The number of components in closed and
open duration distributions indicated that at least 5 closed states and two open
states were needed to fit a full model that included desensitized states, which, prior to fitting, could not be assigned to particular gaps in the data record because of the low agonist concentration. The scheme shown in Model 1 was used to fit rate constants for 0.1 \( \mu \text{M} \), 1 \( \mu \text{M} \) ACh and 1 \( \mu \text{M} \) ACh + 1 \( \mu \text{M} \) dFBr data, where A is the ACh concentration, \( k_{+1} \) and \( k_{+2} \) are association rate constants for binding of 1\textsuperscript{st} and 2\textsuperscript{nd} ligand, and \( k_{-1} \) and \( k_{-2} \) are dissociation rate constants. Likewise, \( k_{O+2} \) governs the concentration dependent rate between O1 and O2, and \( k_{O-2} \) is the dissociation rate constant. \( \beta_1 \) and \( \beta_2 \) are opening rate constants for mono- and diliganded receptors, and \( \alpha_1 \) and \( \alpha_2 \) are corresponding closing rate constants. Rate constants \( \delta_{1+/-} \) and \( \delta_{2+/-} \) govern transitions into and out of states D1 and D2, respectively.

2.2.4 Definition of Bursts

Single-channel recordings often contain several channel openings in succession (bursts) followed by long periods of closings that occur spontaneously. Analysis of bursts allows us to obtain information on how burst characteristics can be changed in the presence of a modulator as well as how accurate the model is. We use bursts to determine the accuracy of our model by comparing mean burst open time and mean burst length values obtained from the physical data to those computed by the model. If the values obtained from the model resemble those obtained from our data, then we have more
confidence that the model is an accurate representation of the data. To do this we define bursts in the data with a critical value \((t_{crit})\) using the equation:

\[
Amp_1 e^{-T_{crit}/\tau_1} = Amp_2 (1 - e^{-T_{crit}/\tau_2})
\]

\(T_{crit}\) is the cutoff duration between exponential components and equalizes the area under the overlapping tails of the distributions. Where \(Amp_1\) and \(Amp_2\) are the weights of the first and second components respectively for bursts. \(\tau_1\) and \(\tau_2\) are the time constants computed from MIL. Each \(\tau\) in the distribution contributes one exponential component to the probability density function (thick line in distributions). Bursts that held doubles were discarded from the analysis, and values were compared to burst durations calculated from the model.

### 2.2.5 Calculation of Burst Parameters

We define bursts as a series of openings that occurs before ligand unbinds from the receptor. Burst parameters were calculated from transition probabilities (Colquhoun and Hawkes, 1982). Since ligand remains bound, a burst was defined as 1 or more openings consisting of transitions between \(O_2\) and \(C_3\) only. The burst begins with an opening to \(O_2\) and ends with a transition to \(D_2\) or, alternatively, to \(C_3\) from which the receptor does not reopen (next transition to \(D_1\) or \(C_2\)). \(O_1\) openings were not included due to the low occupancy
of O1. To determine the distribution of the number of openings per burst, we calculated the probability that a burst contains 1, 2, 3, ..., r openings. For a burst to consist of a single opening, the open channel must make a transition to C3 followed directly by a transition to C2 or D1 (state 7 → state 3 → state 2 or state 7 to state 3 → state 4; see state number assignments, Model 1). Alternatively, the open channel can desensitize to D2 (state 7 → to state 5) without entering C3. The probability that the burst consists of a single opening \( P(1) \) is therefore given by the sum of the probabilities of these events. \( P(1) = \pi_{73} \cdot \pi_{32} + \pi_{73} \cdot \pi_{34} + \pi_{75} \), where \( \pi_{73} = \alpha_2 / (\alpha_2 + \delta_2^+), \ \pi_{32} = k_2 / (k_2 + \beta_2 + \delta_1^+), \ \pi_{34} = \delta_1^+ / (\delta_1^+ + k_2 + \beta_2), \ \text{and} \ \pi_{75} = \delta_2^+ / (\delta_2^+ + \alpha_2). \)

For the probability that the burst consists of two openings, \( P(2) = \pi_{73} \cdot \pi_{37} \cdot \pi_{32} + \pi_{73} \cdot \pi_{37} \cdot \pi_{34} + \pi_{73} \cdot \pi_{37} \cdot \pi_{75} \). The probability of \( r \) openings is given by:

\[
P(r) = \pi_{73}^r \cdot \pi_{37}^{r-1} \cdot (\pi_{32} + \pi_{34}) + \pi_{73}^r \cdot \pi_{37}^{r-1} \cdot \pi_{75}
\]

(6)

Using the identities \( \pi_{32} + \pi_{34} = 1 - \pi_{37} \) and \( \pi_{75} = 1 - \pi_{73} \), the expression can be simplified to the form for a geometric distribution:

\[
P(r) = (\pi_{73} \cdot \pi_{37})^{r-1} \cdot (1 - \pi_{73} \cdot \pi_{37})
\]

(7)
The mean number of openings per burst, $m_r = \sum_{r=1}^{\infty} r \cdot P(r)$, was calculated from the mean of the geometric distribution:

$$m_r = \frac{1}{1 - \pi_2 \pi_3} \quad (8)$$

The sum of the mean open time per burst $[m_r \cdot m_{o2}]$, where $m_{o2} = 1/(\alpha_2 + \delta_{2+})$ and the mean closed time per burst $[(m_r - 1) \cdot m_{c3}]$, where $m_{c3} = 1/(\beta_2 + k_{-2} + \delta_{1+})$, gives the mean burst duration. The probability of being open during the burst was then calculated by dividing the mean open time per burst by the mean burst duration.

Model 1.

2.2.6 Estimation of Channel Open Probability

Estimates of the probability that each patch contained only a single channel were made using an approximation given by Colquhoun & Hawkes
For N independent channels with two states (open and closed), the probability that, if one channel is open, the next transition will be closing of the open channel rather than opening of the second channel is given by:

\[ P_{\text{open}} = \frac{m_0}{m_0 + m_c} \] (m_0 and m_c are the mean open time and mean closed time, respectively) is the probability that a channel is open on an observed record of single openings, and N is the number of channels in the patch. The probability of observing r consecutive single openings before the first double opening is \( P(r) = \pi^{r-1}(1 - \pi) \). The probability of observing at least \( n_0 \) consecutive single openings is then given by the cumulative form of the geometric distribution.

\[ P(r \geq n_0) = \pi^{n_0-1}. \] 

When channels exhibit long (desensitized) closed periods punctuated by bursts with high \( P_{\text{open}} \), a better (more stringent) estimate of the probability of two channels is obtained by treating bursts as single openings (Colquhoun and Hawkes, 1990). We therefore adjusted \( n_0 \) to reflect the number of bursts. 

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\( m_{r_{\text{exp}}} \) by dividing the number of bursts by the openings per burst, where \( m_{r_{\text{exp}}} \) is the experimentally determined number of openings per burst. \( m_{r_{\text{exp}}} \) (no correction for missed events) was obtained in QuB by defining a \( t_{\text{crit}} \) separating components 1 and 2, or components 2 and 3 of closed duration distributions for experiments in ACh and ACh + dFBr, respectively. We defined the \( t_{\text{crit}} \) for dFBr as between 2 and 3 due to the components of 1 and 2 being extremely close to each other. Bursts were defined as runs of openings containing gaps less than \( t_{\text{crit}} \). For the probabilities reported here as \( P(r \geq n_0) \), \( r \) and \( n_0 \) are bursts, not openings.

2.2.7 Simulations

All simulations shown for control (ACh model) and modulated (dFBr model) responses were made using rate constants from the best model of the set as determined by the total number of openings (> 1700, includes openings and closings) (see Table 1). A set was defined as patches that were obtained with the same ACh/dFBr concentration profile. Monte Carlo simulations of macroscopic currents in Figure 2.4A were made in QuB Express for 10,000 channels with a unitary amplitude of 1 pA and standard deviation (SD) of Gaussian noise of 0.15 pA (Nicolai and Sachs, 2013). Current values were converted to and displayed as open probabilities \( (P_{\text{open}}) \) in Igor Pro (WaveMetrics).
For a Markov process, the evolution of state occupancy probabilities is given by:

\[ \frac{ds_i}{dt} = \sum_{j=1}^{n} s_j r_{ji} - \sum_{j=1}^{n} s_i r_{ij} \]  

(4)

where \( s_i \) is the fraction of channels in state \( S_i \), and \( r_{ji} \) and \( r_{ij} \) are the rate constants governing transitions into and out of state \( S_i \), respectively (Colquhoun and Hawkes, 1981). Given an initial condition of \( P_{c1} = 1 \) and a concentration profile \( c(t) \), which sets the values of concentration dependent rates, \( s(t) \) was integrated in xpp (v. 7.0, Bard Ermentrout; www.math.pitt.edu/~bard/xpp/xpp.html) using a Runge-Kutta method and 10 μs time steps.

To examine trajectories of occupancy probabilities, simulations were driven by three different agonist concentration profiles: 1) a jump, 2) a slowly rising exponential function, and 3) a brief pulse of agonist given by an alpha function to mimic transmitter concentration at a fast synapse (\( \tau = 500 \) μs).

\[ c(t) = A \times (t/\tau) \times e^{-(t-\tau)/\tau} \]  

(5)
2.3 Results

2.3.1 Conductance of LS Receptors

Channel openings with two distinct amplitudes were observed in cell-attached recordings with a low concentration of ACh (1 μM). Figure 2.1A shows 1 s records and corresponding open-point amplitude histograms for typical low- and high-amplitude channels recorded at -80 mV. Assuming a reversal potential of 0 mV, fitted histograms yielded chord conductance \( g \) estimates of 19.9±1.4 pS (mean±SD; \( n=13 \) cells) and 28.6±2.5 pS (\( n=19 \) cells). Consistent with previous reports of inward rectification of \( \alpha 4\beta 2 \) receptors (Buisson et al., 1996; Sabey et al., 1999), we did not observe LS receptor openings at assumed patch membrane potentials above 0 mV; however, I-V data collected between -80 and -20 mV for two patches (one in dFBr; Figure 2.1C, open circles) were well fitted by a linear equation, yielding a slope conductance estimate (30 pS) similar to our estimate of \( g \), and a predicted reversal potential near 0 mV (-2.7 mV). Assuming inward currents rectify for depolarizations above -20 mV, our estimate of \( g \) may be somewhat high due to an underestimate of the driving force.

In patches that appeared to contain only one channel conductance, based on the observation of openings to only one level over a minimum recording period of 2.5 min, 71% (15 out of 21 patches) contained LS receptors. Differences in apparent affinity for ACh for the two subtypes of \( \alpha 4\beta 2 \) receptors (LS, \( \alpha 4\beta 2 \) EC\(_{50} \sim 75 \mu M \) vs HS, \( \alpha 4\beta 2 \) EC\(_{50} \sim 1 \mu M \) (Buisson and Bertrand, 2001; Nelson et al., 2003; Moroni et al., 2006)) favors activation of HS receptors;
therefore, our data is consistent with reports that at least 70% of functional α4β2 receptors expressed in HEK 293 cells are LS receptors (Li and Steinbach, 2010; Grupe et al., 2013). Modulation of LS receptor currents by dFBr could arise from effects on channel gating, unitary conductance, or recruitment of silent receptors (Fenster et al., 1999; Zhang and Steinbach, 2003; Li and Steinbach, 2010) (not investigated here). Fitting of open point amplitude histograms for ACh + dFBr (Figure 2.1B) yielded a mean conductance of 28.4±2.0 pS (mean±SD, n=12 cells), similar to the value for ACh. Accordingly, dFBr does not modulate receptor function by altering single channel conductance. Sublevels were observed in patches with and without dFBr (Figure 2.1B, inset). Sublevels appeared more frequently dFBr; however, because they were brief and, even in dFBr, occurred too infrequently to analyze, they were fitted as full amplitude events.

2.3.2 Gating of LS Receptors in ACh and ACh + dFBr

Recordings in 1 μM ACh or, alternatively, 1 μM ACh + 1 μM dFBr yielded 3 patches for each condition (out of a total of ~ 150 patches containing LS receptors) that contained runs of 30 pS openings of sufficient length (> 250 openings, average of 560) and no evidence of simultaneous openings of two channels. When the probability of a channel being open is low, as it was in our experiments (P_{open} ~ 0.01), more than one channel may be present despite the observation of openings to only one level (Colquhoun and Hawkes, 1990). We
estimated the probability of obtaining the observed run of single openings (corrected for high $P_{\text{open}}$ bursts, see Methods) for each patch if, in fact, two channels were present (see Methods). Values of $P(r \geq n_o)$ were 0.01, 0.23, and 0.53 for ACh, and 0.03, 0.45, and 0.11 for ACh + dFBr. Figure 2.2 shows 50 s segments from ACh and dFBr patches with the lowest probability of having two channels.

In single-channel recordings it is often observed that several channel openings occur in rapid succession, where the individual openings are separated by brief closings (Colquhoun and Hawkes, 1982). Analyzing bursts allows all the openings within one burst to be attributed to a single channel, allowing shut periods within bursts to tell us about characteristics of a channel or the effects a modulator has on burst structure. Figure 2.3 shows burst length and open lifetime for 1 $\mu$M ACh (A) and 1 $\mu$M ACh + 1 $\mu$M dFBr (B). Two exponentials were needed to fit the distributions of burst length and open time per burst. Burst length and open lifetime decrease in the presence of dFBr; the fast ($\tau_f$) and slow ($\tau_s$) time components decrease for burst length from 2.86 ms to 1.6 ms for $\tau_f$ and 42.1 ms to 14.7 ms for $\tau_s$. Burst open lifetime distributions showed a similar trend, with $\tau_f$ being reduced from 3 to 0.75 ms and $\tau_s$ being reduced from ~16 to ~11 ms. dFBr decreases the mean burst length and burst open lifetime, which is counter-intuitive to how a potentiator is expected to behave. The decrease in burst length is due to dFBr effectively reducing the intra-burst duration via decreasing the mean lifetime of desensitized state D1 (See 2.3.7), allowing greater simultaneous activation of receptors when exposed to agonist. Burst
parameters for the patches, calculated for zero agonist concentration, were in agreement with the time constants obtained from the fitted burst distributions. The calculated mean burst duration was 51.4 ms (42.1 ms from fits) and 25.7 ms (~15 ms from fits) for ACh and dFBr respectively. The calculated mean open time was also in agreement with values of 12.4 ms (~16 ms for the fits) and 8.1 ms (~11 ms for the fits) in ACh and dFBr respectively. These results give us more confidence that our model is a fairly accurate representation of LS α4β2 nAChR gating.

The stability of the patches for all concentrations (0.1 and 1 μM) and in the presence of modulator (1 μM ACh + 1 μM dFBr) is shown in Figure 2.4. P_{\text{open}} was larger when calculated from the stability plots (0.1 μM ACh: 0.13, 1 μM ACh: 0.16, and + dFBr: 0.14) due to the inclusion of periods of high occupancy.

2.3.3 Fitting Mechanisms to Single Channel Data

We investigated the mechanism of modulation for dFBr on α4β2 nACh receptors by fitting a wide range of models directly to the single-channel data. Figure 2.5 shows models tested (left) along with the final model for 1 μM ACh and 1 μM ACh +1 μM dFBr. Rates plus standard deviations (0.1 μM and 1 μM ± dFBr). Models on the left include Katz and Thesleff, Monod Wyman Changeux (MWC), modified MWC models that include desensitization and two models that did not contain activations steps, linear and star. All models were constrained by the requirement of microscopic reversibility for the cycle, while activation rates
were not constrained. Due to the low probability of unliganded openings under our conditions these states were omitted from MWC and similar models.

More classical mechanisms such as the Katz and Thesleff and MWC models were initially utilized in order gain a better understanding of the data. The Katz and Thesleff model fit the closed time distributions well for all concentrations and one open state fit the 1 μM ACh data but not the 0.1 μM data (Figure A.2.1). Activation rates are 10-fold higher for 1 μM data than the 0.1 μM data, whereas the concentration dependent rates of $2k_{+2}$ and $k_{+1}$ are 100 to 1000x larger for the 0.1 μM data than the 1 μM data (Table A.2.1). Lifetime of C3 is also extremely low and well below our dead time, giving less confidence in the rates leaving C3. The basic MWC model had similar problems in regards to the activation rates being 10 to 100 x larger for 0.1 μM vs 1 μM ACh data. Activation rates for the 1 μM data set were low and gave low $P_{\text{open}}$ in simulated macroscopic currents (not shown). The MWC model clearly showed that more than 3 closed states are required to fit the closed time distributions for our data. Recent work using concatenated receptors provides evidence that $\alpha_4\beta_2$ receptors possess two high affinity binding sites at $\alpha$-$\beta$ interfaces in addition to a 3rd, high affinity site at the $\alpha$-$\alpha$ interface (Harpsoe et al., 2011). Attempts to fit permutations of the MWC mechanism that included an additional agonist binding reaction always produced unreasonably large or small values for at least 1 rate constant. It may be that the additional activation step and subsequent open state have a lifetime very similar to that of either the mono or diliganded activation
steps and open states in our data, thus we are unable to distinguish them from the other states.

2.3.4 Modified MWC Mechanisms

From the classical mechanisms we tested various scenarios of desensitization and its connections to the models. In order to improve fits, we added desensitized states to the MWC mechanism. The MWC modified mechanisms gave the most reliable results for all the cases we tested so we will focus on those mechanisms here. We tested various locations of two or more desensitized states. These included single desensitized states off of either O1, O2, or C3 (not discussed), two desensitized states off of O1/O2, O1/C3, or O2/C3, and a few variations with three desensitized states. Only a few of these variations are discussed in detail here. For these modified models two desensitized states or more gave better results in terms of fits to the closed distributions and reasonable simulated currents. For mechanisms that contained three desensitized states we excluded testing configurations that contained a D state off of O1. This is because any configuration that included a state off of O1 resulted in a difference in the activation rates between concentrations, as well as dFBr not potentiating current when simulations were performed (data not shown), which does not agree with previous findings on dFBr’s apparent activity in whole cell experiments (Weltzin and Schulte, 2010). Desensitized states off of O1 and O2 gave decent fits (Figure A.2.3), but still had the issue of activation rates being
10 to 100x higher for the 0.1μM data and the rates between O1 and O2 are high (Table A.2.3). It was found that a desensitized state was required off of state O2 and off of C3 in order for the activation rates to agree with each other.

We further tested the existence of an additional desensitized state off of O2, leading to two desensitized states off of O2 and one off of C3. The fits to closed and open time distributions improved marginally with this addition and it was difficult to determine if an additional desensitized state was required from the fit alone (Figure A.2.4). With this configuration we saw a large discrepancy between the activation rates and rates entering O1 when comparing multiple concentrations of ligand (Table A.2.4). Simulated macroscopic currents also showed an increase in desensitization compared to the two desensitized state models.

The mechanism that gave us consistent activation rates as well as reasonable simulations for EC_{50} was a two desensitization state configuration with states off of O2 and C3. Consistent with reports in which 2 desensitized states were identified for LS α4β2 receptors (Sabey et al., 1999; Nelson et al., 2003; Paradiso and Steinbach, 2003; Grupe et al., 2013). Rate constants from our two best patches (shown in Figure 2.2) that contained the largest number of events (ACh: 1754; ACh + dFBr: 1785), are shown in Table 2.1 were used for simulations. The average data from analysis of all 3 data sets in each condition are shown in Table 2.2.
For the maximum likelihood fitting algorithm, model rate constants are adjusted to maximize the likelihood of the data (sequence and durations of openings and closings) with correction for missed events (Qin et al., 1996). Estimated rate constants for our model (Figure 2.5) yielded distributions of open and closed durations that agree with the observed data (Figure 2.6). A comparison of the data sets collected under each parameter (ACh and ACh + dFBr) indicated that rate constants for transitions to O1 ($\beta_1$), from O1 to O2 ($k_{+2}$) and transitions out of D1 and D2 ($\delta_1$- and $\delta_2$-) were significantly larger in dFBr (P values: 0.061, 0.008, 0.001 and 0.047 for two-tailed t-test, respectively; Table 2.2). For the fitted rate constants in Figure 2.5, the increase in $\delta_1$ corresponds to a decrease in the mean lifetime of D1 ($1/\delta_1$) from 77 ms to 7.6 ms, and the increase in $\delta_2$ yields a reduction in the mean lifetime of D2 from 1.7 s to 493 ms.

2.3.5 Desensitization Based Mechanisms

At high concentrations of ACh almost all current can be attributed to the activity of diliganded AChRs (Elenes and Auerbach, 2002). The lifetimes of diliganded closed mouse AChRs (~10 $\mu$s) are too brief to be detected (Maconochie and Steinbach, 1998). If concentrations are high enough, then the lifetimes of diliganded closed states are too brief to be detected. To account for this possibility are we fitted models that contained a doubly liganded open state and only desensitization like states for closed states. We made two desensitization based models, linear and star. For the linear and the star models the fits are relatively good for closed time distributions, but open time
distributions for 0.1 μM and 1 μM ACh +1 μM dFBr requires more than one open state (Figure A.2.5 and A.2.6). Majority of rates agreed in the models between concentrations. The exceptions were for linear $k_{23}$ and $k_{21}$, and for star $k_{04}$, $k_{30}$, and $k_{10}$ (Table A.2.5 and A.2.6). Simulations testing the decay time course for these models (starting in state O) revealed dFBr increased the decay time course, which does not coincide with previous reports on the effects of dFBr (Kim et al., 2007; Weltzin and Schulte, 2015). It is not surprising that these schemes failed to describe open-time distributions that clearly contain more than one open component. Also, if our data did not contain any information on activation, then we would see a strong case for dFBr as a potentiator that destabilizes desensitized states in models that lack activation rates. We are confident that the concentrations utilized in the low-sensitivity experiments are low enough to allow the observation of activation kinetics for the α4β2 LS subtype.

2.3.6 Simulations of Macroscopic Currents

To determine if our estimated kinetic parameters supported potentiation over a range of ACh concentrations, we simulated macroscopic responses to concentration jumps (Figure 2.7A). Potentiation of the peak response by dFBr was apparent at 10 μM ACh, less at 100 μM, and negligible at 1 mM ACh where the peak open probability for ACh alone was ~ 0.8. At concentrations of ACh above 1 mM, peak responses were inhibited in dFBr (Figure 2.7B). dFBr yielded 3-3.5-fold potentiation of desensitized (near steady-state) responses, as
expected for shorter lifetimes of desensitized states. Fitted Hill equations to peak responses yielded (mean±SD) EC$_{50}$ values and Hill slopes of 58±4 μM and 0.76±0.04 for ACh, and 17±1 μM and 0.98±0.05 for dFBr. The concentration dependence of dFBr-induced differences (increases) in responses are shown for ACh concentrations up to 1 mM in Figure 2.7C. Our model yielded a maximal P$_{\text{open}}$ of ~ 0.9, consistent with an estimate from noise analysis for LS receptors of > 0.8 (Li and Steinbach, 2010).

When driven with a jump into 1 mM ACh, desensitization exhibited a biphasic, exponential decay with a time constant of 55 ms for the fast, major amplitude component (relative amplitude 0.33) and 220 ms for the slow component (Figure 2.7A). The fitted, steady-state value was 0.07, with steady-state values of P$_{\text{open}}$ for all concentrations (assessed from mean response amplitudes near the end of a 15 s jump) are shown in Figure 2.7B (open squares).

2.3.7 Occupancy Probabilities and Desensitization

An advantage of single-channel methods for studies of mechanism is that once the rate constants for the model are determined, the time-dependent probability of occupancy of each state can be determined for any concentration profile. One aim of examining occupancy probabilities was to determine how changes in occupancy of D1 and D2 contribute to potentiation. Figure 2.8 shows occupancy probabilities for concentration jumps into 10 μM, 100 μM, and 1 mM
ACh. The open probability was approximated by $P_{O2}$ ($P_{O1}$ is less than 0.005 for all simulations). For jumps into a low concentration of ACh (10 μM), dFBr increased the peak response by a factor of 2 ($P_{O2,ACh} = 0.13$, $P_{O2,dFBr} = 0.26$).

During the early phase of macroscopic desensitization (right panels Figure 2.8), the reduction in $P_{D1}$ by dFBr was the significant contributor to potentiation. Even though entry into D1 occurs earlier, $P_{D1,dFBr}$ is lower and drops faster compare to that of $P_{D1,ACh}$. Potentiation is almost entirely due to destabilization of D1. This is most notable at larger concentrations and appears to be negligible at concentrations of ~1 μM. As $P_{O2}$ approaches steady-state (best seen for 1 mM simulation, lower panels), the difference between $P_{D1,ACh}$ and $P_{D1,dFBr}$ declines due to the relatively slow relaxation of ACh model receptors from D1 into D2. Steady-state differences in $P_{O2}$ arise almost entirely from destabilization of D2. The early versus late contribution of destabilization of D1 and D2 are also seen for higher concentration simulations, where the fraction of bound receptors are close to 1 shortly after $P_{O2}$ reaches the peak.

2.3.8 Slow Solution Exchange and Potentiation

In a recent study in which α4β2 receptors were in excess, 3-fold potentiation was observed for 1 μM dFBr with 100 μM ACh (EC75 for that preparation) (Weltzin and Schulte, 2015). Our models yielded a prediction of less than 1.1-fold potentiation at EC75 (~ 300 μM) because of the high $P_{open}$ (~ 0.68) at the peak response to ACh alone. Differences between oocyte results
and model predictions could arise from different gating properties of LS receptors in HEK cells, recruitment of receptors from a large population of nonfunctional, silent receptors (Fenster et al., 1999; Zhang and Steinbach, 2003; Li and Steinbach, 2010), or from a relatively slow activation profile in oocytes due to slow solution exchange (Joshi et al., 2004). To determine if slow solution exchange predicted larger potentiation, we calculated occupancy probabilities using an exponential function for the concentration profile \( c(t) = c_{\text{max}} \times (1 - \exp(-t/\tau_c)) \) (Figure 2.9). A time constant \( (\tau_c) \) of 4 s yielded a time to peak of \( \sim 400 \text{ ms} \) (ACh model) for 300 \( \mu \text{M} \) ACh, similar to published results for \( \alpha 4\beta 2 \) responses in oocytes (Kim et al., 2007; Weltzin and Schulte, 2010). A 400 ms rise time is also expected for the vertical flow exchange apparatus used in those experiments (Joshi et al., 2004). Figure 2.9A shows that a slow rise in ACh concentration yielded a lower peak \( P_{\text{open}} \) for the naive receptor and a large increase in response (1.64-fold) compared to jumps (lower panel).

2.3.9 Deactivation Following a Pulse of High Agonist Concentration

To examine macroscopic responses to an agonist concentration profile typical of fast synapses (Clements et al., 1992; Silver et al., 1996; Beato, 2008), we drove models with an alpha function with peak concentration of 1 mM and half maximum width of 1.2 ms (Figure 2.10). The peak open probability for the ACh model \( (P_{02, \text{ACh}} = 0.28) \) was low in comparison to that found for sustained jumps into the same concentration of ACh. dFBr yielded a 1.6-fold increase in
response amplitude ($P_{02_{dFBr}} = 0.45$) due to an increase in the fraction of bound receptors. For the ACh model, the decay of $P_{02}$ (deactivation) exhibited a fast, major amplitude (0.72) phase with a time constant of 44 ms, in addition to two smaller, slow components with time constants of 133 ms and 5.4 s. The slow time constants matched time constants describing relaxation of $P_{D1}$ and $P_{D2}$, consistent with the view that the time course of LS receptor deactivation reflects a fast phase of channel closing followed by delayed openings arising from receptors leaving D1 and D2.

In single channel recordings of α7 nAChRs, individual openings were found to terminate not by channel closing, but by entry into desensitized states (daCosta et al., 2011) (see also, (Marabelli et al., 2015)). For the LS receptor, relaxation of receptors into D1 and D2 exhibited a time course similar to the fast decay of $P_{02}$ (Figure 2.10A) indicating that bursts of openings may terminate in D1 or D2. To test the hypothesis that the initial phase of deactivation is accelerated by entry into desensitized states, responses were simulated with $\delta_{1+}$ and $\delta_{2+}$ set to zero. The inset shows that the mean decay time of $P_{02}$ was prolonged ~ 3-fold ($\tau=157$ ms). The deactivation time course of LS receptors was therefore not a simple reflection of channel closing.

For the dFBr model, the decay of $P_{02}$ was fitted with the sum of two exponential components. The time constant of the major amplitude ($A_1=0.97$), fast phase (45 ms) was slightly faster than the value found for naive receptors, and the time course of the slow phase ($\tau_2=649$ ms) was similar to the time course
of decay of $P_{D2}$ ($\tau=653$ ms). A decay component for $P_{O2}$ reflecting recovery of receptors from D1 was not observed because the fast time constant for relaxation of $P_{D1}$ ($\tau_1 = 76$ ms; $A_1 = 0.97$) was the same as the fast time constant for $P_{O2}$. $P_{D1}$ is scaled in Figure 2.11B to show this relationship. Despite the fact that deactivation was faster in dFBr, charge transfer at 100 ms was still 1.4-fold larger. In contrast to naïve receptors, removal of desensitization for the dFBr model only produced a small increase in the time constant ($\tau=50$ ms) of a single exponential fit to the decay of $P_{O2}$ (inset). With only $\delta1+$ set to zero (trace not shown), the time constant was slightly shorter ($\tau=39$ ms); therefore, elimination of D1 offsets the effect of removing D2. The calculated mean burst duration in dFBr was 26 ms ($P_{\text{open}}=0.98$, $m_r=3.1$) with desensitization, so a functional burst in dFBr includes, on average, ~1 sojourn in D1. When burst $P_{\text{open}}$ is high, the mean burst duration is a good approximation of the time constant fitted to the decay of aligned bursts (no latency to first opening). Removal of desensitization increased the mean burst duration from 51 ms ($P_{\text{open}}=0.98$, $m_r=4.1$) to 154 ms ($P_{\text{open}}=0.98$, $m_r=8.5$).

The development of and recovery from desensitization following a pulse of agonist concentration was faster in dFBr (Figure 2.11). dFBr decreased the time to peak of $P_{D1}+P_{D2}$ from 187 ms to 69 ms. Recovery from desensitization reflected recovery of both $P_{D1}$ and $P_{D2}$ for ACh and $P_{D2}$ only for dFBr. Despite the fact that receptors were only exposed to saturating agonist for ~1 ms, 54% of naïve receptors bound by agonist were desensitized. In dFBr, only 18% were desensitized.
2.4 Discussion

Studies of \( \alpha 4 \beta 2 \) receptors in whole cells are complicated by expression of two subtypes of receptors (Nelson et al., 2003; Moroni et al., 2006). With oocytes, the ratio of injected RNAs can be altered to favor one subtype (Nelson et al., 2003; Moroni et al., 2006; Weltzin and Schulte, 2015); however, for studies of fast receptors, cell lines are often preferable because solution exchange times are shorter for small cells. When \( \alpha 4 \beta 2 \) receptors are expressed in HEK293 cells, most of the functional surface receptors are the low-sensitivity (\( \alpha 4 \beta 2 \)) subtype (Nelson et al., 2003; Li and Steinbach, 2010), and this preparation has been used successfully to characterize macroscopic properties of LS receptors and, recently, to investigate mechanism for the LS-selective modulator NS9283 (Grupe et al., 2013). Unfortunately, dFBr potentiates both subtypes (Weltzin and Schulte, 2015) so, for us, a single-channel approach provided the most straightforward option for circumventing problems with heterogeneity, with the added benefit of gaining greater insight into mechanism. If full gating models can be obtained (or at least ones that give a fairly complete representation of function), as we did here for naïve and modulated receptors, simulations of macroscopic responses can be made for any profile of agonist concentration. This could prove useful for predicting the consequences of dFBr modulation for signaling at LS receptors because receptors may experience quite different profiles in ACh concentration due to different spatial relationships between receptors and release sites (Descarries et al., 1997; Ren et al., 2011) and, as we
suggest here, the change in the postsynaptic signal may depend on the concentration profile, with larger effects for profiles that favor desensitization.

The model for the naive receptor presented in this study generated predictions for macroscopic behavior that agree with results for human LS receptors from other studies. Our finding of two desensitized states with short and long lifetimes is supported by observations of two phases of current decay in concentration jump experiments (Nelson et al., 2003; Paradiso and Steinbach, 2003; Grupe et al., 2013), and also by experiments documenting corresponding fast and slow phases for recovery of receptors from desensitization (Paradiso and Steinbach, 2003; Grupe et al., 2013). Except perhaps for the slow the time course of desensitization recovery (discussed below), the values of time constants (of fitted exponentials) reported by other groups agree with our values (Nelson et al., 2003; Grupe et al., 2013). The fitted rate constants for our model also yielded expected values for potency and Hill coefficients (Nelson et al., 2003; Moroni et al., 2006; Grupe et al., 2013). Ideally, we would like to perform additional experiments with a range of ACh + dFBr concentrations to verify if the potentiation exhibited by dFBr on a single-channel level depends on agonist concentration. Overall, the model does a remarkably good job of replicating high concentration results.

The primary finding of this study was that dFBr decreased the mean lifetimes of two desensitized states. Our conclusions, and the validity of our models, depend on obtaining correct estimates for the lifetimes of C1, D1, and D2, the longest closed durations in our single-channel records and, therefore, the
most susceptible to underestimation due to the possibility that patches contained multiple channels. For the patches used to fit rate constants for simulations, we found low values (<0.05) for the estimates of the probability that a second channel is present \( P(r \geq n_0) \); however, the estimates for \( P(2) \) channels were larger for other patches. Moreover, if a second channel were active for only part of the record, \( P(r \geq n_0) \) is less reliable. D2 is isolated from other closed states and has a long mean lifetime (1.7 s); therefore, if undetected channels were present, the value of \( \delta_2 \) should be correlated with \( P(r \geq n_0) \). Regression analysis showed that \( P(r \geq n_0) \) and \( \delta_2 \) were not correlated (Pearson \( r = -0.29, P(r) = 0.57 \)). The finding of shorter lifetimes of D1 and D2 in dFBr appears to be well supported; however, we are less certain about the precise values of \( \delta_2 \) and \( k_{+1} \). For rat \( \alpha_4\beta_2 \) receptors activated by ACh, the reported slow time constant for recovery from desensitization was 17 s (Paradiso and Steinbach, 2003). In a recent study of human, mostly LS receptors expressed in HEK 293 cells, the slow phase of recovery was 12.8 s (Grupe et al., 2013). In simulations (not shown), decreasing the value of \( \delta_2 \) from 0.6 s\(^{-1}\) to 0.2 s\(^{-1}\) (a value well within the error reported for group data) prolonged the slow recovery time constant from 5 s to 10 s, and decreased the steady-state \( P_{\text{open}} \) from 0.05 to 0.025. Effects on other macroscopic parameters were negligible. Additional studies are needed to obtain better estimates of \( \delta_2 \) and \( k_{+1} \).

Our findings have potential implications for signaling. First, in cases where receptors are subject to a prolonged elevation in transmitter concentration (Descarries et al., 1997; Zoli, 1999; Ren et al., 2011), destabilization of
desensitized states would result in a substantial increase in charge transfer. Interestingly, dFBr had no effect on response amplitude for agonist concentrations below 1μM, so the possibility exists for functional selectivity based on a threshold of agonist concentration.

At fast synapses, for which low-sensitivity receptors are better suited (Nelson et al., 2003), receptors are exposed to a brief concentration and are far from steady state. Our results predict a steep dependence of receptor occupancy on transmitter concentration. For glutamatergic synapses, estimates of the time constants for the decay of transmitter range from 1.2 to 4 ms (Clements et al., 1992; Diamond and Jahr, 1995). For our model, increasing the duration of the pulse of 1 mM ACh from 1 ms to 4 ms increased the estimated $P_{\text{open}}$ from 0.25 to 0.68. From the perspective of development of positive allosteric ligands, if receptors are saturated at fast cholinergic synapses, dFBr and other compounds that exhibit potentiation by scaling the current, will be relatively ineffective at boosting postsynaptic signals that arise from low frequency inputs. Compounds that enhance charge transfer by prolonging deactivation may be more effective, e.g. NS9283 (Grupe et al., 2013). For high frequency stimuli, compounds that accelerate recovery from desensitization may be preferable.

Paradiso and Steinbach showed that prolonged application of ACh promotes entry into a desensitized state that recovers slowly; the relative amplitude of the slowly recovering component of desensitization increased with longer jumps into agonist (Paradiso and Steinbach, 2003). Our results do not
fully account for their observations, which suggest the existence of a third desensitized state; however, we do observe a redistribution of receptors from D1 to D2 at longer times ($\tau \sim 1$ s). For experiments in which desensitized receptors were found to be reactivated by dFBr (Weltzin and Schulte, 2010), the only mechanism that, for our model, can account for their observed rapid increase in the fraction of open receptors is via an increase in the rate out of D2. Our results do not support the large increases in peak current with high agonist concentrations found in oocyte experiments (Kim et al., 2007; Weltzin and Schulte, 2015); however, at least part of the discrepancy is likely due to slow solution exchange, which allows time for desensitization to develop.

In conclusion, we provide the first model for gating of human $\alpha_{4}\beta_{2}$ receptors from single channel data. We provide evidence favoring the view that the allosteric modulator dFBr improves LS receptor responsiveness by destabilization of desensitized states. The consequences of modulation for signaling will likely depend on the frequency and profile of agonist concentration at the receptors.
2.5 Acknowledgments

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Figure 2.1A: Unitary conductance’s of α4β2 receptors. (A) Sample traces and open point amplitude histograms are shown for low [21.0±1.9 pS (mean±SD); HS receptor; gray traces] and high [30.6±3.3 pS (mean±SD); LS receptor; black traces] conductance channels recorded in 1 μM ACh.
Figure 2.1B: Unitary conductance’s of \( \alpha 4 \beta 2 \) receptors. (B) Fitted amplitude histograms and traces are shown for an LS receptor in 1 \( \mu M \) ACh [33.5\( \pm \)3.9 pS (mean\( \pm \)SD)] and a receptor in 1 \( \mu M \) ACh and 1 \( \mu M \) dFBr [32.3\( \pm \)5.1 pS (mean\( \pm \)SD)]. Inset shows example of occasional sublevel [dotted line at 1.38 pA (17.3 pS)] primarily seen in dFBr.
Figure 2.1C: Unitary conductance’s of α4β2 receptors. (C) I-V plot shows mean values (± SD) of HS (19.9±1.4 pS; n=13 cells; gray star) and LS (28.6±2.5 pS; n=19 cells; black circle, obscured) receptors recorded in 1 μM ACh and LS receptors in 1 μM ACh + 1 μM dFBr (28.4±2.0 pS; n=12 cells; blue square). Additional I-V data fitted with linear equations yielded slopes of ~30 pS and extrapolated reversal potentials near -3 mV.
Figure 2.2: Steady-state recordings of α4β2 receptors. Recordings shown are for 1 μM ACh (top) or 1 μM ACh + 1 μM dFBr (bottom). Contiguous traces represent 50 s of data filtered at 1 kHz (-3 dB, 4-pole Bessel).
Figure 2.3: Burst characteristics for α4β2 receptors. Distributions are fitted with two exponentials for 1 μM ACh (red; tcrit: 1.22 ms) and 1 μM ACh + 1 μM dFBr (blue; tcrit: 1.99 ms). Determination of critical time is described in methods. Distributions are plotted up to 140 ms for burst Length and 100 ms for burst open lifetime for comparison between control and dFBr. (A) Distributions for 1 μM ACh burst length (left; \( \tau_1: 2.86 \text{ms}, \tau_2: 42.1 \text{ms} \)) and burst open lifetime (right; \( \tau_1: 3.0 \text{ms}, \tau_2: 15.9 \text{ms} \)). (B) Distribution for 1 μM ACh + 1 μM dFBr burst length (left; \( \tau_1: 1.6 \text{ms}, \tau_2: 14.7 \text{ms} \)) and burst open lifetime (right; \( \tau_1: 0.75 \text{ms}, \tau_2: 10.9 \text{ms} \)). The presence of dFBr decreases burst length and open lifetime fast and slow time components.
Figure 2.4: Stability plots for $\alpha 4\beta 2\beta 2$ receptors. Stability plots obtained from QuB (50 events per segment, overlap of 25 events) for 0.1 (top) and 1 $\mu$M ACh (middle) and 1 $\mu$M ACh 1 $\mu$M dFB are displayed as segment vs occupancy of opening. An event is composed of either an opening or a closing. All patches are relatively stable with intermittent spikes in occupancy.
Figure 2.5: Models tested on $\alpha_{4}\beta_{2}$ receptors. Left: Examples of models tested, included are activation based mechanisms (from top) Katz and Thesleff, MWC, MWC with 2 desensitized, MWC with 3 desensitized and the desensitization models linear, and star. Right: Final model utilized with rate constants. Rate constants governing transitions $C2\rightarrow O1$, $O1\rightarrow O2$ and out of desensitized states D1 and D2 (circled) were significantly larger ($P < 0.05$) in dFBr.
Figure 2.6: Gating models for naive and modulated LS α4β2 receptors. Open and closed duration distributions from patches shown in Figure 2.2 for 0.1 μM ACh (top), 1 μM ACh (middle) and 1 μM ACh + 1 μM dFBr fitted with the gating model shown in Figure 2.5.
Figure 2.7A: Concentration dependence of potentiation of peak and steady-state currents. (A) Simulated currents are shown for the ACh (black traces) and dFBr (blue traces) models driven by 5 s jumps into 10 µM (top), 100 µM (middle), and 1 mM (bottom) ACh. Exponential functions fitted to the decay phase of currents for 1 mM ACh traces are superimposed (black traces). ACh model: $A_1 = 0.33$, $\tau_1 = 55$ ms, $A_2 = 0.45$, $\tau_2 = 220$ ms, $y_0 = 0.07$. dFBr model: $\tau = 157$ ms, $y_0 = 0.26$. Rise times (10 – 90%) for ACh and dFBr models with 1 mM ACh were 9.2 ms and 7.5 ms, respectively. Calibration bar: units of open probability ($P_{open}$).
Figure 2.7B: Concentration dependence of potentiation of peak and steady-state currents. (B) The channel open probability ($P_{\text{open}}$) of peak (stars) and steady-state (squares) responses are plotted for ACh (black) and dFBr (blue) models as a function of ACh concentration (1, 2.15, 4.64, 10, 21.5, 46.4, 100, 215, 464, 1000, 2150, 4640, and 10000 μM). The maximal (peak) $P_{\text{open}}$ for saturating ACh concentration was ~ 0.92 for ACh and 0.81 for dFBr. Slope and $EC_{50}$ values from fitted Hill equations were $0.76 \pm 0.04$ and $58 \pm 3.6$ μM for ACh, and $0.98 \pm 0.05$ and $17 \pm 1.1$ μM for dFBr.
Figure 2.7C: Concentration dependence of potentiation of peak and steady-state currents. (C) The ratio of dFBr model to ACh model $P_{\text{open}}$ is shown as a function of ACh concentration. Potentiation of peak response (stars) was largest at low ACh (101% increase for 4.64 μM ACh) and declined to zero at 1 mM ACh. At concentrations higher than 1 mM, dFBr responses were smaller (shown in B). Steady-state, desensitized responses (blue squares) were potentiated 3-3.5-fold above 10 μM ACh.
Figure 2.8: Occupancy probabilities using a concentration jump profile for $\alpha_4\beta_2$ receptors. See caption on next page.
Figure 2.8: Occupancy probabilities using a concentration jump profile for α4β2 receptors. Occupancy probabilities (y-axis) of channel states following a concentration jump are shown for 10 (top), 100 (middle), and 1000 (bottom) μM ACh for ACh (solid lines) and dFBr models (dashed lines). The maximum probability of occupancy of O1 was low (< 0.005) and changes in PO1 with dFBr were negligible; therefore, PO2 was used as an estimate of open probability. Occupancy probabilities of states C2 and C3 were also low, except for the first few milliseconds of the high concentration simulation where PC3 reached ~ 0.1 in dFBr.
Figure 2.9A: Slowly rising concentration profile decreases peak $P_{\text{open}}$ and increases potentiation with dFBr. (A) Occupancy probabilities are shown for a slow, exponential rise in ACh concentration according to: $c(t) = A \times (1 - \exp(-t/\tau))$ where $A = 300$ μM (EC75) and $\tau = 4$ s.
Figure 2.9B: Slowly rising concentration profile decreases peak $P_{\text{open}}$ and increases potentiation with dFBr. (B) Response to step increase in ACh concentration to 300 μM.
Figure 2.10A: Response to a pulse of a high concentration of agonist. Occupancy probabilities are shown for models driven by an alpha function to mimic a synaptic agonist concentration profile. (A) Responses are shown for the ACh model. Peak open probability ($P_{o2}$) was 0.28 at 3.7 ms compared with 0.45 at 3.6 ms for dFBr (shown in B; 1.61-fold potentiation). The decay of $P_{o2}$ was fitted with the sum of 3 exponentials ($A_1=0.72$, $\tau_1=44$ ms, $A_2=0.27$, $\tau_2=133$ ms, $A_3=0.012$, $\tau_3=5.4$ s, $y_0=0$). The decay of $P_{d1}$ was fitted with two exponential components ($A_1=0.98$, $\tau_1=147$ ms, $A_2=0.02$, $\tau_2=5.0$ s, $y_0=0$), and $P_{d2}$ was fitted with one component ($\tau=5.2$ s, $y_0=0$). Main inset shows $P_{o2}$ for the ACh model (black trace) with $P_{o2}$ for a modified model (blue trace) in which rates into desensitized states D1 and D2 were set to zero ($\delta_{1+} = \delta_{2+} = 0$). The decay time constant for the fitted exponential was 157 ms ($y_0=0$).
Figure 2.10B: Response to a pulse of a high concentration of agonist. (B) For the dFBr model the decay of $P_{O2}$ was fitted with two exponential components ($A_1=0.97$, $\tau_1=45$ ms, $A_2=0.02$, $\tau_2=649$ ms, $y_0=0$). $P_{D1}$ was fitted with two components ($A_1=0.97$, $\tau_1=76$ ms, $A_2=0.03$, $\tau_2=556$ ms, $y_0=0$), and $P_{D2}$ was fitted with a single exponential ($\tau=653$ ms, $y_0=0$). For dFBr, the inset shows that removal of desensitization had a negligible effect on the decay time course. The time constant ($\tau=50$ ms; $y_0=0$) was similar to the time constant for the fast, major component of decay of $P_{O2}$ with desensitization intact ($\tau=46$ ms).
Figure 2.11A: Development and recovery from desensitized states following a brief pulse of agonist concentration. Responses are shown for the same alpha function pulse in Figure 2.10. The fraction of occupied receptors (1-\(P_c\)) is nominally the same as the sum of receptors in O2, D1, and D2 (\(P_{O2}+P_{D1}+P_{D2}\)) for both models. Insets show responses on expanded time scales. (A) For the ACh model, desensitization (\(P_{D1}+P_{D2}\)) peaked at 197 ms. Relaxation of \(P_{D1}+P_{D2}\) was fitted with the sum of two exponentials with time constants (\(A_1=0.07, \tau_1=200\) ms, \(A_2=0.93, \tau_2=5.2\) s, \(y_0=0\)) similar to relaxation time constants for the components \(P_{D1}\) and \(P_{D2}\) fitted individually.
Figure 2.11B: Development and recovery from desensitized states following a brief pulse of agonist concentration. (B) For dFBr, $P_{D1 + D2}$ peaked at 69 ms, and relaxation of $P_{D1 + D2}$ was well fitted with a single exponential with a time constant ($\tau=654$ ms) identical to that for that for $P_{D1}$ alone (Figure 2.10).
Table 2.1. Rate constants used in simulations for $\alpha_{4}\beta_{2}$ receptors. Errors are the relative estimated error in the fits.

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<th>(+) dFBr</th>
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Table 2.2. Mean values for $\alpha_4\beta_2$ receptor rate constants. Errors are the standard deviations between the three patches.

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2.6 References


Unravelling the mechanism of action of NS9283, a positive allosteric modulator of

patch-clamp techniques for high-resolution current recording from cells and cell-


Unraveling the high- and low-sensitivity agonist responses of nicotinic

Joshi, P.R., A. Suryanarayanan, and M.K. Schulte. 2004. A vertical flow chamber for
Xenopus oocyte electrophysiology and automated drug screening. *J Neurosci


2007. Synthesis of desformylflustrabromine and its evaluation as an alpha4beta2

Li, P., and J.H. Steinbach. 2010. The neuronal nicotinic alpha4beta2 receptor has a

Maconochie, D.J., and J.H. Steinbach. 1998. The channel opening rate of adult- and
fetal-type mouse muscle nicotinic receptors activated by acetylcholine. *J.

Marabelli, A., R. Lape, and L. Sivilotti. 2015. Mechanism of activation of the prokaryotic


Nicolai, C., and F. Sachs. 2013. Solving ion channel kinetics with the QuB software. Biophysical Reviews and Letters. 08:191-211.


Figure A.2.1: LS Katz and Thesleff model fits. Open and closed duration distributions from patches for 0.1 μM (top) and 1 μM ACh (bottom) group fitted with gating model Katz and Thesleff shown in Figure 2.5.
Figure A.2.2: LS MWC model fits. Open and closed duration distributions from patches for 0.1 μM (top) and 1 μM ACh (bottom) group fitted with gating model MWC shown in Figure 2.5.
Figure A.2.3: LS modified MWC model fits. Open and closed duration distributions from patches for 0.1 μM (top) and 1 μM ACh (bottom) group fitted with gating model modified from classical MWC mechanism shown in Figure 2.5 (third from the top). Model is modified to contain two desensitized states off of the mono and di-liganded open states.
Figure A.2.4: LS modified MWC model fits with three desensitized states. Open and closed duration distributions from patches for 0.1 μM (top) and 1 μM ACh (bottom) group fitted with gating model modified from classical MWC mechanism shown in Figure 2.5 (fourth from the top). Model is modified to contain three desensitized states. One off of the doubly liganded closed state and an additional two off of the di-liganded open states.
Figure A.2.5: LS linear model fits. Open and closed duration distributions from patches for 0.1 μM (top) and 1 μM ACh (bottom) group fitted with gating model lacking activation and containing 5 desensitized states shown in Figure 2.5 (fifth from the top). Model lacks activation and effectively begins in the doubly liganded open state. This model assumes that we are not seeing any activation rates in our data.
Figure A.2.6: LS star model fits. Open and closed duration distributions from patches for
0.1 $\mu$M (top) and 1 $\mu$M ACh (bottom) group fitted with gating model lacking activation
and containing 5 desensitized states in a star shape shown in Figure 2.5 (bottom).
Model lacks activation and effectively begins in the doubly liganded open state.
Table A.2.1: Rate constants for Katz and Thesleff mechanism.

![Chemical Reaction Diagram]

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Table A.2.2: Rate constants for MWC mechanism.

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Table A.2.3: Rate constants for a MWC modified mechanism

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<td>s$^{-1}$</td>
<td>0.12 ± 33%</td>
<td>0.04 ± 42%</td>
</tr>
</tbody>
</table>
Table A.2.4: Rate constants for a three desensitized state MWC mechanism.

\[
\begin{align*}
C \xrightarrow{2k_{+1}} C & \xrightarrow{k_{+1}} C \xrightarrow{2k_{-1}} C \\
\xrightarrow{a_1} O & \xrightarrow{\beta_1} O \xrightarrow{a_2} O \\
\xrightarrow{d_1} C & \xrightarrow{d_2} C \xrightarrow{d_3} C
\end{align*}
\]

<table>
<thead>
<tr>
<th>Rate Constant</th>
<th>Unit</th>
<th>0.1 ( \mu M )</th>
<th>1 ( \mu M )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( 2k_{+1} )</td>
<td>M(^{-1})s(^{-1})</td>
<td>1.05E9 ( \pm ) 36%</td>
<td>3.1E7 ( \pm ) 32%</td>
</tr>
<tr>
<td>( k_{+1} )</td>
<td>M(^{-1})s(^{-1})</td>
<td>3.7E9 ( \pm ) 85%</td>
<td>3.8E8 ( \pm ) 33%</td>
</tr>
<tr>
<td>( 2k_{-1} )</td>
<td>s(^{-1})</td>
<td>758 ( \pm ) 85%</td>
<td>419 ( \pm ) 50%</td>
</tr>
<tr>
<td>( k_{-1} )</td>
<td>s(^{-1})</td>
<td>296 ( \pm ) 22%</td>
<td>350 ( \pm ) 99%</td>
</tr>
<tr>
<td>( \beta_1 )</td>
<td>s(^{-1})</td>
<td>0.001 ( \pm ) 99%</td>
<td>74 ( \pm ) 73%</td>
</tr>
<tr>
<td>( \alpha_1 )</td>
<td>s(^{-1})</td>
<td>9.8e-5 ( \pm ) 99%</td>
<td>855 ( \pm ) 55%</td>
</tr>
<tr>
<td>( \beta_2 )</td>
<td>s(^{-1})</td>
<td>2710 ( \pm ) 81%</td>
<td>3570 ( \pm ) 43%</td>
</tr>
<tr>
<td>( \alpha_2 )</td>
<td>s(^{-1})</td>
<td>315 ( \pm ) 82%</td>
<td>79 ( \pm ) 44%</td>
</tr>
<tr>
<td>( k_{+2} )</td>
<td>M(^{-1})s(^{-1})</td>
<td>1E9 ( \pm ) 81%</td>
<td>2E4 ( \pm ) 99%</td>
</tr>
<tr>
<td>( k_{-2} )</td>
<td>s(^{-1})</td>
<td>279 ( \pm ) 83%</td>
<td>4E-5 ( \pm ) 99%</td>
</tr>
<tr>
<td>( d_{+1} )</td>
<td>s(^{-1})</td>
<td>44 ( \pm ) 23%</td>
<td>82 ( \pm ) 18%</td>
</tr>
<tr>
<td>( d_{-1} )</td>
<td>s(^{-1})</td>
<td>4.5 ( \pm ) 25%</td>
<td>0.14 ( \pm ) 19%</td>
</tr>
<tr>
<td>( d_{+2} )</td>
<td>s(^{-1})</td>
<td>6.8 ( \pm ) 91%</td>
<td>0.59 ( \pm ) 99%</td>
</tr>
<tr>
<td>( d_{-2} )</td>
<td>s(^{-1})</td>
<td>0.11 ( \pm ) 54%</td>
<td>0.003 ( \pm ) 66%</td>
</tr>
<tr>
<td>( d_{+3} )</td>
<td>s(^{-1})</td>
<td>262 ( \pm ) 25%</td>
<td>332 ( \pm ) 20.5%</td>
</tr>
<tr>
<td>( d_{-3} )</td>
<td>s(^{-1})</td>
<td>0.69 ( \pm ) 43%</td>
<td>96 ( \pm ) 1%</td>
</tr>
</tbody>
</table>
Table A.2.5: Rate constants for a linear mechanism.

![Diagram of the linear mechanism]

<table>
<thead>
<tr>
<th>Rate Constant</th>
<th>0.1 μM</th>
<th>1 μM</th>
</tr>
</thead>
<tbody>
<tr>
<td>$k_{10}$</td>
<td>1817 ± 10%</td>
<td>3017 ± 9.6%</td>
</tr>
<tr>
<td>$k_{01}$</td>
<td>87 ± 5.4%</td>
<td>188 ± 5.6%</td>
</tr>
<tr>
<td>$k_{12}$</td>
<td>1086 ± 10%</td>
<td>1062 ± 11%</td>
</tr>
<tr>
<td>$k_{21}$</td>
<td>47 ± 24%</td>
<td>129 ± 65%</td>
</tr>
<tr>
<td>$k_{23}$</td>
<td>49 ± 36%</td>
<td>514 ± 43%</td>
</tr>
<tr>
<td>$k_{32}$</td>
<td>8.8 ± 32%</td>
<td>23 ± 35%</td>
</tr>
<tr>
<td>$k_{34}$</td>
<td>2 ± 36%</td>
<td>12 ± 38%</td>
</tr>
<tr>
<td>$k_{43}$</td>
<td>0.84 ± 43%</td>
<td>1.9 ± 32%</td>
</tr>
<tr>
<td>$k_{45}$</td>
<td>0.1 ± 99%</td>
<td>0.27 ± 46%</td>
</tr>
<tr>
<td>$k_{54}$</td>
<td>0.14 ± 67%</td>
<td>0.185 ± 25%</td>
</tr>
</tbody>
</table>
Table A.2.6: Rate constants for a star mechanism.

<table>
<thead>
<tr>
<th>Rate Constant</th>
<th>$0.1 \mu M$</th>
<th>$1 \mu M$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$k_{10}$</td>
<td>612 ± 41%</td>
<td>99 ± 42%</td>
</tr>
<tr>
<td>$k_{01}$</td>
<td>8.9 ± 37%</td>
<td>3.6 ± 18%</td>
</tr>
<tr>
<td>$k_{20}$</td>
<td>3.3 ± 24%</td>
<td>0.13 ± 11%</td>
</tr>
<tr>
<td>$k_{02}$</td>
<td>13 ± 14%</td>
<td>11 ± 14%</td>
</tr>
<tr>
<td>$k_{30}$</td>
<td>43 ± 29%</td>
<td>2.11 ± 44%</td>
</tr>
<tr>
<td>$k_{03}$</td>
<td>9.8 ± 17%</td>
<td>7.2 ± 16%</td>
</tr>
<tr>
<td>$k_{40}$</td>
<td>0.25 ± 18%</td>
<td>0.97 ± 99%</td>
</tr>
<tr>
<td>$k_{04}$</td>
<td>6.7 ± 20%</td>
<td>1e-4 ± 99%</td>
</tr>
<tr>
<td>$k_{50}$</td>
<td>3696 ± 14%</td>
<td>3553 ± 5.6%</td>
</tr>
<tr>
<td>$k_{05}$</td>
<td>53 ± 9.6%</td>
<td>62 ± 5.8%</td>
</tr>
</tbody>
</table>
CHAPTER 3: Synchronization of High-Sensitivity $\alpha 4\beta 2$

Nicotinic Acetylcholine Receptors

Abstract

Understanding the mechanism of modulation of nAChRs by potential compounds such as desformylflustrabromine (dFBr) allows for greater insight into the potential uses and drawbacks of the drug on a synaptic level. We tested a variety of classical mechanisms to determine the best kinetic representation of our single-channel data. Our data for the human high sensitivity, $(\alpha 4\beta 2)3$, nicotinic receptors was best explained using an 8-state MWC based gating model. It was found that dFBr increased the open probability via an increase in the opening frequency of the diliganded open state in dFBr (by increasing the opening rate of the diliganded receptor). Also, dFBr increased the mean open time of the mono-liganded open state. In order to determine the consequences of the effects of dFBr on signaling at cholinergic synapses, we simulated macroscopic responses to concentration jumps. At high simulated concentrations of ACh, potentiation was primarily exhibited through the increase in open frequency into the diliganded open state, while at low concentrations of ACh, potentiation primarily occurred through the increase in mean open time for the mono liganded open state. For 1 mM ACh simulated macroscopic responses, we found an increase in peak $P_{\text{open}}$ in the presence of dFBr, increasing $P_{\text{open}}$ from 0.4 to 0.6. The increase in peak $P_{\text{open}}$ in the presence of dFBr is due to the synchronization of openings, however, dFBr reduces overall charge transfer at simulated concentrations above 1 $\mu$M ACh. The simulated responses demonstrated that the decrease in charge transfer is primarily due to an increase in probability of being in the desensitized state D2. The negative effect dFBr is having on
the high sensitivity receptor could be due to the presence of the Good’s buffer HEPES in the recording solution. Future studies will need to consider the buffer’s potential effects when analyzing results.²

3.1 Introduction

The current knowledge on the physiological role and activation of $\alpha_4\beta_2$ nicotinic acetylcholine receptors (nAChRs) in the central nervous system (CNS) is limited. The heteromeric $\alpha_4\beta_2$ AChRs are widely distributed throughout the brain and may be involved in nicotine addiction as well as a variety of neurological disorders, including Alzheimer’s disease and autism. The high-sensitivity (HS) $\alpha_4\beta_2$ nAChR is the least studied of the two stoichiometric forms and very little is known about their synthesis and regulation (Changeux, 2010). Due to its potential role in several psychiatric and neurological disorders, HS $\alpha_4\beta_2$ nAChRs are an important therapeutic target. (Nelson et al., 2003; Khiroug et al., 2004).

A novel strategy in treating a variety of cognitive diseases is to utilize compounds that enhance responses of nAChRs and potentially increase transmission at synapses where nAChRs are deficient, thus restoring functional properties. Allosteric modulators are a class of compounds that bind at a site distinct from the endogenous ligand acetylcholine, and therefore do not compete with the agonist, preserving the natural cholinergic signal. Desformylflustrabromine (dFBr) has been previously identified as a novel positive allosteric modulator (PAM) of $\alpha_4\beta_2$ nAChRs with no apparent potentiation of other subtypes (Sala et al., 2005). Previous studies have suggested that the potentiation exhibited by dFBr is due to an alteration in channel gating kinetics, but the mechanism by which dFBr modulates the HS $\alpha_4\beta_2$ receptor is currently unknown (Weltzin and Schulte, 2010; Sala et al., 2005; Kim et al., 2007).
In this study we are taking a first step in determining how dFBr modulates by providing a comprehensive gating model for the high sensitivity receptors. Once it is known how these receptors are activated in the CNS, then our model will provide a framework for how dFBr will potentiate on a synaptic level. Obtaining a mechanism that can describe the effects of the positive allosteric modulator, dFBr, will allow greater insights into its uses and potential shortcomings in future studies.

3.2 Materials and Methods

3.2.1 Cell Techniques

HEK-293 cells were maintained in minimum essential Eagle medium (MEME) supplemented with 1% Glutamax (Gibco), 10% bovine growth serum, antibiotic penicillin streptomycin (200 µg ml⁻¹), 100 µg/mL G418 (Thermo-Fisher) and pH adjusted to 7.4. Cells were incubated at 37°C in a humidified atmosphere of 5% CO₂, passaged weekly and plated onto 35mm tissue culture dishes (Becton Dickinson) one day prior to experiments. Human α4 and β2 neuronal nicotinic receptor subunits were stably transfected into HEK-293 cells and passaged weekly one day prior to experiments. The cells were provided to us by Dr. Henry Steinbach at Washington University, St. Louis, MO.
3.2.2 Electrophysiology Techniques

All salts were obtained from Sigma Aldrich. To avoid receptor rundown all patch-clamp recordings were made in the cell-attached configuration at room temperature. Membrane potential was zeroed with a K+-based extracellular solution containing (in mM): 142 KCl, 5.4 NaCl, 2 CaCl₂, 1.7 MgCl₂, 5 HEPES, 10 glucose and pH adjusted to 7.4. Pipette solution contained (in mM): 5.4 KCl, 142 NaCl, 2 CaCl₂, 1.7 MgCl₂, and 5 HEPES pH adjusted to 7.4. Agonist and PAM solutions were made on the day of the experiment.

Experiments were performed using a Nikon Eclipse FN1 microscope. Pipettes were pulled to a resistance of 10-14 MΩ using a PMP-102 Micropipette puller (Microdata Instrument Inc.) and thick-walled borosilicate capillary glass tubing with flame polished ends (Warner Instruments). In order to enhance signal to noise ratio pipette tips were coated with Sylgard (Dow Corning 184). Prior to recording the pipettes were front and back filled with the pipette solution containing the appropriate concentration of ACh (0.1 -10 μM) with or without dFBr (1 μM). Patches were voltage-clamped at -80 mV with a HEKA EPC-10 Elektronik amplifier using PATCHMASTER software. Data were low-pass filtered (4-pole Bessel, -3dB) at 10 kHz and digitized to the computer at 100 kHz.
3.2.3 Definition of Bursts and Clusters

Bursts were defined as in section 2.2.4 and by a critical value \((t_{\text{crit}})\) using the equation:

\[
Amp_1 e^{-\text{crit} / \tau_{\text{au}1}} = Amp_2 (1 - e^{-\text{crit} / \tau_{\text{au}2}})
\]

\(T_{\text{crit}}\) is the cutoff duration between exponential components and equalizes the area under the overlapping tails of the distributions. Where \(Amp_1\) and \(Amp_2\) are the weights of the first and second components respectively for bursts or second and third components for clusters. \(\tau_{\text{au}1}\) and \(\tau_{\text{au}2}\) are the time constants computed from MIL. Each \(\tau\) in the distribution contributes one exponential component to the probability density function (thick line in distributions). Bursts and clusters that held doubles were discarded from the analysis.

3.2.4 Model Fitting and Simulations

Idealized records of single-channel events were constructed and dwell time distributions fitted with models in QuB as previously described. Rate constants were obtained using an algorithm (MIL described) (Qin, 2004; Qin et al., 1997, 1996). All simulations performed utilized scheme 2 for control (1 \(\mu\)M ACh) and modulated (1\(\mu\)M ACh + 1 \(\mu\)M dFBr). Responses were made using rate constants from a group fit that contained greater than \(n=3\) for all conditions, except 10 \(\mu\)M data. The open and closed distribution used in figures was based on patches that contained the most events.
(openings + closings). In order to distinguish between models, we relied on various factors including, the fit of the probability density function to the data, the degree of agreement in activation steps between 0.1 and 1.0 μM ACh, as well as simulated peak $P_{\text{open}}$ and EC$_{50}$ values. Log likelihood was not used as a best fit parameter due to its inability to distinguish between models containing a similar number of states (Elenes and Auerbach, 2002). Monte Carlo simulations of macroscopic currents in Figure 3.4A were made in QuB Express for 10,000 channels with a unitary amplitude of 1 pA and standard deviation (SD) of Gaussian noise of 0.15 pA (Nicolai and Sachs, 2013). Current values were converted to and displayed as open probabilities ($P_{\text{open}}$) in Igor Pro (WaveMetrics). Stability plots were generated using QuB’s stability function with a window size (events) of 15 and an overlap (events) of 2.
3.3 Results

3.3.1 Conductance of HS Receptors in the Presence of dFBr

Allosteric modulators can exhibit potentiation a variety of ways, including increasing an ion channels conductance. To determine whether dFBr increases chord conductance as a means of potentiating the receptor channel openings with one primary amplitude were analyzed from cell-attached recordings obtained using low concentrations of ACh (0.1 and 1.0 µM ACh). Similar to previous reports we found that α4β2 receptors demonstrated a high sensitivity component to ACh, having a conductance of ~20 pS (Buisson and Bertrand, 2001). Figure 3.1 shows open point amplitude histograms for HS receptors in 1 µM ACh and 1 µM ACh + 1 µM dFBr. Fitted open point amplitude histograms yielded a chord conductance of 19.5±1.8 pS (mean±SD; n=5) and 19.9±0.34 pS (n=4) in the absence and presence of dFBr respectively. The presence of dFBr had no significant effect on the HS conductance levels. Traces (1 s) reveal that dFBr does not significantly alter the mean open time (the mean duration of a single opening). This suggests that when dFBr is applied the average open state lifetime of the receptor is not affected by the drug. Inward rectification of α4β2 has been observed in previous reports (Buisson et al., 1996) and our recordings did not exhibit HS openings above 0 mV. I/V data was collected for 1 µM ACh (n=3) and 1 µM ACh + 1 µM dFBr (n=4) between -80 mV and -30 mV and was fitted with a linear equation. The fit yielded slope conductances of 19.9 for ACh and 19.8 for ACh + dFBr. The I/V relationships demonstrate that dFBr does not increase the chord conductance as a means of potentiation. The control values are very close to the
previously reported conductance values of ~20 pS (Buisson et al., 1996). Long
contiguous sweeps (50 s) do not show the obvious effect of dFBr on HS receptor gating.

3.3.2 Gating of HS Receptors in ACh and ACh + dFBr

Recordings of HS α4β2 in 0.1 (n=4), 1 (n=4), 10 (n=2) μM ACh or 1 μM ACh + 1
μM dFBr (n=3) yielded multiple patches for all conditions that contained enough
openings for analysis. The average number of openings for all ACh patches were
917±395, with a range from 295 to 1250 openings. The average number of openings
for 1 μM ACh + 1 μM dFBr were 1698 openings. Incidents of double openings were
observed in 1 out of 3 of the 0.1 or 1 μM ACh patches, but all other conditions showed
evidence for more than one channel. The probability of a channel being open for the
HS subtype is very low (~ 0.001 to 0.005). These values are taken from the most stable
0.1 and 1.0 μM ACh patches, which contained two observable channels. Figure 3.2
shows 50s segments for 1 μM ACh and 1 μM ACh + 1 μM dFBr. These two data sets
show HS receptor gating for ACh with and without dFBr. Visually an effect is not
immediately apparent.

Single-channel recordings often contain several channel openings in succession
followed by long periods of closings. Our HS data contained none of these classical
bursts, and most ‘bursts’ in our data contained only one opening. Figure 3.3 shows the
burst length and apparent open lifetime of the bursts for 0.1-10 μM ACh and 1 μM ACh
+ 1 μM dFBr. Two exponentials were needed to fit the distribution of the burst length
and also to fit the total open time per burst. Both of these were very similar since the
majority of bursts consisted of one apparent opening. As the concentration increased, the slower component (time constant $\tau_s$) becomes shorter. The presence of dFBr increased the apparent open lifetime, increasing time constants $\tau_f$ from 2.9 to 4.5 ms and $\tau_s$ from 3.7 and 10.7 ms. There is a similar effect seen on the burst length increasing constants $\tau_f$ from 3.12 to 4.5 ms and $\tau_s$ from 3.4 and 5.04 ms. Our calculated burst parameters were in agreement for dFBr (estimated burst open lifetime 4.7 ms) with the time constants obtained from the fits, whereas our ACh calculated burst parameters were larger than our fitted time constants (estimated burst open lifetime 12.9 ms). A possible explanation for this discrepancy is an underestimation in one of the desensitized states by the model. Clusters were defined by a $t_{crit}$ in between 2nd and 3rd components (see methods). Cluster length (Figure 3.4 A) and apparent open lifetime (Figure 3.4 B) required two exponentials to fit. For burst length $\tau_f$ showed no significant difference between 0.1 to 10 $\mu$M ACh, whereas $\tau_s$ decreased from 4.01 ms to 1.83 ms for 0.1 to 10 $\mu$M ACh. The slow and fast time constants for the cluster open lifetimes decreased as concentration increased from 0.78 to 0.33 ms for $\tau_f$ and 2.1 to 1.6 ms for $\tau_s$ for 0.1 to 10 $\mu$M ACh. Clusters were dominated primarily by single openings and lower concentrations having more openings (3.4 C).

It is a common problem in the single-channel analysis of low $P_{open}$ receptors to have a level of uncertainty in the number of channels within the data. Therefore, it is beneficial to determine if instability exists in the data and how that will affect kinetic analysis. Stability plots for ACh concentrations of 0.1 $\mu$M to 10 $\mu$M ACh and 1 $\mu$M ACh + 1 $\mu$M dFBr were generated using QuB’s stability function and segments contained 15
events (Figure 3.5, see methods). These plots originated from the same patches used in model analysis, as well as burst and cluster distributions. Both 0.1 and 1 μM ACh data sets contained more than one observable channel and were stable with a $P_{\text{open}}$ of 0.0067 and 0.016 respectively. The 0.1 and 1 μM ACh patches were the only completely stable patches despite showing more than one observable channel. Other 0.1 and 1 μM ACh patches had similar low occupancy (<0.01), and contained only one observable channel, but had decreased activity after ~800 segments. The 10 μM ACh patches activity decreased after ~500 segments. The 1 μM ACh + 1 μM dFBr patches were not stable near the end of the recordings and activity decreased after ~1000-1200 segments. The observed occupancy for the 10 μM ACh (0.0011) data was lower when compared to the 0.1 and 1 μM patches, even though more than one channel was observed in the 10 μM data. All 1 μM ACh + 1 μM dFBr patches showed some amount of instability ($P_{\text{open}}= 0.05$, Figure 3.5). This instability is likely caused by channels dropping out as the patch progresses or receptors entering an extremely long lived desensitized state that is concealed by more than one channel being present in the patch. Instability due to channels dropping is the most likely culprit with all of the dFBr data, but this is less likely for the 10 μM data, which begins with very low occupancy and contain more than one observable channel.
3.3.3 Fitting Mechanisms to Single-Channel Data

The mechanism of modulation exhibited by dFBr was investigated by fitting a variety of mechanisms to single-channel data. Mechanisms that contained activation steps were fitted independently to each single-channel concentration group which included 0.1, 1.0 and 10 µM ACh and 1 µM ACh + 1 µM dFBr. A concentration group was defined as all patches of the same concentration, (e.g. all 1 µM), fitted simultaneously as a group. The records were kept intact and were not divided up due to bursts and clusters primarily consisting of single openings (see methods). We tested 4 different types of activation mechanisms, Katz and Thesleff based model, Monod-Wyman-Changeux (MWC) based model, and MWC mechanisms that contained two or three desensitized states. In our MWC models we omitted the unliganded open state due to the low probability of these openings occurring in naïve HS nAChRs. All models utilized activation rate constant constraints of $2k_{+1}$ and $k_{+1}$ for the binding rates, and $2k_{-1}$ and $k_{-1}$ for the unbinding rates, which assumes that both sites on the receptor are available for agonist binding. Mechanisms were also constrained with the requirements of microscopic reversibility.

Classical mechanisms such as the Katz and Thesleff and the MWC have been previously utilized to describe the gating kinetics of ACh receptors. We first tested the Katz and Thesleff based mechanism (See Figure 3.6) and found that 10 µM errors were extremely high (>100%) and most of the rates were vastly different from the other patches (See Table 1). Activation rates between 0.1 and 1 µM ACh changed 5-fold, while the activation rates for the 10µM data were 1000x smaller than for the other two
concentrations. The overall fits of the probability density function to the data was decent, with 10 μM ACh data fitting the worst (Figure 3.7). Deactivation rates in the Katz and Thesleff model for the dFBr data are much lower when compared to controls, while β2 is 10x lower than control. When simulated, the Katz and Thesleff configuration gave extremely high $P_{\text{open}}$ (~0.85) values and the presence of dFBr displayed inhibition.

The MWC mechanism has been widely used to describe many ion channels. This mechanism postulates that there are two conformational states a channel can take on, open and closed, each with different affinities for ligand. For the MWC we imposed constraints so that the activation rate constants are equivalent for each ligand bound. For this model type we have two molecules of agonist binding, with one cycle, and 6 free parameters. Of the two classical mechanisms the MWC model gave consistent activation rates between the 0.1 and 1 μM ACh data. Examples of open and shut time distributions are shown in Figure 3.8. Since the MWC scheme lacks desensitized states the fits were poor and thus we decided to use the MWC as a base model. One to three desensitized states were added to the basic model and their positions tested. Positions included a D state distal to an O state or to the diliganded closed state.
3.3.4 Modified MWC Schemes

It was difficult to find a desensitization configuration that would result in a slow desensitization time course as observed in previous reports (Weltzin et al., 2014). We first tested multiple two-desensitized state configurations, and scheme 1 is an example of one such configuration (Figure 3.6). This particular configuration gave the lowest errors, unlike configurations that had desensitized states distal to O1 and O2 or C3 and O2. With only two desensitized states, placing one distal to O1 and O2 resulted in no observable desensitization in simulated macroscopic responses. While having the desensitized states distal to C3 and O2 resulted in large error values and the lifetime of C3 being much lower than our imposed dead time (Figure A.3.1 and Table A.3.1). Closed and open time distributions are shown for scheme 1 in Figure 3.9 and rate constants in Table 3.3. Multiple configurations that contained three desensitized states were also tested. Schemes included all three desensitized states off O2 as well as a desensitized state off of O1 and two off of O2. Both configurations resulted in large errors for all conditions and poor fits to the closed time distributions. All three desensitized states being adjacent to O1, O2 and C3 was also tested, but the errors and fits gave us less confidence in the mechanism compared to scheme 2 (Figure 3.10 and Table 3.4).

All of the MWC based models that were tested gave a consistent high $P_{\text{open}}$ (>0.6) for 0.1 μM ACh data. The current cause for this is most likely due to larger rate constants between O1 and O2, as lowering those rates in simulations to values seen in the 1 μM ACh model resulted in a lower simulated $P_{\text{open}}$ values, which were comparable to the 1 μM data (~0.3-0.4).
3.3.5 Desensitization Based Mechanisms

We also investigated the possibility that our data lacks activation information due to the closed time distributions only displaying closings affiliated with entry into long lived closed states. When agonist concentrations are high, then almost all current can be attributed to the activation of diliganded AChRs (Elenes and Auerbach, 2002). High ACh concentrations can therefore lead to diliganded closed states being too brief to be detected. Mechanisms that only contained desensitized states as closed states were investigated and group fitted to data at different ACh (0.1, 1 and 10 μM ACh) and ACh + dFBr concentrations (1 μM ACh + 1 μM dFBr). Mechanisms are based off of QuB’s model search function and were the most common outcomes.

All desensitization models contain only one open state, which results in the probability density function not fitting the open duration distributions. Most group fits of the desensitization based models do not fit 10 μM data as well as the 0.1 or 1 μM ACh data. Scheme 5 (star) (Figure A.3.4) had the best overall visual fits compared to the other desensitization schemes. Whereas, scheme 6 (linear, Figure A.3.5) and scheme 7 (Figure A.3.6) had the worst fits. The 10 μM data fits were poor for all desensitization schemes. In some of the schemes the lifetimes of all of the desensitized states increased with dFBr, including scheme 4 and scheme 7, with scheme 7 fits being equivalent to scheme 6. For all of the schemes it was noted that dFBr reduced the likelihood of entry into the majority (4/5 or 3/4) of the desensitized states within the models. Also, there were cases in all of the desensitized models where dFBr increased the lifetime of some of the desensitized states. Of all of the desensitization schemes
the most probable would be Scheme 5 (star) due to the probability density function fitting the best out of all three and lowest overall errors modulated and unmodulated.

### 3.3.6 Simulation of Macroscopic Currents

Conducting simulations of macroscopic responses with fitted models is a powerful tool in determining if the estimated kinetic parameters support potentiation. Scheme 2 was utilized to investigate dFBr’s potentiation over a range of ACh concentrations by simulating macroscopic responses to concentration jumps (Figure 3.11A). Potentiation of peak currents occurred at all concentrations with the largest percentile increase occurring at low concentrations (≤5 μM), and an increase in peak current maintained at larger concentrations (≥ 50 μM) (Figure 3.11 B). Fitted Hill equations to peak responses yielded (mean ±SD) EC$_{50}$ values and Hill slopes of 23.5±1.3 μM and 1.1±0.05 for ACh, and 18.1±0.68 μM and 0.91±0.03 for dFBr. Peak open probability for 1 mM concentration jumps of ACh was ~0.42, dFBr increased peak open probability to ~0.63. The concentration dependence of changes induced by dFBr for macroscopic responses is shown for 1 μM-1 mM ACh (Figure 3.11 C). The decay of responses at all concentrations were best fit with two exponentials. At concentrations of 1 mM, ACh gave a fast time component of ~17 ms (τ$_{f}$) and a slow time component (τ$_{s}$) of ~790 ms. The presence of dFBr significantly reduces these time components, bringing τ$_{f}$ down to 5 ms and τ$_{s}$ down to 6 ms.
We simulated low concentrations of ACh using scheme 2 for ACh and ACh + dFBr in order to determine if dFBr affected low concentrations of agonist the same as high concentrations. Figure 3.12 A shows the effect of dFBr at low concentrations of ACh (1, 3 and 5 μM). At concentrations greater than 3 μM, dFBr gives increased desensitization, whereas at concentrations lower than 1 μM, dFBr increased potentiation without increasing desensitization. We were interested to determine if charge transfer changed significantly in the presence of dFBr, and if that change is correlated to agonist concentration. At ACh concentrations of 1 mM dFBr decreases charge transfer 5.2-fold down to 1.8-fold at 5 μM. Charge transfer did not increase in the presence of dFBr until concentrations were ≤1 μM. We were interested to find out if dFBr has a greater impact on the fast (τf) or slow (τs) time component. While dFBr does decrease τf for the HS receptors, it was found that τs was more affected by dFBr. For the ACh model, τf maintains a steady level at ~100 μM reaching ~17 ms, while in dFBr this occurs at lower concentrations (<10 μM) (Figure 3.12 B). dFBr sees a significant reduction in the slow time course of decay, reducing τs to one tenth its ACh value at concentrations greater than 50 μM (Figure 3.12 C).

When individual rate constants were changed during macroscopic simulations it was found that potentiation was different based upon the concentration of ACh used for stimulus. At high (1mM) simulated ACh concentrations the primary mode of potentiation exhibited by dFBr is an increased rate of entry into the diliganded open state O2. This potentiation is offset by an increase into the desensitized state D2, accounting for the observed increase in desensitization as the simulated concentration of ACh increases.
(Figure 3.11A). At low concentrations (<10 μM) potentiation primarily occurs by an increase in the mean open lifetime of O1. dFBr increases the open probability, synchronizing openings in simulated macroscopic currents, giving potentiation at the peak.

3.3.7 Occupancy Probabilities and Agonist Concentration

The aim of examining occupancy probabilities was to determine how dFBr increased macroscopic desensitization, and if potentiation was equivalent at all concentrations. Figure 3.13 shows the occupancy probabilities for concentration jumps into 10 μM, 100 μM, and 1 mM ACh. At high concentrations (1mM) the open probability is approximated by $P_{O2}$ for ACh and ACh + dFBr models, because $P_{O1}$ is <0.005, as the simulated concentration decreases, $P_{O1_{ACh}}$ increases to 0.032. At high concentrations, dFBr increased the peak response 1.6-fold ($P_{O2_{ACh}} = 0.38$, $P_{O2_{dFBr}} = 0.601$). For jumps into low concentration of ACh (10 μM) dFBr increased open probability by a factor of 2.3 ($P_{O2_{ACh}} + P_{O1_{ACh}} = 0.1$, $P_{O2_{dFBr}} + P_{O1_{dFBr}} = 0.23$).

For concentrations greater than 100 μM an increase in $P_{O2}$ was the primary contributor to potentiation, whereas below 100 μM $P_{O1}$ becomes a significant contributor. For all concentrations $P_{D2}$ and $P_{D3}$ both contributed to desensitization with $P_{D3}$ contributing the most in ACh, whereas $P_{D2}$ contributed the most in dFBr. In all cases $P_{D1}$ contributed very little to desensitization. For concentrations greater than 100 μM $P_{D2_{dFBr}}$ was larger than $P_{D2_{ACh}}$, and is the primary cause for increased desensitization seen by dFBr in macroscopic simulations. $P_{D3_{dFBr}}$ is < 0.1 compared to $P_{D3_{ACh}}$ being
>0.65 at 0.500 ms. The lower probability of entering D3 leads to more entries into O2 contributing to initial potentiation.

### 3.3.8 Potentiation During a Pulse of High Agonist Concentration

We examined the macroscopic responses to an agonist concentration profile typical of fast synapses by driving models with an alpha function with an agonist concentration of 1mM. Peak open probability in ACh ($P_{O2_{ACh}} = 0.35$) was slightly lower compared to concentration jumps ($P_{O2_{ACh}} = 0.38$) of the same concentration. $P_{open}$ was larger for dFBr when a similar comparison was made ($P_{O2_{dFBr}} = 0.634$) when driven with an alpha function versus ($P_{O2_{dFBr}} = 0.601$) when driven with a concentration jump. Overall, dFBr yielded a 1.8-fold increase in peak amplitude.

The development of desensitization following a pulse of agonist was faster with dFBr, while recovery from desensitization was significantly slower. The presence of dFBr decreased the time to peak of $P_{D2}$ ($P_{D2_{ACh}} = 41.2$ ms, $P_{D2_{dFBr}} = 27$ ms) and $P_{D3}$ ($P_{D3_{ACh}} = 8.1$ ms, $P_{D3_{dFBr}} = 5.3$ ms). $P_{D2_{dFBr}}$ was lower with the alpha function versus the concentration jump, while $P_{D3}$ maintained similar levels, despite having a longer decay time course in the presence of dFBr ($P_{D3_{ACh} \tau=6.7}$ s, $P_{D3_{dFBr} \tau=21.8}$ s).
3.4 Discussion

3.4.1 Fitted Models

We attempted to fit HS α4β2 nAChR single channel data with 3 model types and their variants. The desensitization model types were just a subset of possible configurations that would be viable for describing the desensitization time course of the receptors. The classical mechanisms such as Katz and Thesleff and MWC have been successfully utilized in prior studies to describe nicotinic receptors (Edelstein et al., 1996; Colquhoun and Sakmann, 1985). In their simpler forms the primary classical mechanisms of Katz and Thesleff and MWC failed to describe the single-channel data that they were fitted to, even when the Katz and Thesleff contained a modification of an additional open state. Modifying the MWC to account for desensitization proved to be the most successful. We found that a model containing 3 desensitized states gave the best overall fits, error values, and simulated currents. Some combinations gave simulated currents that demonstrated extremely long (>20s) desensitization time courses similar to or longer than those seen in previous whole cell experiments on oocytes, but these models contained the largest amount of error (Weltzin and Schulte, 2015). In our data we could not justify adding more than 3 desensitized states in an activation mechanism due to the lack of complexity in the closed time distribution and the possibility of more than one channel in our data. The likelihood that we are not capturing any activation information in our patches was also investigated. This scenario is plausible given that 1 μM ACh is around the EC_{50} for these receptors and previous reports on nicotinic receptors have observed a lack of activation data at high concentrations (Elenes and Auerbach, 2002).
Several desensitization schemes were explored for all concentrations, but an additional open state was found to be needed to describe the open lifetime distribution. While the closed time distributions appear rather simple, they most likely contain several desensitized states that are difficult to resolve. This added with the instability at the end of the dFBr data adds additional complexity to result interpretation.

3.4.2 Scheme 2 Predictions

The MWC based mechanisms gave consistent activation rates across the two concentrations of 0.1 and 1 μM ACh. The MWC modified scheme 2 gave the most promising results in terms of errors, peak open probability and desensitization. We discounted any schemes that when simulated over long periods (>25s) gave high steady state currents and no visible desensitization. In order to test the effects dFBr would have on macroscopic currents we used models obtained from 1 μM ACh and the 1 μM ACh + 1 μM dFBr data. The primary finding was that dFBr potentiates peak currents of HS receptors across all concentrations. At higher concentrations (≥5 μM) potentiation is offset by an increase in desensitization, resulting in overall lower charge transfer, while at very low concentrations (≤ 3 μM) dFBr increases the desensitization time course, and charge transfer increases. Potentiation at high concentrations is brought about via an increase in the amount of diliganded openings, while increased desensitization occurs via an increase into the desensitized state D2. This increase directly affects the slow desensitization time course, while having little effect on the fast time component. At low concentrations, potentiation primarily occurs via an increase in
the mean open time of O1, as well as a decreased rate of entry into the desensitized state off of the diliganded closed state. While the concentration is not high enough for the change in desensitization to override potentiation, we do observe that receptors will be ‘pushed’ into the long lived desensitized states by the increased rate into the diliganded open state.

The potential implications our findings have for signaling will largely depend on where HS receptors are expressed and in how they are exposed to agonist. For signaling at classical fast synapses, dFBr could have significant consequences based upon the agonist concentration profile HS receptors are exposed to. At synapses where there are brief, high concentrations of ACh, dFBr will initially potentiate receptors before pushing them into desensitization. Based on our simulations this would result in an overall decrease in charge transfer. However, if the receptors are exposed to long lasting ambient (low) levels of ACh concentration, such as in a ‘volume transmission’ model (Ren et al., 2011; Descarries et al., 1997), then dFBr is capable of boosting receptor responsiveness and increasing overall charge transfer for HS receptors.

3.4.3 Model Discrepancies

All of our dFBr data contained instability that arose near the end of the patch. This instability is due to either multiple receptors dropping out as the patch ages or dFBr having a greater effect on the open channel duration which becomes weaker as the patch progresses. The instability could also be a product of both scenarios. Loss of receptors is the more likely culprit as more than one observable channel is seen in the
first half of the records. The 0.1 and 1 μM ACh data is very stable, and we have very high confidence in the results for those patches. Of all of our data for the 10 μM ACh patches consistently had only one observable channel, but were unstable near the end of the records and the initial $P_{\text{open}}$ obtained from stability plots is very low. It was noted that at 10 μM the open time distributions became shorter, and it is possible that block is involved even at concentrations this low. Another consistent discrepancy was in the 10 μM data all gave significantly lower activation rates compared to other concentrations. Whether this is due to open channel block on the single-channel level or another mechanism is currently unknown. Properties of potential block will have to be investigated further on a single-channel level with higher concentrations of agonist (25-100 μM ACh).

There were consistent discrepancies throughout all of the MWC models. The most notable of these was that the 0.1 μM ACh data gave higher $P_{\text{open}}$ simulated currents compared to the 1 μM ACh. This larger $P_{\text{open}}$ is due to the fact that the 0.1 μM data has a 20-fold larger rate of entry into the diliganded open state. In our 1 μM data we do begin to see short closings disappearing which would account for the discrepancy between the two concentrations. All of the 0.1 μM models had higher rates between the singly-ligated and diliganded open state, which did not appear to have a large effect on simulated currents. We were unable to determine the reason behind the 0.1 μM data having an increase in these rates. In future studies we will need to investigate lower concentrations of ACh (1-50 nM) with dFBr and see if we still see similar effects as we have observed with 1 μM data.
3.4.4 HEPES Effects on HS Single-Channel Data

In previous studies, HEPES appeared to be a PAM selective for the HS subtype compared to LS $\alpha4\beta2$ and $\alpha7$ nicotinic acetylcholine receptors (Weltzin et al., 2014). HEPES is thought to potentiate and inhibit by binding to the $\beta2+/\beta2-$ interface of the HS $\alpha4\beta2$ receptors. It was also found that, inhibition of HEPES potentiated responses occurs at concentrations greater than 100 mM HEPES. Due to the binding location of HEPES, the presence of the Good’s buffer could be altering our HS dFBr data.

In oocyte preparations, the presence of a HEPES buffer altered the efficacy of dFBr for the HS $\alpha4\beta2$ receptors. The lower efficacy of dFBr in the HS subtype may be the result of competition between HEPES and dFBr (Weltzin et al., 2014). In our preparations we utilized 5 mM of HEPES buffer, as is common in a lot of electrophysiology recordings. We believe that the presence of HEPES on HS receptors gives the appearance of increased desensitization in dFBr. This is due to HEPES binding at the $\beta2+/\beta2-$ interface, offsetting dFBr’s steady state potentiating effects by inhibiting the receptor. Therefore, it would be wise in future single molecule studies of HS $\alpha4\beta2$ receptors to not utilize HEPES as a buffer in order to eliminate any unnecessary alterations in the observed gating.

Overall, we have provided a larger picture of the kinetics on a single-channel level of the high sensitivity $\alpha4\beta2$ nAChR and give more insight into the potential effects of dFBr on these receptors. Our data supports dFBr exhibiting potentiation by increasing peak current at all concentrations, but an overall decrease in charge transfer is observed at higher concentrations potentially due to the presence of HEPES. The
effect this has on desensitization largely depends on the concentration profile of ACh, with increased desensitization existing at higher concentrations. Our model offers one possibility to what we observed on a single-channel level for the HS receptors, and provides a starting point for future analysis of dFBr analogues.
3.5 Acknowledgements

We would like to thank Dr. Joe Henry Steinbach for the stably transfected HEK293 cell line.
Figure 3.1: Unitary conductance’s of $\alpha_4\beta_2\delta_3$ receptors. (A) Open point amplitude distribution for high sensitivity (HS) conductance channels in the absence [19.5 pS ± 1.8 (mean±SD), black traces] and presence [19.9 pS ±0.34 (mean ±SD), blue traces] of dFBr with respective traces underneath. Calibration: 200 ms and 1.5 pA. (B) I-V plot shows mean value (± SD) for HS [19.9 ± 1.1 pS; n=7; black circles] recorded in 1 $\mu$M ACh and HS receptors recorded in 1 $\mu$M ACh + 1 $\mu$M dFBr [19.8 ± 0.34 pS; n=4; blue squares]. I/V data gave slopes of ~20 pS in the absence and presence of dFBr with reversal potentials of 6.3 and 1.2 mV respectively.
Figure 3.2: Steady state recordings of $\alpha_4\beta_2\gamma_3$ receptors. Steady state recordings of HS in 1 $\mu$M (top) and 1 $\mu$M ACh + 1 $\mu$M dFBr. Contiguous recordings of 50 s of data filtered at 1kHz (-3dB, 4-pole Bessel). Calibration: 100ms, 1.5 pA.
Figure 3.3: Burst characteristics of α4β2 receptors. See caption on next page
Figure 3.3: Burst characteristics of α4β2 receptors. Distributions of burst length (left) and open lifetime (right). Distributions are fitted with two exponentials (red line) and are plotted up to 20 ms for comparison. Varied concentrations (0.1 μM, 1 μM, and 10 μM) of ACh are compared [tcrit: 1.13 ms, 1.65 ms, and 2.8 ms respectively] to 1 μM ACh + 1 μM dFBr (tcrit: 0.57 ms). Bursts usually contained single opening at concentrations greater than 0.1 μM and are defined in methods. Bursts length time constants slightly diminishes with increasing concentration [0.1 μM: τ1= 1.2 ms, τ2= 5.9 ms; 1 μM: τ1= 3.12 ms, τ2= 3.4 ms; 10 μM: τ1= 0.96 ms, τ2= 1.23 ms] with the presence of dFBr increasing τs [τ1= 4.5 ms and τ2= 5.04 ms]. A similar effect is observed with open lifetime [0.1 μM: τ1= 1.1 ms, τ2= 5.8 ms; 1 μM: τ1= 2.9 ms, τ2= 3.7 ms; 10 μM: τ1= 0.9 ms, τ2= 1.3 ms] with dFBr also increasing apparent open lifetime [τ1=4.5 ms and τ2= 10.7 ms].
Figure 3.4: Cluster characteristics of $\alpha_4\beta_2$ receptors. See caption on next page.
Figure 3.4: Cluster characteristics of α4β2 receptors. Distributions of cluster length (left), open lifetime (middle), and openings per cluster (right). Distributions are fitted with two exponentials (red line) for 0.1 μM (t_{crit}: 31.5 ms), 1 μM (t_{crit}: 42.2 ms) and 10 μM ACh (t_{crit}: 20.4 ms) and 1 μM ACh + 1 μM dFBr (t_{crit}: 5.33 ms). The determination of critical time is described in methods. (A) Cluster length for 0.1-10 μM ACh and 1 μM ACh + 1 μM dFBr. Distributions are plotted up to 80 ms for comparison between them. Cluster length decreases with increasing ACh concentrations (0.1 μM: τ_1 = 0.297 ms, τ_2 = 4.01 ms; 1 μM: τ_1 = 0.49 ms, τ_2 = 2.43 ms; 10 μM: τ_1 = 0.37 ms, τ_2 = 1.83 ms). The presence of dFBr increased the long time constant for cluster length (τ_1 = 1.89 ms, τ_2 = 8.48 ms). (B) Distribution of open lifetime for 0.1 to 10 μM ACh and 1 μM ACh + 1 μM dFBr. Distributions are plotted up to 40 ms for comparison. Open lifetimes decrease with increasing concentration [0.1 μM: τ_1 = 0.78 ms, τ_2 = 2.1 ms; 1 μM: τ_1 = 0.41 ms, τ_2 = 1.9 ms; 10 μM: τ_1 = 0.33 ms, τ_2 = 1.6 ms]. (C) Histograms for openings per burst for 0.1-10 μM ACh and 1 μM ACh + 1 μM dFBr
Figure 3.5: Stability plots for $\alpha_4\beta_2$ receptors. See caption on next page.
Figure 3.5: Stability plots for $\alpha_4\beta_2$ receptors. Segments were plotted vs occupancy of the open state for 0.1-10 $\mu$M ACh and 1 $\mu$M ACh + 1 $\mu$M dFBr. 0.1 $\mu$M [top-left; 0.0067±0.014 (mean±SD)] and 1 $\mu$M ACh [top-right; 0.016±0.022 (mean±SD)], both contained one observable channel and were the most stable. 10 $\mu$M [bottom-left; 0.0011±0.0009 (mean±SD)] loses activity after ~500 segments and 1 $\mu$M ACh + 1 $\mu$M dFBr [bottom-right; 0.0502±0.04 (mean±SD)] loses activity after ~1200 segments, both contained more than one observable channel.
Figure 3.6: $\alpha_{4}\beta_{2}$ receptor mechanisms. Mechanisms fitted to 0.1-10 $\mu$M ACh and 1 $\mu$M ACh + 1 $\mu$M dFBr data. Scheme 1-3 are mechanisms that include activation including classical Katz and Theslaeff mechanisms. Scheme 3-7 are desensitization models based upon most common outcomes for QuB’s model search.
Figure 3.7: Katz and Thesleff model fits. Fits for the modified Katz and Thesleff shown in Figure 3.6 for ACh concentrations 0.1 -10 µM and 1 µM ACh + 1 µM dFBr (bottom).
Figure 3.8: MWC model fits. Fits for the MWC shown in Figure 3.6 for ACh concentrations 0.1 -10 μM and 1 μM ACh + 1 μM dFBr (bottom).
Figure 3.9: Scheme 1 model fits. Open and closed duration distributions from patches for 0.1 μM, 1 μM ACh and 1 μM ACh 1 μM dFBr (top, middle and bottom respectively) fitted with gating model scheme 1 shown in Figure 3.6. Scheme 1 is a modified version of the MWC model in Figure 3.4.
Figure 3.10: Scheme 2 model fits. Open and closed duration distributions from patches for 0.1 μM and 1 μM ACh (top, middle and bottom respectively) fitted with gating model scheme 2 shown in Figure 3.6. Scheme 2 is modified version of MWC models shown in Figure 3.4.
Figure 3.11A: Concentration dependence of peak and steady-state currents for scheme 2. (A) Simulated currents are shown for the ACh (black traces) and dFBr (blue traces) MWC based scheme 2 driven by 2.5 s jumps into 10 μM (top), 100 μM (middle), and 1 mM (bottom) ACh. Exponential functions were fitted to the decay phase of currents for all concentrations. For 1 mM ACh the exponential functions were ACh scheme 2: $A_1=0.37$, $\tau_1=17$ ms, $A_2=0.23$, $\tau_2=794$ ms, $y_0=0.008$. dFBr scheme 2: $A_1=0.28$, $\tau_1=0.006$ ms, $A_2=0.32$, $\tau_5=0.006$ ms, $y_0=0.05$. Rise times (10 – 90%) for ACh and dFBr models with 1 mM ACh were 5.4 ms and 2.1 ms, respectively. Calibration bar: units of open probability ($P_{\text{open}}$).
Figure 3.11B: Concentration dependence of peak and steady-state currents for scheme 2. (B) Open probability ($P_{\text{open}}$) of peak (squares) and steady-state (stars) responses are plotted for ACh (black) and dFBr (blue) scheme 2 models as a function of ACh concentration (1, 5, 10, 50, 100, 200, 300, 500, and 1000 µM). The $P_{\text{open}}$ at the peak for saturating ACh concentration was ~ 0.42 for ACh and 0.63 for dFBr.
Figure 3.11C: Concentration dependence of peak and steady-state currents for scheme 2. (C) The ratio of dFBr to ACh scheme 2 $P_{\text{open}}$ is shown as a function of ACh concentration. Potentiation of peak response (stars) was largest at very low ACh (600% increase for 1 µM ACh) and declined to a steady value of 150% increase at 50 µM ACh. Steady-state, desensitized responses (blue star) were diminished ~25% at concentrations above 1 µM ACh.
Figure 3.12A: Potentiation of peak and steady-state currents at low concentrations of ACh. (A) Simulated currents at low concentrations are shown for the ACh (black traces) and dFBr (blue traces) scheme 2 models driven by 15 s jumps into 1 μM (top), 3 μM (middle), and 5 μM (bottom) ACh. Calibration bar: units of open probability ($P_{open}$).
Figure 3.12B: Potentiation of peak and steady-state currents at low concentrations of ACh. (B) The fast time ($\tau_f$) component are plotted for ACh (black) and dFBr (blue) scheme 2 models as a function of ACh concentration (1, 3, 5, 10, 50, 100, 200, 300, 500, 1000 $\mu$M). dFBr only showed slower $\tau_f$ at 1 $\mu$M ACh, higher concentrations yielded faster decay time courses.
Figure 3.12C: Potentiation of peak and steady-state currents at low concentrations of ACh. (C) The slow time ($\tau_s$) component are plotted for ACh and dFBr scheme 2 models as a function of ACh concentration. Slower time component at concentrations higher than 10 $\mu$M, almost completely disappears in the presence of dFBr.
Figure 3.13: Occupancy probabilities of α2β2 receptors using a concentration jump profile. Left: Occupancy probabilities (y-axis) of channel states after a concentration jump are shown for 10, 100 and 1000 μM ACh (top to bottom) using either the scheme 2 for ACh (solid lines) or scheme 2 for dFBr (dashed lines). Right: enlarged portion (grey squares, y-axis: 0.5 and x-axis: 0.15).
Figure 3.14A: Response to a pulse of a high concentration of agonist. Occupancy probabilities are shown for models driven by an alpha function to mimic a synaptic agonist concentration profile. (A) Responses shown for the ACh model. Peak open probability ($P_{o2}$) was 0.352 at 3.4 ms compared with 0.634 at 2.1 ms for dFBr (shown in B; 1.8-fold potentiation). The decay of $P_{o2}$ was fitted with one exponential ($\tau$=12.9 ms). $P_{d2}$ was fitted with two exponential components ($A_1$=0.60, $\tau_1$=526 ms, $A_2$=0.40, $\tau_2$=633 ms, $y_0$=0), and $P_{d3}$ was fitted with one components ($\tau$=6.7 s, $y_0$=0).
Figure 3.14B: Response to a pulse of a high concentration of agonist. (B) For the dFBr model the decay of $P_{02}$ was also fitted with one exponential component ($\tau=5$ ms). $P_{D2}$ was fitted with two components ($A_1=0.54$, $\tau_1=18.1$ s, $A_2=0.46$, $\tau_2=15.2$ s, $y_0=0$), and $P_{D3}$ was fitted with a single exponential ($\tau=21.8$ s, $y_0=0$).
Table 3.1: Rate constants for Katz and Thesleff mechanism

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Table 3.2: Rate constants for MWC mechanism.

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<th>10 μM</th>
<th>dFBr</th>
</tr>
</thead>
<tbody>
<tr>
<td>$2k_{+1}$</td>
<td>M$^{-1}$s$^{-1}$</td>
<td>4.9E7 ± 10%</td>
<td>2.4E7 ± 24%</td>
<td>4E4 ± 100%</td>
<td>4E5 ± 300%</td>
</tr>
<tr>
<td>$k_{+1}$</td>
<td>M$^{-1}$s$^{-1}$</td>
<td>2.4E7 ± 10%</td>
<td>1.2E7 ± 24%</td>
<td>2E4 ± 100%</td>
<td>2E5 ± 300%</td>
</tr>
<tr>
<td>$2k_{-1}$</td>
<td>s$^{-1}$</td>
<td>1080 ± 16%</td>
<td>931 ± 20%</td>
<td>0.85 ± 21%</td>
<td>1.7 ± 12%</td>
</tr>
<tr>
<td>$k_{-1}$</td>
<td>s$^{-1}$</td>
<td>2167 ± 16%</td>
<td>466 ± 20%</td>
<td>0.42 ± 21%</td>
<td>0.85 ± 12%</td>
</tr>
<tr>
<td>$\beta_1$</td>
<td>s$^{-1}$</td>
<td>362 ± 24%</td>
<td>405 ± 24%</td>
<td>1.2 ± 100%</td>
<td>5.5 ± 100%</td>
</tr>
<tr>
<td>$\alpha_1$</td>
<td>s$^{-1}$</td>
<td>818 ± 10%</td>
<td>28 ± 33%</td>
<td>1385 ± 100%</td>
<td>246 ± 100%</td>
</tr>
<tr>
<td>$\beta_2$</td>
<td>s$^{-1}$</td>
<td>15 ± 21%</td>
<td>303 ± 49%</td>
<td>1.4 ± 100%</td>
<td>4E-5 ± 100%</td>
</tr>
<tr>
<td>$\alpha_1$</td>
<td>s$^{-1}$</td>
<td>0.1 ± 24%</td>
<td>85 ± 77%</td>
<td>329 ± 100%</td>
<td>6E-4 ± 100%</td>
</tr>
<tr>
<td>$k_{+2}$</td>
<td>M$^{-1}$s$^{-1}$</td>
<td>2.2E9 ± 17%</td>
<td>580 ± 100%</td>
<td>92 ± 100%</td>
<td>3.8E7 ± 100%</td>
</tr>
<tr>
<td>$k_{-2}$</td>
<td>s$^{-1}$</td>
<td>103 ± 26%</td>
<td>0.001 ± 100%</td>
<td>0.001 ± 100%</td>
<td>103 ± 100%</td>
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Table 3.3: Rate constants for scheme 1 fits.

<table>
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<th>1 µM</th>
<th>+dFBr</th>
</tr>
</thead>
<tbody>
<tr>
<td>2k⁺₁</td>
<td>M⁻¹s⁻¹</td>
<td>7.6E7 ± 98%</td>
<td>2.32E7 ± 26%</td>
<td>3.8E7 ± 100%</td>
</tr>
<tr>
<td>k⁺₁</td>
<td>M⁻¹s⁻¹</td>
<td>3.8E7 ± 98%</td>
<td>1.16E7 ± 26%</td>
<td>1.9E7 ± 100%</td>
</tr>
<tr>
<td>2k⁻₁</td>
<td>s⁻¹</td>
<td>1712 ± 35%</td>
<td>2455 ± 12%</td>
<td>1472 ± 8%</td>
</tr>
<tr>
<td>k⁻₁</td>
<td>s⁻¹</td>
<td>856 ± 35%</td>
<td>1227 ± 12%</td>
<td>736 ± 8%</td>
</tr>
<tr>
<td>β₁</td>
<td>s⁻¹</td>
<td>396 ± 59%</td>
<td>114 ± 32%</td>
<td>137 ± 18%</td>
</tr>
<tr>
<td>α₁</td>
<td>s⁻¹</td>
<td>287 ± 28%</td>
<td>671 ± 30%</td>
<td>175 ± 18%</td>
</tr>
<tr>
<td>β₂</td>
<td>s⁻¹</td>
<td>37 ± 22%</td>
<td>1543 ± 59%</td>
<td>519 ± 100%</td>
</tr>
<tr>
<td>α₂</td>
<td>s⁻¹</td>
<td>0.1 ± 100%</td>
<td>29 ± 98%</td>
<td>22.4 ± 100%</td>
</tr>
<tr>
<td>k⁺₂</td>
<td>M⁻¹s⁻¹</td>
<td>9.1E8 ± 100%</td>
<td>1.1E6 ± 100%</td>
<td>0.0048 ± 0.01%</td>
</tr>
<tr>
<td>k⁻₂</td>
<td>s⁻¹</td>
<td>147 ± 32%</td>
<td>0.76 ± 100%</td>
<td>1.3E⁻⁸ ± 100%</td>
</tr>
<tr>
<td>d⁺¹</td>
<td>s⁻¹</td>
<td>307 ± 60%</td>
<td>54 ± 11%</td>
<td>0.18 ± 53%</td>
</tr>
<tr>
<td>d⁻¹</td>
<td>s⁻¹</td>
<td>0.49 ± 70%</td>
<td>0.14 ± 11%</td>
<td>8.6E⁻⁴ ± 11%</td>
</tr>
<tr>
<td>d⁺²</td>
<td>s⁻¹</td>
<td>33 ± 35%</td>
<td>99 ± 5%</td>
<td>99 ± 3.3%</td>
</tr>
<tr>
<td>d⁻²</td>
<td>s⁻¹</td>
<td>361 ± 58%</td>
<td>5.3 ± 8%</td>
<td>0.7 ± 3.7%</td>
</tr>
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</table>
Table 3.4: Rate constants for scheme 2 fits.

![Chemical reaction scheme](image)

Table 4.4

<table>
<thead>
<tr>
<th>Rate</th>
<th>Unit</th>
<th>0.1 (\mu)M</th>
<th>1 (\mu)M</th>
<th>+dFBr</th>
</tr>
</thead>
<tbody>
<tr>
<td>2(k_+1)</td>
<td>(M^{-1}s^{-1})</td>
<td>9E7 ± 13%</td>
<td>2.8E7 ± 32%</td>
<td>3.5E7 ± 94%</td>
</tr>
<tr>
<td>(k_+1)</td>
<td>(M^{-1}s^{-1})</td>
<td>4.5E7 ± 13%</td>
<td>1.4E7 ± 33%</td>
<td>1.7E7 ± 94%</td>
</tr>
<tr>
<td>2(k_-1)</td>
<td>(s^{-1})</td>
<td>794 ± 12%</td>
<td>1261 ± 20%</td>
<td>1678 ± 7.8%</td>
</tr>
<tr>
<td>(k_-1)</td>
<td>(s^{-1})</td>
<td>397 ± 12%</td>
<td>631 ± 20%</td>
<td>839 ± 7.8%</td>
</tr>
<tr>
<td>(\beta_1)</td>
<td>(s^{-1})</td>
<td>82 ± 52%</td>
<td>52 ± 42%</td>
<td>178 ± 95%</td>
</tr>
<tr>
<td>(\alpha_1)</td>
<td>(s^{-1})</td>
<td>1090 ± 8.5%</td>
<td>417 ± 35%</td>
<td>181 ± 90%</td>
</tr>
<tr>
<td>(\beta_2)</td>
<td>(s^{-1})</td>
<td>6370 ± 18.6%</td>
<td>323 ± 71%</td>
<td>1000 ± 24%</td>
</tr>
<tr>
<td>(\alpha_2)</td>
<td>(s^{-1})</td>
<td>300 ± 17%</td>
<td>40 ± 95%</td>
<td>27 ± 98%</td>
</tr>
<tr>
<td>(k_+2)</td>
<td>(M^{-1}s^{-1})</td>
<td>3.3E9 ± 25%</td>
<td>2E3 ± 99%</td>
<td>2.7 ± 74%</td>
</tr>
<tr>
<td>(k_-2)</td>
<td>(s^{-1})</td>
<td>214 ± 18%</td>
<td>0.003 ± 99%</td>
<td>7.2E-6 ± 99%</td>
</tr>
<tr>
<td>(d_+1)</td>
<td>(s^{-1})</td>
<td>3 ± 99%</td>
<td>2.7E-4 ± 99%</td>
<td>0.39 ± 56%</td>
</tr>
<tr>
<td>(d_-1)</td>
<td>(s^{-1})</td>
<td>0.35 ± 98%</td>
<td>0.023 ± 99%</td>
<td>3.7E-5 ± 99%</td>
</tr>
<tr>
<td>(d_+2)</td>
<td>(s^{-1})</td>
<td>57 ± 29%</td>
<td>44 ± 23%</td>
<td>195 ± 3.5%</td>
</tr>
<tr>
<td>(d_-2)</td>
<td>(s^{-1})</td>
<td>0.33 ± 18%</td>
<td>4.3 ± 35%</td>
<td>0.77 ± 4%</td>
</tr>
<tr>
<td>(d_+3)</td>
<td>(s^{-1})</td>
<td>975 ± 25%</td>
<td>236 ± 55%</td>
<td>63 ± 55%</td>
</tr>
<tr>
<td>(d_-3)</td>
<td>(s^{-1})</td>
<td>43 ± 50%</td>
<td>0.19 ± 31%</td>
<td>0.08 ± 41%</td>
</tr>
</tbody>
</table>
3.6 References


Figure A.3.1: Model based off LS receptor mechanism. Open and closed duration distributions from patches for 0.1 μM, 1 μM and 1 μM ACh + 1 μM dFBr (top, middle and bottom respectively) group fitted with the LS gating model shown in Figure 2.6.
Figure A.3.2: Scheme 3 model fits. Open and closed duration distributions from patches for 0.1 μM, 1 μM and 10 μM ACh (top, middle and bottom respectively) group fitted with gating model scheme 3 shown in Figure 3.6. Scheme 3 contains no activation steps.
Figure A.3.3: Scheme 4 model fits. Open and closed duration distributions from patches for 0.1 μM, 1 μM and 10 μM ACh (top, middle and bottom respectively) group fitted with gating model scheme 4 shown in Figure 3.6.
Figure A.3.4: Scheme 5 model fits. Open and closed duration distributions from patches for 0.1 μM, 1 μM and 10 μM ACh (top, middle and bottom respectively) group fitted with gating model scheme 5 (star) shown in Figure 3.6.
Figure A.3.5: Scheme 6 model fits. Open and closed duration distributions from patches for 0.1 μM, 1 μM and 10 μM ACh (top, middle and bottom respectively) group fitted with gating model scheme 6 (linear) shown in Figure 3.6.
Figure A.3.6: Scheme 7 model fits. Open and closed duration distributions from patches for 0.1 μM, 1 μM and 10 μM ACh (top, middle and bottom respectively) group fitted with gating model scheme 7 shown in Figure 3.6.
Table A.3.1 LS based mechanism rate constants.

<table>
<thead>
<tr>
<th>Rate</th>
<th>Unit</th>
<th>0.1 μM</th>
<th>1 μM</th>
<th>+dFBr</th>
</tr>
</thead>
<tbody>
<tr>
<td>2k⁺₁</td>
<td>M⁻¹s⁻¹</td>
<td>1.3E8 ± 10%</td>
<td>2.32E7 ± 26%</td>
<td>3.8E7 ± 100%</td>
</tr>
<tr>
<td>k⁺₁</td>
<td>M⁻¹s⁻¹</td>
<td>6.7E7 ± 10%</td>
<td>1.16E7 ± 26%</td>
<td>1.9E7 ± 100%</td>
</tr>
<tr>
<td>2k⁻¹</td>
<td>s⁻¹</td>
<td>1571 ± 8%</td>
<td>2455 ± 12%</td>
<td>1563 ± 8%</td>
</tr>
<tr>
<td>k⁻¹</td>
<td>s⁻¹</td>
<td>785 ± 8%</td>
<td>1227 ± 12%</td>
<td>781 ± 8%</td>
</tr>
<tr>
<td>β₁</td>
<td>s⁻¹</td>
<td>106 ± 9%</td>
<td>114 ± 32%</td>
<td>139 ± 99%</td>
</tr>
<tr>
<td>α₁</td>
<td>s⁻¹</td>
<td>539 ± 5%</td>
<td>671 ± 30%</td>
<td>185 ± 99%</td>
</tr>
<tr>
<td>β₂</td>
<td>s⁻¹</td>
<td>4100 ± 14%</td>
<td>1543 ± 59%</td>
<td>221 ± 100%</td>
</tr>
<tr>
<td>α₂</td>
<td>s⁻¹</td>
<td>101 ± 13%</td>
<td>29 ± 98%</td>
<td>16 ± 100%</td>
</tr>
<tr>
<td>k⁺₂</td>
<td>M⁻¹s⁻¹</td>
<td>7.1E8 ± 12%</td>
<td>1.1E6 ± 100%</td>
<td>7.6E-7 ± 100%</td>
</tr>
<tr>
<td>k⁻₂</td>
<td>s⁻¹</td>
<td>82 ± 9%</td>
<td>0.76 ± 100%</td>
<td>3.5E-6 ± 100%</td>
</tr>
<tr>
<td>d⁺₁</td>
<td>s⁻¹</td>
<td>2863 ± 18%</td>
<td>51 ± 11%</td>
<td>993 ± 33%</td>
</tr>
<tr>
<td>d⁻₁</td>
<td>s⁻¹</td>
<td>0.3 ± 10%</td>
<td>0.2 ± 11%</td>
<td>2.5 ± 23%</td>
</tr>
<tr>
<td>d⁺₂</td>
<td>s⁻¹</td>
<td>2E-7 ± 100%</td>
<td>0.5 ± 5%</td>
<td>196 ± 4%</td>
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<tr>
<td>d⁻₂</td>
<td>s⁻¹</td>
<td>6E-9 ± 100%</td>
<td>2.6E-10 ± 8%</td>
<td>0.33 ± 4%</td>
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Table A.3.2: Rate constants for scheme 3 and scheme 4.

<table>
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<tr>
<th>Rate constants</th>
<th>Scheme 3</th>
<th>+dFBbr</th>
<th>Scheme 4</th>
<th>+dFBbr</th>
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</thead>
<tbody>
<tr>
<td>k₀₁</td>
<td>67 ± 14%</td>
<td>122 ± 3.5%</td>
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<td>17 ± 7%</td>
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<tr>
<td>k₁₀</td>
<td>6580 ± 13%</td>
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<td>6457 ± 11%</td>
<td>961 ± 11%</td>
</tr>
<tr>
<td>k₀₂</td>
<td>193 ± 57%</td>
<td>47 ± 9%</td>
<td>345 ± 2.4%</td>
<td>46.2 ± 8.9%</td>
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<tr>
<td>k₂₀</td>
<td>2.4 ± 27%</td>
<td>0.98 ± 6%</td>
<td>18 ± 16%</td>
<td>7.5 ± 7.7%</td>
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<tr>
<td>k₂₃</td>
<td>0.4 ± 156%</td>
<td>0.002 ± 80%</td>
<td>147 ± 21%</td>
<td>0.002 ± 90%</td>
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<tr>
<td>k₃₂</td>
<td>0.3 ± 42%</td>
<td>43500 ± 100%</td>
<td>15 ± 11%</td>
<td>0.002 ± 90%</td>
</tr>
<tr>
<td>k₀₄</td>
<td>185 ± 59%</td>
<td>22.2 ± 8%</td>
<td>32.4 ± 13%</td>
<td>117 ± 3.7%</td>
</tr>
<tr>
<td>k₄₀</td>
<td>41 ± 52%</td>
<td>1308 ± 12%</td>
<td>0.1 ± 10%</td>
<td>6.3 ± 5.5%</td>
</tr>
<tr>
<td>k₄₅</td>
<td>208 ± 30%</td>
<td>6.8 ± 57%</td>
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<tr>
<td>k₅₄</td>
<td>6 ± 116%</td>
<td>0.002 ± 600%</td>
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</tr>
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</table>
Table A.3.3: Rate constants for scheme 5.

\[
\begin{align*}
C & \xrightarrow{k_{a1}} C \\
C & \xrightarrow{k_{c1}} C \\
C & \xrightarrow{k_{c2}} C \\
C & \xrightarrow{k_{c3}} C \\
C & \xrightarrow{k_{c4}} C
\end{align*}
\]

Table A.3.3

<table>
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<tr>
<th>Rate constants</th>
<th>Scheme 5</th>
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<tr>
<td>k1(O)</td>
<td>6955 ± 12%</td>
<td>13.8 ± 38%</td>
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<tr>
<td>k0(1)</td>
<td>67.5 ± 13%</td>
<td>45 ± 65%</td>
</tr>
<tr>
<td>k2(O)</td>
<td>279 ± 21%</td>
<td>0.002 ± 90%</td>
</tr>
<tr>
<td>k0(2)</td>
<td>28 ± 10%</td>
<td>0.11 ± 68%</td>
</tr>
<tr>
<td>k3(O)</td>
<td>3.5 ± 13%</td>
<td>1316 ± 13%</td>
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<tr>
<td>k0(3)</td>
<td>117 ± 11%</td>
<td>20 ± 8%</td>
</tr>
<tr>
<td>k4(O)</td>
<td>0.06 ± 13%</td>
<td>5 ± 21%</td>
</tr>
<tr>
<td>k0(4)</td>
<td>17 ± 17%</td>
<td>81 ± 35%</td>
</tr>
<tr>
<td>k5(O)</td>
<td>1 ± 0.01%</td>
<td>1 ± 0.01%</td>
</tr>
<tr>
<td>k0(5)</td>
<td>217 ± 6%</td>
<td>40 ± 6%</td>
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Table A.3.4: Rate constants for scheme 6 and scheme 7.

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<th>Scheme 7</th>
<th>+dFBr</th>
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<tbody>
<tr>
<td>k1O</td>
<td>1001 ± 690%</td>
<td>143 ± 15%</td>
<td>6428 ± 13%</td>
<td>1429 ± 13%</td>
</tr>
<tr>
<td>kO1</td>
<td>444 ± 610%</td>
<td>186 ± 2%</td>
<td>65 ± 13%</td>
<td>22 ± 9%</td>
</tr>
<tr>
<td>k12</td>
<td>5496 ± 861%</td>
<td>1088 ± 13%</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>k21</td>
<td>114 ± 473%</td>
<td>54 ± 9%</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>k23</td>
<td>173 ± 492%</td>
<td>2.5 ± 22%</td>
<td>7.9 ± 39%</td>
<td>2.2 ± 65%</td>
</tr>
<tr>
<td>k32</td>
<td>18 ± 292%</td>
<td>3 ± 26%</td>
<td>0.007 ± 42%</td>
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<td>k34</td>
<td>0.23 ± 148%</td>
<td>0.43 ± 61%</td>
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<td>k43</td>
<td>0.34 ± 109%</td>
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<td>k45</td>
<td>5.7e-12 ± 2600%</td>
<td>0.008 ± 0.06%</td>
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<td>k54</td>
<td>1.8e-9 ± 2056%</td>
<td>0.002 ± 0.01%</td>
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<td>kO2</td>
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<td>31 ± 10%</td>
<td>127 ± 3%</td>
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<td>k2O</td>
<td>-</td>
<td>-</td>
<td>208 ± 20%</td>
<td>7.5 ± 8%</td>
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<tr>
<td>kO4</td>
<td>-</td>
<td>-</td>
<td>185 ± 59%</td>
<td>1 ± 0.001%</td>
</tr>
<tr>
<td>k4O</td>
<td>-</td>
<td>-</td>
<td>1.95 ± 7%</td>
<td>0.008 ± 50%</td>
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<td>kO5</td>
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<td>k5O</td>
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Chapter 4: Conclusions and Future Directions

4.1 General Overview

We have used single molecule methods to investigate the action of a positive allosteric modulator, dFBr. We have provided the first step in developing a model that describes the gating of the human α4β2 nicotinic receptors. Full gating models that give a fairly accurate representation of function allows naïve and modulated receptors to be simulated under a range of agonist profiles and concentrations. By determining the mechanism of modulation for HS and LS α4β2 receptors on a single-channel level, we have greatly enhanced the understanding of the potential therapeutic benefit dFBr has on α4β2 receptors at cholinergic synapses. How dFBr modulates would be determined by the agonist concentration profile at a synapse. The concentration differences would be dependent upon the spatial differences between receptors and local neurotransmitter release sites.

4.1.1 dFBr Destabilizes Desensitized States of α4β2 Receptors

Models were constructed for the LS subtype of the α4β2 nAChR. It was found that a 7 state model based off of an MWC mechanism described our data well and yielded unique rate constants for naïve and modulated receptors. The model contained two open states; a moniliganded open state off of C2 (O1) and a diliganded open state off of C3 (O2), as well as 5 closed states: two ligand binding steps (C1- C2 and C2 to
C3) and two desensitized states, one distal to C3 (D1) and the other distal to O2 (D2). For the LS subtype ($\alpha 4\beta 2$) nAChRs the primary finding was that dFBr destabilized the two desensitized states D1 and D2 by decreasing their mean lifetimes. The lifetime of D1 was decreased from 77 ms to 7.6 ms, and D2 decreased from 1.7 s to 490 ms.

Macroscopic simulations provided insight into the contribution destabilization of desensitized states has on modulation. State occupancy probabilities were investigated by driving the model with various profiles of ACh concentration. Potentiation at peak current exhibited by dFBr was primarily due to the destabilization of D1, which slowed the initial fast rate of macroscopic desensitization. Whereas, potentiation of steady state currents arose from a lower probability of occupying D2. The simulated dose response curves did not have a biphasic quality to them as seen in previous reports for LS receptors. The biphasic quality seen in LS receptors is due to the unique $\alpha 4-\alpha 4$ binding site, which allows for the binding of one additional agonist. Unlike the LS which contains two unique binding sites, the HS variant of the receptor only contains one type of agonist binding site, and therefore does not display a biphasic dose response curve (Harpsøe et al., 2011). Our model does not contain 3 activation steps, but still has an EC$_{50}$ (74 $\mu$M) close to the reported values for LS receptors (75 $\mu$M) (Nelson et al., 2003). While an additional activation step was tested, we found that the presence of a 3rd activation step did not give an accurate representation of the data. It may be that our concentration of ACh is too low to observe activation of the low sensitivity component of the receptor. However, this does not account for why our simulated dose response curves give EC$_{50}$ values similar to reported values.
We investigated previous reports that dFBr increases peak current 3-fold under high agonist concentrations. Our models using single-channel data predicted less than 1.1-fold potentiation at EC$_{75}$ (300 μM), compared to the 3-fold potentiation observed in previous studies (Weltzin and Schulte, 2010). When occupancy probabilities were calculated, we found that a slow rise in ACh concentration yields a lower peak $P_{\text{open}}$ for naive receptors and a larger increase in response when dFBr was applied. Our results do not support a large increase in peak current when high ACh concentrations are applied as seen in oocyte experiments. This discrepancy is likely in part due to slow solution exchange, which allows for desensitization to develop resulting in a lower peak amplitude in oocyte experiments.

The effects that dFBr would have on α4β2 receptors would largely be dependent on the agonist concentration profile at the synapse. Under simulated agonist concentration profiles similar to that at fast synapses, dFBr increased the open response amplitude. If the nicotinic receptors at fast cholinergic synapses become saturated with agonist, then dFBr’s ability to increase peak amplitude responses would be relatively ineffective at increasing postsynaptic signals. For LS receptors dFBr may be more suited for stimuli arising from a volume transmission scenario, where increasing the recovery from desensitization would have the greatest impact.
4.1.2 Synchronization of α4β2nAChR Receptors

For the HS α4β2 nAChR mechanisms were constructed and their viability as an accurate representation of channel gating was determined. It was found that an 8 state mechanism, based off of a simple MWC mechanism, best described our experimental single-channel HS data. The mechanism was composed of two activation steps (C1->C2 and C2->C3), two open states: a monoligated open state off of C2 (O1) and a diliganded open state off of C3 (O2). The open states contained transitions between one another. The HS model that best described our data and contained the best overall fits, error values and simulated currents contained 3 desensitized states. One distal to C3, with an additional two desensitized states off of O2.

For the HS nAChr subtype the primary finding was that dFBr synchronizes receptor openings at high concentrations by increasing the opening rate. dFBr potentiates peak currents across all concentrations, but charge transfer is reduced at simulated macroscopic responses with a concentration greater than or equal to 5 µM. The decrease in charge transfer at concentrations greater than 5 µM is at due to an increase in desensitization. For concentrations lower than 3 µM, the presence of dFBr resulted in an increase in charge transfer, reflective of a decrease in the likelihood of entering the desensitized state D2. The potentiation observed at low concentrations is brought on by the increase in the mean open lifetime of O1. While at higher concentrations, the increase into O2 has a larger effect on potentiation and the receptors desensitize more quickly.
With regards to signaling, the effects dFBr would have is dependent upon the synapse and the concentration of ACh in the synaptic cleft. dFBr has the ability to potentiate peak amplitude responses of HS receptors at low and high concentrations, but an increase in charge transfer is only observed at low concentrations. When in the presence of dFBr an overall decrease in charge transfer occurs, and postsynaptic signals generated from low frequency inputs may not benefit even though an increase in peak amplitude is observed over a wide range of agonist concentrations. However, dFBr would be the most beneficial if the receptors are exposed to low concentrations of agonist.

4.1.3 Implications for Nicotine Addiction

The pleasurable effects of nicotine occur due to the desensitization of high affinity nAChRs in the VTA. One of the most prominent types, α4β2 receptors, is mostly expressed on GABAergic neurons and dopaminergic inputs. These receptors desensitize rapidly when they are exposed to nicotine with the reinforcing nature of nicotine due to upregulation of α4 containing nAChRs on VTA GABAergic neurons (Nashmi et al., 2007). Receptor upregulation of α4β2 is directly related to desensitization, and receptors become hypersensitive after chronic exposure to nicotine leading to increased inhibition of dopaminergic activity (Hilario et al., 2012; Ngolab et al., 2015).

Our findings demonstrate that dFBr has the potential to be useful for the treatment of nicotine addiction. It is well known that the α4β2 receptors are upregulated
when chronically exposed to nicotine and there is some evidence that the high sensitivity form may be most affected. The effect that dFBr would have on HS receptors is unclear in our data as we observe increased desensitization due to the presence of channel block or HEPES, but we find that dFBr would increase receptors responsiveness when exposed to ambient concentrations of ACh. In the VTA α4β2 receptors are largely expressed on presynaptic inputs into dopaminergic neurons, and limiting the receptors ability to desensitize would be beneficial in reducing upregulation of the receptors. For LS receptors, the application of dFBr could lift receptors out of desensitization by destabilizing of desensitized states. This would give increased activity from inputs into dopaminergic neurons and could potentially lead to downregulation of α4β2 receptors back to normal levels, since receptor desensitization and upregulation have been shown to be linked (Fenster et al., 1999). The increase in activity within the VTA without the presence of nicotine would aid in eliminating symptoms of withdrawal that accompanies smoking cessation, which is one of the factors that prevents success in quitting (West et al., 1989).

4.1.4 Implications for the Treatment of Alzheimer’s Disease

One possible therapeutic approach for the treatment of Alzheimer’s disease is through the use of compounds that interfere with nAChR-Aβ interactions. Previous studies have looked at the interactions of dFBr and Aβ on α4β2 receptors and found that dFBr dissociates Aβ from the receptors. In our studies we found that dFBr increases responses for both α4β2 subtypes, making it a potential candidate for
Alzheimer’s disease if used in conjunction with other drugs. dFBr may be best paired with compounds that demonstrate the ability to dissociate Aβ from α7 nAChRs, antibodies that target Aβ, or both of these options. Compounds such as 2-[2- (4-bromophenyl)-2-oxoethyl]-1-methyl pyridinium (S 24795), a partial α7 nAChR agonist, have been found to dissociate Aβ in a concentration dependent manner (Wang et al., 2009, 2010). Combining dFBr with compounds like S-24795 will combat the negative effects Aβ binding has on α4β2 and α7 nAChRs, thereby potentially preventing further progression of the disease. The use of immunotherapy is still in its infancy (Goure et al., 2014), but the presence of compounds such as S-24795 and dFBr coupled with immunotherapy could potentially drastically reduce the concentration of Aβ in the brain. dFBr’s ability to increase responses by destabilizing desensitized states for LS receptors and increasing open time for HS receptors could help enhance cognitive ability, while simultaneously dissociating Aβ from receptors. Combating Alzheimer’s disease would require a therapy that combats the disease from many angles potentially utilizing multiple compounds such as dFBr.

4.2 Future Work

The investigation of dFBr on a single molecule level focused on characterizing the mechanisms of modulation for the α4β2 LS and HS nicotinic receptors. The goal was to provide a comprehensive gating model for both stoichiometric forms of the α4β2 receptors and determine the mechanism of action for the PAM dFBr. Determining how allosteric modulators would behave under different forms of synaptic transmission is
crucial to understanding the effect potentiation has on cholinergic synapses. Further investigation is needed into the effect structural changes have on receptors and allosteric modulators, and how these changes would change the mechanism of modulation. Understanding the structure-function relationship between dFBr analogs and nACHRs will help determine whether these changes increase dFBr’s viability as a therapeutic agent, aiding in rational drug design.

4.2.1 Effect of Linked Subunits on α4β2 Gating Models

One of the primary difficulties of doing whole cell and single-channel studies on receptors that contain more than one subtype, is receptor heterogeneity. While single-channel helps circumvent heterogeneity, more than one subtype can still be present in a patch adding complexity to data analysis. Many attempts have been made to address this problem and express α4β2 (LS) and α2β2 (HS) receptors separately. Strategies include using a 1:4 or 4:1 (or larger) ratio of α4:β2 subunit cDNAs for injecting or transfecting into the appropriate cell line. This technique gives monophasic concentration profiles, which suggests that there are homologous population of either α4β2 or α2β2 subtype (Moroni et al., 2006; Nelson et al., 2003), but it is currently unknown if this technique abolishes the unwanted stoichiometric form of the receptor. Alternatively, constructs with two units attached via synthetic linkers can be used to constrain the stoichiometry of α4β2 receptors (Zhou et al., 2003). These studies would be done on two linker configurations α4-6-β2 and β2-6-α4 in order to determine if the position of the linker has any noticeable effect on the desensitization or bursting characteristics of the receptor. With these studies the linker would have to be added to
the transfection medium in a 2:1 ratio of α4-6-β2/β2-6-α4 to α4/β2 cDNAs, depending on the subtype being studied. Linkers also have the advantage of being able to be combined with mutated subunits, allowing for greater insight into the effects of mutations on gating. Conducting single molecule studies on receptors that contain mutations would increase understanding of the structure function relationship for the receptors, and determine if mutations affect activation, burst, and desensitization kinetics. Preliminary work for this project involved introducing the linked subunit cDNA into a mammalian vector. Investigation into whether the naive receptors kinetics differ from that of linked receptors would also be beneficial for future studies utilizing linked subunits.

4.2.2 Investigation of dFBr Analogues

The compound being studied, dFBr, has potential drawbacks in its use as an effective therapeutic agent. Being a natural product, dFBr could have poor transit across the blood brain barrier, selectivity, and efficacy. A nicotine self-administration study conducted with rats administered with dFBr demonstrated that dFBr is present in cerebral fluid. This finding indicates that dFBr crosses the blood-brain barrier, but is available only at one-third the concentration compared to blood. There is room for improvement with dFBr’s qualities, including the compounds ability to transit across the blood brain barrier (Liu, 2013). In order for dFBr to be considered an effective therapeutic treatment dFBr needs to be improved and made more selective in its therapeutic target. To accomplish this, analogues can be tested to determine what
structural aspects of dFBr are important for modulation. Single-channel analysis would help in determining if a change in structure affected dFBr's mechanism of modulation and therefore which components are most important. These studies would be done using five dFBr analogues with EC\textsubscript{50}s ranging from 1 to 6.3 \mu{M} and IC\textsubscript{50}s from 8 to 90 \mu{M}, with some analogues only displaying inhibition. An optimal ligand for both receptor subtypes would destabilize desensitized states as well as increase peak current across a wide range of concentrations. Determining the structural components that exhibit potentiation via destabilization of desensitized states versus potentiation of peak current would allow for a therapeutic agent that has more specificity. Currently, at high concentrations of agonist dFBr only destabilizes desensitized states on the LS subtype, while increasing desensitization on the HS subtype. Peak current at high concentrations of agonist for the HS subtype is still modulated by dFBr, whereas the LS subtype is not. Analogues would help elucidate if dFBr can become more selective in its actions for the HS versus LS subtype, enhancing its potential as a therapeutic agent. Studies investigating the effects structural modifications have on a single-channel level furthers the drug design process and leads to more rational drug design.

4.2.3 Investigation of the Mechanisms of Action of NS206 and NS9283

The positive allosteric modulator dFBr displays two types of modulation, affecting desensitization as well as peak currents. Unlike dFBr, NS206 and NS9283 compounds each display a single type of modulation via different mechanisms. The compounds differ in their structure as well as their binding sites, and thus behave very differently.
NS206 potentiates both stoichiometric forms of the receptor displaying higher efficacy on HS receptors increasing ACh evoked currents 3-4fold and potentiates LS receptors 1.5 fold. While NS206 potentiates both stoichiometric forms of the receptor, NS9283 potentiates only the LS form of the α4β2 receptor (Grupe et al., 2013). Further comparison on the differences between the two compounds, it is found that NS9283 increases potency and gives monophasic properties. Whereas NS206 displays no change in ACh potency and for the LS subtype continues to have biphasic properties (Olsen et al., 2013). Both compounds do not affect receptor desensitization for either stoichiometry and when co-applied, work additively. It would be worthwhile investigating how the mechanisms of modulation differ between these two compounds, whether there is an additive effect on the single-channel level, and if the mechanisms are similar in form to dFBr’s mechanism. Conducting single-channel analysis on two compounds that give rise to individual types of modulation, differ in structure, and mechanism would help increase understanding on what aspects of the α4β2 nAChR kinetic mechanism the compounds affect the most and if all PAMs display similar modes of action. Developing an understanding of the relationship between compound structure, kinetic mechanisms and binding sites is an important aspect of intelligent and efficient drug development.
4.3 References


