Single Channel Properties of P2X$_2$ Purinoceptors

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Abstract The single channel properties of cloned P2X$_2$ purinoceptors expressed in human embryonic kidney (HEK) 293 cells and Xenopus oocytes were studied in outside-out patches. The mean single channel current-voltage relationship exhibited inward rectification in symmetric solutions with a chord conductance of ~30 pS at ~100 mV in 145 mM NaCl. The channel open state exhibited fast flickering with significant power beyond 10 kHz. Conformational changes, not ionic blockade, appeared responsible for the flickering. The equilibrium constant of Na$^+$ binding in the pore was ~150 mM at 0 mV and voltage dependent. The binding site appeared to be ~0.2 of the electrical distance from the extracellular surface. The mean channel current and the excess noise had the selectivity: K$^+ >$ Rb$^+ >$ Cs$^+ >$ Na$^+ >$ Li$^+$. ATP increased the probability of being open ($P_o$) to a maximum of 0.6 with an EC$_{50}$ of 11.2 μM and a Hill coefficient of 2.3. Lowering extracellular pH enhanced the apparent affinity of the channel for ATP with a pK$_a$ of ~7.9, but did not cause a proton block of the open channel. High pH slowed the rise time to steps of ATP without affecting the fall time. The mean single channel amplitude was independent of pH, but the excess noise increased with decreasing pH. Kinetic analysis showed that ATP shortened the mean closed time but did not affect the mean open time. Maximum likelihood kinetic fitting of idealized single channel currents at different ATP concentrations produced a model with four sequential closed states (three binding steps) branching to two open states that converged on a final closed state. The ATP association rates increased with the sequential binding of ATP showing that the binding sites are not independent, but positively cooperative. Partially liganded channels do not appear to open. The predicted $P_o$ vs. ATP concentration closely matches the single channel current dose-response curve.

Key words: adenosine triphosphate • kinetics • permeation • selectivity • channel

Introduction

P2X purinoceptors are ligand-gated cation channels that are activated by extracellular ATP and its analogues. These receptors exist in excitable and nonexcitable cells, including neurons, smooth and cardiac muscles, glands, astrocytes, microglia, and B lymphocytes (Nakazawa et al., 1990a; Bean, 1992; Walz et al., 1994; Bretschneider et al., 1995; Capiod, 1998; McQueen et al., 1998). During the past few years, seven P2X purinoceptor subunits (P2X$_1$–P2X$_7$) have been cloned (Brake et al., 1994; Valera et al., 1994; Bo et al., 1995; Lewis et al., 1995; Chen et al., 1995; Buell et al., 1996; Seguela et al., 1996; Soto et al., 1996; Surprenant et al., 1996; Wang et al., 1996; Rassendren et al., 1997b). The P2X family has a distinctive motif for ligand-gated ion channels, with each subunit containing two hydrophobic transmembrane domains (M1 and M2) joined by a large intervening hydrophilic extracellular loop (Brake et al., 1994). The cDNA of each receptor is ~2,000 bp in length and has a single open reading frame encoding ~400 amino acids. A comparison of the amino acid sequences of the seven members shows an overall similarity of 35–50% (Collo et al., 1996; North, 1996; Surprenant et al., 1996).

Dose–response analyses of the cloned receptors made with whole cell currents revealed a Hill coefficient larger than 1, suggesting that activation of the channel involves more than one agonist. This is consistent with experiments on the native receptors in PC12 cells and sensory neurons (Friel, 1988; Nakazawa et al., 1991; Bean, 1992; Ugur et al., 1997). Studies aimed at measuring the subunit stoichiometry predict that the naturally assembled form of P2X receptor channels contains three subunits (Nicke et al., 1998).

All of the P2X clones can be expressed in heterologous cells, such as HEK 293 cells and Xenopus oocytes. ATP is a potent agonist for all cloned P2X receptors, and the receptors are highly selective for ATP over most other adenosine derivatives. However, benzyl-ATP is 10-fold more potent than ATP in activating P2X$_7$ receptors (Surprenant et al., 1996). It is interesting to point out that α,β-methylene ATP is a poor agonist for the subtypes that do not show desensitization: P2X$_2$, P2X$_4$, P2X$_5$, P2X$_6$, and P2X$_7$, but is a potent agonist for P2X$_1$ and P2X$_3$ receptors that do desensitize.
Most studies of cloned P2X receptors have focused on the primary structure and pharmacology based on whole cell currents, while only a small amount of work has been done on the single channel properties. Single channel currents from P2X<sub>1</sub> receptors expressed in Xenopus oocytes were reported to have a mean amplitude of ~2 pA at −140 mV and a chord conductance of 19 pS between −140 and −80 mV (Valera et al., 1994). The conductance for P2X<sub>1</sub>, P2X<sub>2</sub>, and P2X<sub>4</sub> channels expressed in Chinese hamster ovary cells were ~18, 21, and 9 pS, respectively, at −100 mV with 150 mM extracellular NaCl, but the openings of P2X<sub>3</sub> were not resolved (Evans, 1996).

To provide a firmer basis for further analysis of the P2X family, we have examined P2X<sub>2</sub> receptors at the single channel level. We have characterized the current-voltage (I-V)<sup>1</sup> relationships, cation selectivity of permeation, ATP sensitivity, proton modulation, and gating kinetics.

**Materials and Methods**

**Expression Systems**

P2X<sub>2</sub> receptors were expressed either in stably transfected human embryonic kidney 293 (HEK 293) cells or in Xenopus oocytes by mRNA injection (Rudy and Iverson, 1992). Since receptor expression is generally too high to obtain patches with only a single channel, we decreased the expression of the receptors in Xenopus oocytes by reducing the amount of mRNA to 25 ng, lowering the incubation temperature from 17° to 14°C, and shorting the incubation time to 16 h.

For electrophysiological experiments, HEK 293 cells were cultured at 37°C for 1-2 d after passage. The medium for HEK 293 cells contained 90% DMEM/F12, 10% heat inactivated fetal calf serum, and 300 μg/ml genetin (G418). The media were adjusted to pH 7.35 with NaOH and sterilized by filtration. The incubation medium (ND96) for Xenopus oocytes contained (mM): 96 NaCl, 2 KCl, 1 MgCl<sub>2</sub>, 1.8 CaCl<sub>2</sub>, 5 HEPES, titrated to pH 7.5 with NaOH. All chemicals were purchased from Sigma Chemical Co.

**Electrophysiology**

We made patch clamp recordings from HEK 293 cells 1-2 d after passage and from Xenopus oocytes 16 h after injecting mRNA. Single channel currents from outside-out patches and whole cell currents were recorded at room temperature (Hamill et al., 1981). Recording pipettes, pulled from borosilicate glass (World Precision Instruments, Inc.) and coated with Sylgard, had resistances of 10-20 MΩ.

For recording from HEK 293 cells, the pipette solution contained (mM): 140 NaF, 5 NaCl, 11 EGTA, 10 HEPES, pH 7.4. The bath solution and control perfusion solutions were the same and contained (mM): 145 NaCl, 2 KCl, 1 MgCl<sub>2</sub>, 1 CaCl<sub>2</sub>, 11 glucose, 10 HEPES, pH 7.4. For Xenopus oocytes, the pipette solution contained 90 mM NaF instead of 140 mM NaF and other components were the same as for HEK 293 cells; the bath and control perfusion solutions were the same as those used for HEK 293 cells except that they contained 100 instead of 145 mM NaCl. The patch perfusion solutions were the same as the bath solutions, except for modified divalent and ATP concentrations. Perfusion was driven by an ALA BPS-4 perfusion system (ALA Scientific Instruments). To investigate the cation selectivity of the channels, we substituted different cations for Na<sup>+</sup> ion in the perfusate. To investigate the affinity of Na<sup>+</sup> for the channel pore, we varied the extracellular NaCl concentration without compensation by other ions, while the pipette solution was kept constant. The resulting change of ionic strength caused the development of small liquid junction potentials between the bulk solution and the perfusate. We calculated these potentials according to the Henderson equation (Barry and Lynch, 1991). For solution exchanges from 100 to 150, 125, 100, 75, 50, and 25 mM NaCl, the junction potentials were −2.1, −1.2, 0, 2.5, 3.6, and 7.3 mV, respectively. Because these values are small compared with the holding potentials, we did not correct the membrane potential when we calculated the chord conductances.

Currents were recorded with a patch clamp amplifier (AXOPATCH 200B, Axon Instruments), and stored on videotape using a digital data recorder (VR-10A; Intrustron Corp.). The data were low-pass filtered at 5–20 kHz bandwidth (~3 dB) and digitized at sampling intervals of 0.025–0.1 ms using a LabView data acquisition program (National Instruments).

**Analysis**

Amplitudes and excess channel noise. The mean amplitudes of single channel currents were determined by all-points amplitude histograms that were fit to a sum of two Gaussian distributions. Chord conductances were calculated assuming a reversal potential of 0 mV. The excess open channel noise (σ<sub>eo</sub>) was computed as the root mean square (rms) difference between the variances of the open channel current and the shut channel current (Sivilotti et al., 1997).

The probability of the channel being open, P<sub>o</sub>, was defined as the ratio of open channel area (A<sub>o</sub>) to the total area (A<sub>o</sub> + A<sub>c</sub>). In the all-points amplitude histogram:

\[
P_o = \frac{A_o}{A_o + A_c}.\]

This calculation is insensitive to short events. In our initial analysis, we treated the channel as having two amplitude classes, open and closed. All rapid gating events associated with the open channel were treated as noise. Essentially we defined open as "not closed."

Power spectra. Power spectra, S(f), were computed using the fast Fourier transform routine in LabView™. We used records of 50–100-ms duration associated with open and closed states. The power spectrum of the excess noise was obtained by subtracting the spectrum of the closed state spectrum from that of the open state. The spectra were fit with the sum of a Lorentzian plus a constant:

\[
S(f) = \frac{S(0)}{1 + (f/f_0)^2} + S_1.
\]

The rms noise, σ<sub>1</sub>, from the Lorentzian component is:

\[
\sigma_1 = \sqrt{\frac{\pi}{2}} \frac{1}{f_0} S(0).
\]

<sup>1</sup>Abbreviations used in this paper: AIC, asymptotic information criterion; HEPK cell, human embryonic kidney cell; I–V, current-voltage; rms, root mean square.
Thermal and shot noise. The thermal (or Johnson) and shot noise contributions were calculated according to Defelice (1981). The power spectra of thermal noise is white; i.e., it does not vary with frequency. Its value, $S_{th}$, is given by:

$$S_{th} = \frac{4k_BT}{R}, \quad (4)$$

where $k_B T$ is Boltzmann’s constant times the absolute temperature, and $R$ is the equivalent resistance of the open channel. The rms noise in a bandwidth $f$ is given by:

$$\sigma_{th} = \sqrt{\frac{4k_BTf}{R}}. \quad (5)$$

The power spectrum of shot noise, $S_{sh}$, is also white and given by:

$$S_{sh} = 2q_{i}, \quad (6)$$

where $q$ is the charge of an elementary charge carrier, and $i$ is the single channel current amplitude. The rms amplitude of shot noise over bandwidth $f$ is:

$$\sigma_{sh} = \sqrt{2q_{i}f}. \quad (7)$$

Multiple conductance levels. We tried to see if there were discrete subconductance levels making up the open channel “buzz mode” by using a maximum-point likelihood method (MPL; www.qub.buffalo.edu) that use the Baum-Welch algorithm (Chung et al., 1990). While we did get convergence at four to six levels, these levels were not consistent among data from different ATP concentrations. When we performed global fitting on data from different ATP concentrations, the rates were assumed to be exponential functions of voltage; i.e., $k_{ij} = k_{ij}(0)\exp(-z_{0ij}/k_B T)$, where $k_{ij}(0)$ is the apparent rate constant at 0 mV and the specified ATP concentration, and $z_{0ij}$ is the effective sensing charge.

We used Akaike’s asymptotic information criterion (AIC) to rank different kinetic models (Vandenberg and Horn, 1984; Horn, 1987):

$$AIC = 2 \left[ \sum_{j=1}^{n} \log(\text{maximum likelihood}) - (\text{number of free parameters of the model}) \times (\text{number of data sets}) \right] \quad (8)$$

where $j$ is the number of the data set. The model with a higher AIC is considered a better fit.

Simulation. We used Origin (Microcal Software, Inc.) and Scientist (MicroMath Scientific Software, Inc.) software to simulate and fit data.

Results

Basic Features of Single Channel Currents

Fig. 1 A shows a typical single channel current, activated by 1.5 mM ATP at a membrane potential of $-100$ mV from a stably transfected HEK 293 cell. Channel openings appeared as flickery bursts with ill-defined conductance levels. There were a few clear closures and subconductance levels within a burst, but discrete levels could not be resolved from the all-points histogram (see Fig. 1 B). The spread of current levels was reflected by the much larger standard deviation of the open than the closed component, each of which could be fit reasonably well with a single Gaussian. In Fig. 1 B, the mean open current amplitude is $3.2$ pA, equivalent to a chord conductance of $32$ pS. The standard deviations of the open and closed histograms are $0.95$ and $0.24$ pA, respectively, so that the excess open channel noise $\sigma_{ex}$ is $0.92$ pA; i.e., $29\%$ of the mean current amplitude. Since the mean is clearly less than the peak current, we obtained a closer estimate of the maximal open channel current by measuring the mean of extreme values. Comparing the upper $5\%$ of the two distributions, the peak amplitude and conductance were $4.3$ pA (Fig. 1 B, arrow) and $43$ pS, respectively, a closer estimate of the maximum ion flux. However, for convenience in discussing later results, unless specified otherwise, “channel current” and “conductance” will refer to the mean rather than the peak values.

To further characterize the open channel noise, we followed the procedures of Sigworth (1985a) for noise analysis using differential power spectra. We compared...

\[ k_{ij}(0) \text{[ATP]}, \quad \text{where } k_{ij}(0) \text{ is an intrinsic rate constant at the specified voltage}, \quad \text{while the other rates were assumed to be independent of ATP. When we performed global fitting on data from different voltages at same ATP concentration, the rates were assumed to be exponential functions of voltage; i.e., } k_{ij} = k_{ij}(0)\exp(-z_{0ij}/k_B T), \quad \text{where } k_{ij}(0) \text{ is the apparent rate constant at } 0 \text{ mV and the specified ATP concentration, and } z_{0ij} \text{ is the effective sensing charge.} \]

...
the power spectra of excess open channel fluctuations with the expected thermal and shot noise (Fig. 1 C). The open channel spectrum was well fit with a Lorentzian with \( f_c = 264 \text{ Hz} \), equivalent to a relaxation time of 0.62 ms, plus a constant. This noise is much larger than the expected thermal or shot noise, suggesting that the fluctuations most likely arise from rapid conformational changes in the channel.

I–V Curve

Fig. 2 A shows the single channel currents recorded from HEK 293 cells activated by 2 \( \mu \text{M} \) ATP at different holding potentials using outside-out patches with symmetrical \( \text{Na}^+ \). The currents became small and noisy at positive holding potentials so that the unitary currents were not discernible. The single channel I–V curve (Fig. 2 B) exhibited a strong inward rectification similar to whole cell currents recorded under the same conditions (Fig. 3, A and B). The rapid fluctuation of current in the “open” state was maintained at all holding potentials.

Dose–Response Curve of Single Channel Currents

We were unsuccessful in obtaining outside-out patches containing only a single P2X\(_2\) channel when the receptor was expressed in HEK 293 cells. However, we were
Figure 2. The current-voltage relationship of single channel currents. (A) Single channel currents of an outside-out patch from HEK 293 cells activated by 2 μM ATP (1 mM extracellular Mg^{2+} and Ca^{2+} at different membrane potentials, symmetrical Na^{+} solutions containing 145 mM extracellular NaCl/145 mM intracellular NaF). The data were filtered at 5 kHz and digitized at 10 kHz. All of the current traces in this figure are from the same patch. (B) Mean I–V relationship of single channel currents. The error bars indicate the standard deviation of the single channel currents from the all-points histograms. The single channel I–V relationship shows strong inward rectification despite exposure to identical Na^{+} solutions across the patch. The same result was obtained when single channel currents were recorded in the absence of Mg^{2+} and Ca^{2+}; therefore, divalent cations are not responsible for the rectification.

Figure 3. The current-voltage relationship of whole cell currents (WCCs). (A) Whole cell currents from an HEK 293 cell at different holding potentials: −100 to +80 mV at 20-mV intervals. Voltage drops from incomplete series resistance compensation were subtracted from the membrane potential. The currents were activated by 10 μM ATP in the presence of 1 mM extracellular Ca^{2+} with symmetrical Na^{+} solutions: 145 mM extracellular NaCl/145 mM intracellular NaF. The data were filtered at 2 kHz and digitized at 5 kHz. (B) The mean WCCs (±SD) (○, n = 5) and predicted WCCs activated by 10 μM ATP. The I–V curve exhibits strong inward rectification, similar to the single channel currents shown in Fig. 2. The reversal potential was −0 mV. The predicted WCCs were calculated using Eq. 15 where i (single channel amplitude) was taken from Fig. 2, and P_o from the calculations for Model 1-4 (△) (Fig. 13), and Eq. 16 (□) as a function of voltage. The number of channels was chosen so that the predicted WCC at −60 mV was equal to the experimental data. The predicted WCCs match reasonably well with the experimental data.
able to obtain patches with a single channel from Xenopus oocytes provided we carefully controlled the amount of mRNA, and the time and temperature of incubation. Fig. 4 A shows the single channel currents from an outside-out patch activated by different concentrations of ATP. As expected, increasing the concentration of ATP increased $P_o$. The all-points histograms (Fig. 4 B) show that the average current and excess noise were independent of ATP concentration ($i = 3.5 \text{ pA}$ and $\sigma_{ex} = 1.6 \text{ pA}$ at $-120 \text{ mV}$). Thus there is no indication that ATP is blocking the open channel. We calculated the probability of being open at each ATP concentration from the amplitude histograms using ratio of the open area to the total area (Fig. 4 C). The open probability saturated when the ATP concentration reached 30 $\mu M$. The dose-response curve was fitted by the Hill equation with a Hill coefficient of 2.3, an $EC_{50}$ of 11.2 $\mu M$, and a maximal open probability of 0.61. The Hill coefficient and $EC_{50}$ are similar to those obtained from the dose-response curves of whole cell currents of our own data (not shown) and the literature (Brake et al., 1994), indicating that there are at least three subunits in a functional P2X$_2$ receptor ion channel. The data from this patch were very stable and used later in the comparison of kinetic models.

Na$^+$ Conduction through the Ion Channel

To investigate the affinity of Na$^+$ for the open channel, we measured single channel amplitudes at holding potentials of $-80$, $-100$, $-120$, $-140 \text{ mV}$ for extracellular NaCl concentrations of 10, 25, 50, 75, 100, 125, and 150...
The data were digitized at 20 kHz and low pass filtered at 10 kHz.

The equilibrium constant, \(K_s\), increased with depolarization (Fig. 6 B). The relationship between the conductance, \(g\), and the ATP concentration (Table I), the selectivity of the relative noise for Li\(^+\) suggests that it can affect the flickering kinetics. We compared \(\sigma_{ex}/i\) with the thermal, \(\sigma_{th}\), and shot, \(\sigma_{sh}\), noise (Table I). Again, \(\sigma_{th}\) and \(\sigma_{sh}\) were very small compared with \(\sigma_{ex}\), and the ratio \((\sigma_{th} + \sigma_{sh})\) to \(\sigma_{ex}\) ranged from 8 to 14% depending on the ions. The relative noise caused by the open channel fluctuations, when corrected for thermal and shot noise \((\sigma_{ex}^2 - \sigma_{th}^2 - \sigma_{sh}^2)^{1/2}/i\), followed the same cation sequence as \(\sigma_{ex}/i\).

Effect of pH

The effect of pH on channel activation. The effect of extracellular pH on the channel currents is illustrated in Fig. 8 A. Multiple-channel outside-out currents activated by 2 \(\mu\)M ATP increased ~10-fold when pH was decreased from 8.3 to 6.8, and saturated with further decreases in pH (Fig. 8 B). The fluctuations in these multichannel currents at higher pH appeared to be dominated by the overlap of independent channels, so that at pH 8.3, where the mean current is small, single channel events were visible. At pH 6.3, the current saturated and the frequency of fluctuations increased dramatically, apparently dominated by the open channel noise. As is clear from the rise time of the currents, the activation rate decreased with increasing pH, and the fall time remained constant (Fig. 8 C). The potential of channel activity by protons is similar to the effect of increasing the ATP concentration, suggesting that protons may increase the affinity of the binding

\[
\gamma = \frac{\gamma_{\max}}{1+K_s/\text{[NaCl]}},
\]

\[
K_s(V) = K_s(0) \exp(-z\delta V/k_B T),
\]
site for ATP. The pK_s was \( \approx 7.9 \) and the Hill coefficient was 2.5, again suggesting that there are more than two subunits in the channel.

Effect of pH on single channel properties. The results above and published studies on the effect of pH were based on whole cell or multi-channel recordings (Li et al., 1996, 1997; King et al., 1997; Stoop et al., 1997; Wildman et al., 1998). To explore the possible effect of pH on gating and channel conductance, we examined the effect of pH on single channels. To obtain single channel activity from the stable cell lines, we exposed the patch to ATP for long times, so that run down reduced the number of active channels. Fig. 9 A shows these currents recorded at different values of extracellular pH. We measured the single channel amplitude and excess noise from the all-points amplitude histograms. The mean amplitude of the current was independent of pH, but the excess open channel noise increased with decreasing pH (Table II). As visible in Fig. 9 A, the frequency of brief closures within open channel bursts appeared to increase as pH decreased. These interruptions were longer than the normal fast “flickery” behavior. It has been suggested that protons may block an open channel (Yellen, 1984).

To further characterize this phenomenon, we computed the power spectra of the open channel fluctuations (Fig. 9 B), and fit them with a Lorentzian plus a constant (Eq. 3). The Lorentzian represents a two-state relaxation process.
whose characteristic time constant $\tau$ is related to the corner frequency, $f_c$, by:

$$\tau = \frac{1}{2\pi f_c}. \quad (11)$$

Diffusional block of the open channel can be described as a two-state model (Scheme I), where O is the open state and Cb is the protonated-blocked state.

$$\begin{align*}
O + H^+ &\xrightarrow{\alpha} \text{Cb} \\
\text{Cb} &\xrightarrow{\beta} O.
\end{align*} \quad \text{(Scheme I)}$$

$\alpha$ and $\beta$ are the blocking and unblocking rate constants. The relaxation time for this two-state process is related to the rate constants by:

$$\frac{1}{\tau} = 2\pi f_c = \alpha[H^+] + \beta. \quad (12)$$

The prediction of proton block is that $f_c$ increases linearly with increasing proton concentration. However, our data show that $f_c$ decreased with increasing proton concentration (Fig. 9B). To further examine the possibility of proton block, we analyzed bursts kinetically using the maximum likelihood method with a two-state model. We fit the extracted $\alpha$'s and $\beta$'s at different pH to an equation of the form:

$$\alpha = \alpha_0[H^+]^n, \quad (13)$$

where $\alpha_0 = 224 \mu M^{-0.32} s^{-1}$, $n_\alpha = 0.33$, $\beta_0 = 1,493 \mu M^{-0.16} s^{-1}$, and $n_\beta = 0.16$. Since $\alpha$ and $\beta$ are not directly proportional to the proton concentration, a single site model appears to be inappropriate. We speculate that we may be titrating several sites that display negative cooperativity. The effect of pH on the open channel is to modify the conformation of the channel rather than to provide a simple proton block. Table II summarizes the open channel properties at different pH. Remarkably, the effect of pH on the mean open channel current is negligible.

**Kinetic Analysis with the Maximum-Interval Likelihood Method**

To understand the kinetics of agonist binding and channel gating, we applied the maximum likelihood method to data from outside-out patches that were stable over time and ATP concentration (Fig. 4A). We began by fitting simple noncyclic models using the maximum-likelihood interval analysis and used AIC ranking to select a preferred model. The analysis was hierarchical in the sense that we fit portions of the reaction scheme under restricted conditions, and then merged these models to create a full description. The kinetic description required: (a) the number of closed and open states, (b) the connections between states, and (c) the values of the rate constants between the states and their dependence on concentration and voltage.

**Mean open and close times.** The data was idealized into two classes: open and closed (see Fig. 10A). We did not attempt to idealize the data making up the bulk of the flickery open channel activity since the amplitudes

**Table I** The Effects of Different Cations on Open Channel Properties

<table>
<thead>
<tr>
<th>Ion</th>
<th>Li$^+$</th>
<th>Na$^+$</th>
<th>K$^+$</th>
<th>Rb$^+$</th>
<th>Cs$^+$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$i$ (pA)</td>
<td>3.48</td>
<td>4.17</td>
<td>5.86</td>
<td>5.63</td>
<td>5.36</td>
</tr>
<tr>
<td>$\gamma$ (pS)</td>
<td>29.0</td>
<td>34.8</td>
<td>48.8</td>
<td>46.9</td>
<td>44.7</td>
</tr>
<tr>
<td>$\sigma_{ex}$ (pA)</td>
<td>0.92</td>
<td>1.40</td>
<td>1.90</td>
<td>1.96</td>
<td>1.79</td>
</tr>
<tr>
<td>$\sigma_{th}$ (pA)</td>
<td>0.646</td>
<td>0.336</td>
<td>0.034</td>
<td>0.334</td>
<td>0.344</td>
</tr>
<tr>
<td>$\sigma_{sh}$ (pA)</td>
<td>0.049</td>
<td>0.053</td>
<td>0.063</td>
<td>0.062</td>
<td>0.061</td>
</tr>
<tr>
<td>$\sigma_{sh}$ (pA)</td>
<td>0.075</td>
<td>0.082</td>
<td>0.097</td>
<td>0.096</td>
<td>0.094</td>
</tr>
<tr>
<td>$\sigma_{sh}$ (pA)</td>
<td>0.035</td>
<td>0.033</td>
<td>0.324</td>
<td>0.348</td>
<td>0.333</td>
</tr>
</tbody>
</table>

The currents were recorded at $-120$ mV. Liquid junction potentials were not corrected. $\gamma$ is chord conductance, $\sigma_{ex}$ excess open channel noise, $\sigma_{th}$ thermal noise, and $\sigma_{sh}$ shot noise.

**Table II** Summary of Open Channel Properties at Different pH

<table>
<thead>
<tr>
<th>pH</th>
<th>Amplitude</th>
<th>$\sigma_{ex}$</th>
<th>$f_c$</th>
<th>$\tau$</th>
<th>$S_0$</th>
<th>$S_1$</th>
<th>$\sigma_i$</th>
</tr>
</thead>
<tbody>
<tr>
<td>6.3</td>
<td>3.61</td>
<td>1.33</td>
<td>205</td>
<td>0.72</td>
<td>1.01</td>
<td>2.31</td>
<td>0.57</td>
</tr>
<tr>
<td>6.8</td>
<td>3.60</td>
<td>1.32</td>
<td>224</td>
<td>0.66</td>
<td>1.38</td>
<td>3.00</td>
<td>0.70</td>
</tr>
<tr>
<td>7.3</td>
<td>3.73</td>
<td>1.13</td>
<td>240</td>
<td>0.61</td>
<td>1.34</td>
<td>3.64</td>
<td>0.71</td>
</tr>
<tr>
<td>7.8</td>
<td>3.50</td>
<td>1.12</td>
<td>295</td>
<td>0.50</td>
<td>0.94</td>
<td>3.79</td>
<td>0.66</td>
</tr>
<tr>
<td>8.3</td>
<td>3.54</td>
<td>1.17</td>
<td>336</td>
<td>0.44</td>
<td>0.62</td>
<td>3.37</td>
<td>0.57</td>
</tr>
</tbody>
</table>

The currents were recorded at $-120$ mV. Liquid junction potentials were not corrected. $\gamma$ is chord conductance, $\sigma_{ex}$ excess open channel noise, $\sigma_{th}$ thermal noise, and $\sigma_{sh}$ shot noise.
were uncertain, but instead defined open as a single conductance state possessing a lot of noise. The probability of a channel being open increased with ATP, as shown in Fig. 4 C. This could result from an increase in mean open time, a decrease in mean closed time, or a combination. Fig. 10 B shows the mean open and closed times calculated from idealized single channel currents, and plotted as a function of ATP concentration. The mean closed time dramatically decreased with the increase in ATP and saturated at 30 \( \text{mM} \), while the mean open time was not affected by ATP. The results indicate that ATP controls the rate at which the channel opens, but not the rate at which it closes.

Duration histograms. Fig. 10 C shows the open-and closed-time histograms from idealized single channel currents induced by different ATP concentrations (Fig. 10 A). The open-time histograms have two peaks and the closed-time histograms have at least three peaks at low concentration. When the ATP concentration was increased, the intermediate and long time constant peaks of the closed time distribution merged and only two peaks were visible. These results indicate that the channel has at least three closed and two open states.

Kinetic model comparison. We made quantitative comparisons of various kinetic models to determine which model best described the behavior. The models were limited to three closed and two open states (of the same conductance), and at most 10 rate constants. These constraints proved necessary to obtain unique solutions during optimization. There are 98 unique models with that many states. Further constraints were imposed to simplify analysis. (a) We discarded models in which the unliganded states were open because we did not see any spontaneous openings in the absence of ATP. (b) Following traditional models for other ligand-gated channels, the closed states were connected so as to represent the binding of ATP. To evaluate the possible topologies, we used the program MSEARCH (www.qub.buffalo.edu) to compare the likelihood of all remaining models. The program evaluates all topologically unique models having a specified number of states of each conductance and optimizes the rate constants for each one. For this stage of the
analysis, we used three data sets obtained at 5, 10, and 15 μM ATP. We calculated the likelihood of each model by adding the log-likelihoods from each concentration. This is more a test of the topology of the models than a test of the optimal values of the rate constants since the rate constants will change over concentration, but the connectivity won’t. Fig. 11 shows the eight kinetic models that converged on all data sets within 100 iterations. They are listed in the order of AIC rank.

To determine which model was best, we compared the log(maximum likelihood)s and AIC rankings (Table III). The likelihoods of Models 1, 2, and 3 (Fig. 11) are the same, but Model 3 has two more parameters and, hence, a lower AIC rank. Models 1 and 2 have the same number of parameters, likelihood, and AIC rank, so we can not tell the difference between them. Model 7 (Fig. 11) has a larger likelihood than Models 1 and 2; however, its AIC ranking is much lower because of the increased number of parameters. Model 8 (Fig. 11), which has a partially liganded open state, has the smallest likelihood and lowest AIC rank. When Models 1 and 2 were compared across concentration, they were indistinguishable and, for simplicity in what follows, we arbitrarily selected Model 1. In both models, state C1 is unliganded, C2 and C3 are liganded, and O4 and O5 are open. k12 and k23 are the agonist association rates, k21 and k32 are the agonist dissociation rates, k34 and k35 are the channel opening rates, and k43 and k53 are the channel closing rates.

The rate constants governing ATP binding and gating were solved by fitting across a range of ATP concentrations. Fig. 12 shows the rates from the model at bottom (from Fig. 11, Model 1) as a function of ATP concentration when the data from each concentration were fit independently. The association rates k12 and k23 showed a strong dependence on ATP concentration in the 5–20 μM range. However, when the ATP concentration was >20 μM, the rate constants appeared to satu-
rate and the error limits on the parameters increased. A concentration-driven rate should not saturate, but there are a few explanations. First, there may be a concentration-independent state not contained in the model. Second, $k_{12}$ and $k_{23}$ approach $k_{35}$ at high ATP concentration, making $k_{35}$ rate limiting and rendering the optimizer incapable of properly solving the model. Third, if $k_{12}$ and $k_{23}$ are linearly proportional to concentration, then the intrinsic rate constants of both $k_{12}$ and $k_{23}$ are $\sim 2 \times 10^7$ M$^{-1}$s$^{-1}$, which is approaching the diffusion limit.

We tested the first possibility by adding concentration-independent states to the model in Fig. 12, but that did not prevent the association rates from saturating. We tested the identifiability of the model by simulating the model across concentrations (SIMU; www.qub.buffalo.edu) and attempting to extract the rate constants using maximal interval likelihood. Fitting the simulated data, we found that the estimated rate constants also saturated (see below) so that the correct model is not identifiable with data from a single concentration. As far as the diffusion limit providing a true saturation, further experiments are required to test that prediction. However, we currently believe that the apparent saturation is an artifact caused by the lack of identifiability of the model at high concentrations.

(Details of the test on the artifactual origin of saturating rates. We simulated data using Model 1 [Fig. 11], with the rate constants $k_{12}$ and $k_{23}$ increasing linearly with ATP in the 5–50 μM range. The intrinsic rate constants $k_{12}(0)$ and $k_{23}(0)$, obtained from the slope of $k_{12}$ and $k_{23}$ versus ATP from 5–30 μM (Fig. 12), were 14...
and 22 μM⁻¹s⁻¹, respectively. All other rate constants were made independent of ATP and set to values averaged across the data sets. We then analyzed the simulated data as if it were experimental data. The recovered rate constants were similar to the values used to simulate the data for ATP₂₀μM. At higher ATP levels, however, the estimated values of $k_{12}$ saturated and $k_{21}$ even decreased. Large error limits also occurred in $k_{12}$ and $k_{21}$ at the high concentrations [Table IV]. Thus, Model 1 [Fig. 11] cannot uniquely fit data at single high ATP concentration.

To improve identifiability, we fit the data simultaneously across all concentrations. Such global fitting makes the likelihood surface steeper (Qin et al., 1996a). We assumed that the association rate constants were proportional to the ATP concentration, and the other rate constants were independent of ATP (see materials and methods). This time, the rate constants derived from global fitting of simulated data were very close to the values used for simulation (Table V). The results of global fitting to the experimental data are listed in Table VI. It is worth noting that the second ATP association rate constant, $k_{23}(0)$, is larger than the first, $k_{12}(0)$. This result shows that the binding sites are not independent, but that binding to one site modifies binding to the other. With independent sites, the association rate should decrease as the number of free sites decreases. The conclusion is quite model independent; for every model we tested, the association rates increased with proximity to the open state (see below).

Model simplification and expansion. As ATP concentration increased, the three peaks in the closed time duration histograms became two (Fig. 10 C), suggesting that at high ATP concentration, Model 1 (Fig. 11) could be simplified by removal of state C₁ (Fig. 13, Model 1-1). When we fit the kinetics of high concentrations of ATP

**Table III**

<table>
<thead>
<tr>
<th>Model</th>
<th>log(Maximum likelihood)</th>
<th>AIC rank</th>
<th>Number of parameters</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>0</td>
<td>1</td>
<td>8</td>
</tr>
<tr>
<td>2</td>
<td>0</td>
<td>1</td>
<td>8</td>
</tr>
<tr>
<td>3</td>
<td>0</td>
<td>2</td>
<td>10</td>
</tr>
<tr>
<td>4</td>
<td>−0.6</td>
<td>3</td>
<td>8</td>
</tr>
<tr>
<td>5</td>
<td>−0.6</td>
<td>3</td>
<td>8</td>
</tr>
<tr>
<td>6</td>
<td>−0.6</td>
<td>4</td>
<td>10</td>
</tr>
<tr>
<td>7</td>
<td>+3.6</td>
<td>5</td>
<td>10</td>
</tr>
<tr>
<td>8</td>
<td>−46</td>
<td>5</td>
<td>8</td>
</tr>
</tbody>
</table>

Data sets from three different ATP concentrations were used for the calculation of likelihoods and AIC rankings. The lowest number of AIC rank is considered best.
by Model 1 and Model 1-1, the likelihoods were equal. Thus, at high ATP, $k_{12}$ gets so fast that $C_1$ is rarely occupied and Model 1-1 is sufficient to describe the kinetics. However, a large difference in maximum likelihoods arose when we fitted Models 1 and 1-1 to the data at low concentrations of ATP. Model 1-1 can well describe the kinetics of single channel currents of high ATP, but not low.

Our model has only two binding steps. The fact that the $P_o$ curve has a Hill coefficient of 2.3 suggests that there are at least three binding sites in the P2X$_2$ channel. Since it is a homomer, this implies that three or more subunits are needed to form the channel. A more realistic model should have at least one additional partially liganded closed state (Fig. 13, Model 1-2).

The rate constants from Model 1-2 (Fig. 13) are shown in Table VI. Again in this model, the first ATP binding step speeds up the transition rates near the open state. The transition rates near the open state are similar between Model 1 (Fig. 11) and Model 1-2. While Model 1-2 has two more free parameters than Model 1, it has 5.4 units higher likelihood so that Model 1-2 is preferred (see Table VIII). The predicted $P_o$ as a function of ATP concentration is plotted in Fig. 4 C (◊) and fit with the Hill equation with a Hill coefficient of 1.5, an EC$_{50}$ of 17.4 μM, and maximal $P_o$ of 0.74. However, compared with the experimental data, the EC$_{50}$ and maximal $P_o$ are too large and the Hill coefficient too small. These discrepancies can be reduced by connecting an ATP-independent closed state to the open states. Additional evidence for this closed state comes from the closed time histogram that has two components at saturating ATP (Fig. 10 C). Adding a closed state to the right of the open states in Model 1-2 produces Model 1-4 (Fig. 13). This modification corrects the prediction of the dose–response curve. Similarly, adding one more closed state to Model 1 produces Model 1-3 (Fig. 13).

Constraining Models 1-3 and 1-4 (Fig. 13) with detailed balance in the loops, and globally fitting the data from 5 to 50 μM ATP, we obtained rate constants with small error limits (Table VII). The relative likelihoods and the AIC ranking of Model 1 (Fig. 11) and its expanded versions, Models 1-2, 1-3, and 1-4 (Fig. 13) are listed in Table VIII. Models 1-3 and 1-4, which contain loops, have much higher likelihoods than Model 1 or 1-2. Model 1-4 has the highest AIC rank, and therefore is

---

**Figure 12.** The effect of ATP concentration on the rate constants near the open states. The rate constants are based on kinetic Model 1 (shown at bottom), with each data set fit separately. Missed events were corrected by imposing a dead time of 0.05 ms. $k_{12}$ and $k_{23}$ are the most sensitive to the concentration of ATP, increasing over the range from 5 to 20 μM, and saturating at higher concentrations. The rate constants are plotted in two panels to avoid overlap (the scales in both plots are identical).
the preferred model. The rate constants are listed in Table VI and VII. The transition rates near the open states for Models 1 and 1-2, and Models 1-3 and 1-4 are very similar, supporting the hierarchical approach. The predicted probability densities for the open and closed lifetimes of Model 1-4 are shown in Fig. 10C and match the histograms reasonably well. Again, we found that the ATP association rate constants increased with proximity to the open states: \( k_{32}(0) < k_{23}(0) < k_{34}(0) \). This is opposite to what would be expected from independent subunits. Each binding step makes the next faster. This cooperativity of binding appears model independent since all models tested had the same trend. From the Eyreng model for the rates, the energy landscape for the whole reaction is shown in Fig. 14. The kinetic model fits the single channel data quite well. In Fig. 4C, the predicted \( P_0 \) (●) from Model 1-4 (Fig. 13) and its fit to the Hill equation (solid line) are plotted as a function of ATP. The maximal \( P_0 \) (0.64) and \( EC_{50} \) (13.3 \( \mu \)M) are close to those of the experimental data, although the Hill coefficient (1.8) is slightly smaller. These values are much closer to experi-

### Table IV

Comparison of Simulation Values and Fitted Values of Rate Constants from Individual Fitting Based on Model 1 (Fig. 11)

<table>
<thead>
<tr>
<th>[ATP] ( \mu )M</th>
<th>5</th>
<th>10</th>
<th>15</th>
<th>20</th>
<th>30</th>
<th>50</th>
</tr>
</thead>
<tbody>
<tr>
<td>( k_{12} ) Simulation</td>
<td>70</td>
<td>140</td>
<td>210</td>
<td>280</td>
<td>420</td>
<td>700</td>
</tr>
<tr>
<td>Fitted</td>
<td>70 ± 17</td>
<td>146 ± 39</td>
<td>168 ± 99</td>
<td>203 ± 123</td>
<td>107 ± 85</td>
<td>192 ± 160</td>
</tr>
<tr>
<td>( k_{21} ) Simulation</td>
<td>95</td>
<td>95</td>
<td>95</td>
<td>95</td>
<td>95</td>
<td>95</td>
</tr>
<tr>
<td>Fitted</td>
<td>124 ± 60</td>
<td>121 ± 70</td>
<td>42 ± 55</td>
<td>39 ± 39</td>
<td>5 ± 9</td>
<td>5 ± 9</td>
</tr>
<tr>
<td>( k_{32} ) Simulation</td>
<td>110</td>
<td>220</td>
<td>330</td>
<td>440</td>
<td>660</td>
<td>1100</td>
</tr>
<tr>
<td>Fitted</td>
<td>135 ± 38</td>
<td>203 ± 51</td>
<td>236 ± 49</td>
<td>316 ± 54</td>
<td>424 ± 43</td>
<td>726 ± 71</td>
</tr>
<tr>
<td>( k_{33} ) Simulation</td>
<td>1192</td>
<td>1192</td>
<td>1192</td>
<td>1192</td>
<td>1192</td>
<td>1192</td>
</tr>
<tr>
<td>Fitted</td>
<td>985 ± 130</td>
<td>902 ± 106</td>
<td>899 ± 96</td>
<td>867 ± 87</td>
<td>729 ± 81</td>
<td>662 ± 93</td>
</tr>
<tr>
<td>( k_{34} ) Simulation</td>
<td>1145</td>
<td>1145</td>
<td>1145</td>
<td>1145</td>
<td>1145</td>
<td>1145</td>
</tr>
<tr>
<td>Fitted</td>
<td>922 ± 100</td>
<td>978 ± 72</td>
<td>964 ± 65</td>
<td>983 ± 56</td>
<td>916 ± 48</td>
<td>896 ± 45</td>
</tr>
<tr>
<td>( k_{35} ) Simulation</td>
<td>365</td>
<td>365</td>
<td>365</td>
<td>365</td>
<td>365</td>
<td>365</td>
</tr>
<tr>
<td>Fitted</td>
<td>354 ± 22</td>
<td>358 ± 15</td>
<td>356 ± 13</td>
<td>354 ± 11</td>
<td>355 ± 10</td>
<td>351 ± 9</td>
</tr>
<tr>
<td>( k_{36} ) Simulation</td>
<td>418</td>
<td>418</td>
<td>418</td>
<td>418</td>
<td>418</td>
<td>418</td>
</tr>
<tr>
<td>Fitted</td>
<td>307 ± 55</td>
<td>326 ± 39</td>
<td>321 ± 34</td>
<td>325 ± 31</td>
<td>293 ± 26</td>
<td>288 ± 24</td>
</tr>
<tr>
<td>( k_{37} ) Simulation</td>
<td>5335</td>
<td>5335</td>
<td>5335</td>
<td>5335</td>
<td>5335</td>
<td>5335</td>
</tr>
<tr>
<td>Fitted</td>
<td>4795 ± 1060</td>
<td>5494 ± 810</td>
<td>4957 ± 653</td>
<td>4857 ± 540</td>
<td>4750 ± 487</td>
<td>4673 ± 450</td>
</tr>
</tbody>
</table>

\( k_{12} = k_{35}(0)[ATP] \) and \( k_{33} = k_{36}(0)[ATP] \). \( k_{32}(0) \) and \( k_{34}(0) \) are 14 and 22 \( \mu \)M \(^{-1}\) s \(^{-1}\), respectively. All rate constants have units of s \(^{-1}\). Simulation values are rate constants used to simulate single channel currents. Fitted values are the rate constants obtained from maximal likelihood analysis of simulated data, using Model 1. Missed events were corrected by imposing a dead time of 0.05 ms.

### Table V

Comparison of Simulation Values and Fitted Values of Rate Constants from Global Fitting Based on Model 1 (Fig. 11)

<table>
<thead>
<tr>
<th>Rate constant</th>
<th>Simulation</th>
<th>Fitted</th>
</tr>
</thead>
<tbody>
<tr>
<td>( k_{32}(0) ), (( \mu )M (^{-1}) s (^{-1}))</td>
<td>14</td>
<td>13.9 ± 2.0</td>
</tr>
<tr>
<td>( k_{32} ), (s (^{-1}))</td>
<td>95</td>
<td>92 ± 20</td>
</tr>
<tr>
<td>( k_{33}(0) ), (( \mu )M (^{-1}) s (^{-1}))</td>
<td>22</td>
<td>18.6 ± 1.0</td>
</tr>
<tr>
<td>( k_{33} ), (s (^{-1}))</td>
<td>1192</td>
<td>890 ± 38</td>
</tr>
<tr>
<td>( k_{34}(0) ), (( \mu )M (^{-1}) s (^{-1}))</td>
<td>1145</td>
<td>978 ± 20</td>
</tr>
<tr>
<td>( k_{34} ), (s (^{-1}))</td>
<td>365</td>
<td>356 ± 4</td>
</tr>
<tr>
<td>( k_{35}(0) ), (s (^{-1}))</td>
<td>418</td>
<td>319 ± 11</td>
</tr>
<tr>
<td>( k_{35} ), (s (^{-1}))</td>
<td>5335</td>
<td>4793 ± 208</td>
</tr>
</tbody>
</table>

Simulation values are the rate constants used to simulate single channel currents. Fitted values are the values of rate constants obtained from maximum likelihood analysis of simulated data, using Model 1. Missed events were corrected by imposing a dead time of 0.05 ms.

Single channel currents activated by 5, 10, 15, 20, 30, and 50 \( \mu \)M ATP were used for global fitting. The rate constants near the open states were very close between the two models. Missed events were corrected by imposing a dead time of 0.05 ms.
mental data than that from Model 1-2 (Fig. 13), again suggesting that Model 1-4 is better than Model 1-2.

The dependence of $P_o$ and rate constants on membrane potential. We next tried to determine whether the rate constants were dependent on membrane potential using Model 1-4 (Fig. 13). Fig. 15 A shows the single channel currents activated by 30 $\mu$M ATP at voltages from $-120$ to $-80$ mV. Fig. 15 B shows the voltage dependence of the mean open and closed times obtained from idealized currents, and Fig. 15 C shows $P_o$ as a function of voltage. The mean open time decreased with depolarization, while the mean closed time increased. The closing and opening rates are both voltage dependent, and the overall effect is to reduce the open probability with depolarization. $P_o$ values calculated from the all-points histogram were slightly larger than those calculated from the idealized currents, suggesting that some short lived events were missed, but the trend was the same; i.e., $P_o$ decreased with depolarization.

Figure 13. Simplified and expanded versions of Model 1 from Fig. 11 (1-1, 1-2, 1-3, and 1-4), and other kinetic models (9 and 8-1) that have been used in the literature.
To examine which rate constants vary with voltage, we globally fit the data between -80 and -120 mV with Model 1-4 (Fig. 13). Each rate constant was taken to be of the form:

$$k_{ij}(v) = k_{ij}(0) \exp(-z\delta_i V / k_B T), \quad (14)$$

where $k_{ij}(0)$ is the rate constant at 0 mV, $k_B T$ is Boltzmann’s constant times absolute temperature, and $z\delta_i$ is the effective charge (in a lumped parameter model, a product of the sensing charge and the fraction of the total electric field felt at the location of the sensor). However, this model has many parameters and did not converge [with a detailed balance constraint in loop, there are 30 parameters, including 15 $k_{ij}(0)$ and 15 $z\delta_i$]. We had to apply further constraints to reduce the number of parameters. Since it is presumed that the ATP binding site is located in the extracellular loop (Brake et al., 1994), it is reasonable to assume that it is outside the electric field, and therefore the association and dissociation rates are voltage independent. We fixed them to the values obtained by global fitting based on Model 1-4 (Table VII). The likelihood estimator converged, but with large error limits for $k_{ij}$ and $z\delta_i$ (Table IX). Therefore, we could not make a firm conclusion regarding the voltage dependence for any individual rate constant. However, the predicted $P_o$ using the mean

---

**Figure 14.** Representation of the channel activation pathway in terms of energy barriers and wells based on Model 1-4 (Fig. 13). The free energy landscape of the reaction scheme calculated from the transition rates. The relation between $k_B$ and free energy is defined by the Eyring equation:

$$k_{ij} = \kappa \frac{k_B T}{h} \exp \left( -\frac{G_i - G_j}{k_B T} \right),$$

($k_B$ in this case are the rate constants at -120 mV). $\kappa$ is the transmission coefficient (assumed to be 1; Hille, 1992), $k_B$ is Boltzmann’s constant, $h$ is Planck’s constant, and $T$ is the absolute temperature. At 20°C, $k_B T / h$ equals $6.61 \times 10^{3}$ s$^{-1}$. $G_{ij}$ is the free energy at the top of the barrier between states $i$ and $j$, and $G_i$ is the free energy of state $i$. The free energies are arbitrarily referenced to a solution of 1 M ATP. The use of $k_B T / h$ as the preexponential term of the rates is undoubtedly far off for a macromolecule. However, it is a maximum estimate that will cause the energy barriers to also be maximal estimates. The relationship of the well (state) energies, however, is much more likely to be correct since these energies are determined by ratios of rate constants where the preexponential terms will tend to cancel.

---

The table VI shows the rate constants obtained from global fitting based on Models 1-3 and 1-4 (Fig. 13).

<table>
<thead>
<tr>
<th>Table VII: Rate Constants Obtained from Global Fitting Based on Models 1-3 and 1-4 (Fig. 13)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Model 1-3</strong></td>
</tr>
<tr>
<td>Rate constant</td>
</tr>
<tr>
<td>$k_{12}(0)$, (µM$^{-1}$s$^{-1}$)</td>
</tr>
<tr>
<td>$k_{21}$, (s$^{-1}$)</td>
</tr>
<tr>
<td>$k_{32}(0)$, (µM$^{-1}$s$^{-1}$)</td>
</tr>
<tr>
<td>$k_{31}$, (s$^{-1}$)</td>
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<td>$k_{34}$, (s$^{-1}$)</td>
</tr>
<tr>
<td>$k_{43}$, (s$^{-1}$)</td>
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<td>$k_{53}$, (s$^{-1}$)</td>
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<td>$k_{65}$, (s$^{-1}$)</td>
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<td>$k_{64}$, (s$^{-1}$)</td>
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<tr>
<td>$k_{65}$, (s$^{-1}$)</td>
</tr>
<tr>
<td>$k_{66}$, (s$^{-1}$)</td>
</tr>
</tbody>
</table>

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The table VIII shows the comparison of expanded versions of Model 1 (Fig. 11).

<table>
<thead>
<tr>
<th>Table VIII: Comparison of Expanded Versions of Model 1 (Fig. 11)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Model</strong></td>
</tr>
<tr>
<td>1-4</td>
</tr>
<tr>
<td>1-3</td>
</tr>
<tr>
<td>1-2</td>
</tr>
<tr>
<td>1</td>
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</tbody>
</table>

Single channel currents activated by 5, 10, 15, 20, 30, and 50 µM ATP were used for the calculations of maximal likelihoods and AIC ranks. All data were from the same patch. The lowest number of AIC rank is considered best. Detailed balance constraint was applied in the loop of Models 1-3 and 1-4. Missed events were corrected by imposing a dead time of 0.05 ms. (Models 1-2, 1-3, and 1-4 are from Fig. 13.)
values of rate constants does decrease with depolarization and is similar to the $P_o$ from idealized currents (Fig. 15 C). Fig. 15 D shows the open and closed interval histogram and the predicted probability densities (solid lines) from the rate constants.

We predicted the shape of the whole cell I–V relationship by combining $P_o$ from outside-out patches and the single channel conductance. The whole cell current $I$ is determined by the product of single channel current, $i$, the number of channels, $n$, and the open probability, $P_o$:

\[ I = n i P_o. \]  

$P_o$ obtained from histograms can be fit with the Boltzmann equation:

\[ P_o(V) = P_o(0) \exp\left(-\frac{z\delta V}{k_B T}\right), \]

where $P_o(0)$ is $P_o$ at 0 mV, and is equal to 0.11. $z\delta$ is equal to 0.34, indicating that a hyperpolarization of 74 mV is needed for an e-fold increase of $P_o$. If we presume that the voltage dependence of $P_o$ at different ATP concentrations is the same, we can use this result, together with the dose–response curve (Fig. 4 C), to estimate $P_o(V)$ of a single channel. Multiplying $P_o(V)$ by the single channel current (Fig. 2) predicts the shape.
of the whole-cell I–V relationship. It is close to that predicted by Fig. 13, Model 1-4 (Fig. 3 B).

discussion

In this study, we have characterized the single channel properties of cloned P2X2 receptor ion channels. The characterization included general gating features, permeation properties, ATP concentration dependence, effects of pH, and kinetic analysis.

Single Channel Current Behavior and Excess Open Channel Noise

The typical single channel Na\(^+\) current has a chord conductance of \(~30\) pS at \(-100\) mV (Fig. 1). The open channel current shows high frequency, high amplitude flickering with some apparent full closures. Because of the difficulty in resolving the fluctuations comprising this “buzz mode,” rather than build a substrate model to characterize the open channel behavior, we characterized it as a single conductance with noise. The standard deviation of the excess open channel current is \(~30\)% of the mean. This is much larger than that of the acetylcholine receptor, for example, where the noise is only 2–5% of the mean (Auerbach and Sachs, 1984; Sigworth, 1985, 1986). The excess noise does not arise from thermal or shot noise nor from the voltage noise of the amplifier (in the worst case, this is 10^{-8} \times the thermal noise of the channel; Sigworth, 1985). The fluctuations appear to represent rapid conformational changes that modulate the open channel conductance. Occasionally, we saw relatively long-lived subconductance levels (see Fig. 1 D), but these were too infrequent to be evident in the all-points histograms. We attempted to estimate whether the fluctuations were to a discrete number of conducting states using a maximum-point likelihood approach. However, we could not find a consistent set of substate amplitudes between records from different patches, although it is clear that the flickers do not represent simple band-limited full closures of the channel. It is possible that there are actually a large number of states better described by a noise rather than a state model. The presence of these rapid fluctuations means that attempts to estimate the unitary channel current with noise analysis are prone to large, bandwidth-dependent errors.

We do not think that the flickers arise from channel block by a diffusible agent, a mechanism that is often seen in other channels. Ca\(^2+\)-activated K\(^+\) channels can be blocked in a flickery manner by Na\(^+\) (Yellen, 1984), and cardiac Ca\(^2+\) channels are discretely blocked by divalent ions (Lansman et al., 1986). Since our currents were equally noisy with (Fig. 1 A) and without (Figs. 4 A and 5) extracellular divalent ions, we do not believe that the excess noise comes from the block of divalent ions. ATP is not a candidate for blocking the channels since the mean amplitude and the excess noise are independent of ATP concentration (Fig. 4 A). The voltage dependence of the excess noise is also not significant, suggesting that the flickers do not involve processes that sense the electric field. The simplest interpretation is that the excess noise arises from conformational transitions of the channel itself.

The general features of the cloned P2X2 receptors we have discussed are similar to data recorded from native receptors in rat sensory neurons and PC12 cells (Krishnamoorthy et al., 1988; Nakazawa and Hess, 1993). In rat dorsal root ganglion (DRG) cells, single channel currents flickered much more rapidly than in PC12 cells—so rapidly that the lifetime of both states was almost always too short to be resolved by the recording system (Bean et al., 1990). The mean amplitude of the open state with 150 mM extracellular Na\(^+\) was only 0.9 pA at \(-130\) mV with an equivalent chord conductance of 7 pS and no obvious substrates. In contrast, we observed a mean single channel current of 3.2 pA, much larger than in the rat DRG cells, with 145 mM extracellular Na\(^+\) at \(-100\) mV with an equivalent mean chord conductance of 32 pS. The true maximum conductance is even larger since the difference in amplitude of the upper 5% of closed and open distributions is 4.3 pA, \(~34\)% larger than the mean current.

I–V Relationship

The single channel I–V relationship of the cloned P2X2 receptors exhibited strong inward rectification (Fig. 2). This result is consistent with the whole cell I–V relationship (Fig. 3) (Brake et al., 1994; Valera et al., 1994). Zhou and Hume (1998) studied the mechanisms of inward rectification of P2X2 receptors. In their data, both

<table>
<thead>
<tr>
<th>Table IX</th>
<th>Voltage Dependence of Rate Constants Based on Model 1-4 (Fig. 13)</th>
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<tr>
<td></td>
<td>(k_i(-120))</td>
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<tr>
<td>(C_i\rightarrow O_j)</td>
<td>(s^{-1})</td>
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<tr>
<td>(C_1\rightarrow O_3)</td>
<td>344</td>
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<tr>
<td>(O_2\rightarrow C_4)</td>
<td>1150</td>
</tr>
<tr>
<td>(C_4\rightarrow O_6)</td>
<td>330</td>
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<tr>
<td>(O_4\rightarrow C_7)</td>
<td>2166</td>
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<tr>
<td>(O_3\rightarrow C_7)</td>
<td>27</td>
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<tr>
<td>(C_7\rightarrow O_5)</td>
<td>261</td>
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<tr>
<td>(O_5\rightarrow C_7)</td>
<td>214</td>
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<tr>
<td>(C_7\rightarrow O_6)</td>
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gating and single channel conductance contributed to the inward rectification. They also reported that inward rectification did not require intracellular Mg$^{2+}$ or polyamines, and was present when the same solution was used on both sides of the patch. Our data supports these results. The currents in Fig. 2 were recorded in the presence of 1 mM extracellular Ca$^{2+}$ and Mg$^{2+}$; however, the single channel current I–V relation showed similar inward rectification when currents were recorded in the absence of divalent cations (data not shown).

Since the mean open and closed times vary with voltage (Fig. 15 B), the opening and closing rate constants are voltage dependent. Although our kinetic analysis was unable to make a firm assignment of the voltage dependence to particular rate constants, $P_o$ did decrease with depolarization (Fig. 15 C). The predicted whole cell currents (Fig. 3 B) based on Model 1-4 (Fig. 13) and single channel I–V curve matched reasonably well with the data. These results also suggest that both the instantaneous conductance and voltage-dependent gating contribute to the inward rectification. The dual mechanisms of inward rectification in this receptor are similar to the neuronal nicotinic acetylcholine receptor (Sands and Barish, 1992). The most important feature of the voltage dependence of P2X$_2$ kinetics is that it is minor.

**Probability of Being Open**

ATP is a potent agonist for cloned P2X receptors, except P2X$_7$, and the receptors are highly selective for ATP over most other adenosine derivatives. Dose–response studies of whole cell currents reveal a Hill coefficient larger than 1 (Brake et al., 1994), suggesting that activation requires more than one agonist. This is reasonable since the channels are composed of multiple homomeric subunits. We studied $P_o$ over a wide range of ATP concentrations with outside-out patches (Fig. 4) and showed that the $P_o$ curve has a Hill coefficient of 2.3, an $EC_{50}$ of 11.2 $\mu$M, and a maximum of 0.61. A Hill coefficient of 2.3 suggests that there are at least three binding sites in the receptor. (That the Hill coefficient is only a lower estimate of the stoichiometry is emphasized in Fig. 4 C, where data from a simulation of Model 1-4 [Fig. 13] with three binding steps could be well fit with a Hill coefficient of 1.8.) Presumably, the cooperativity arises from the multimeric structure of the channel. Based on refolding studies of the P2X$_2$ extracellular domain (P2X$_2$-ECD), Kim et al. (1997) predicted that the naturally assembled form of P2X$_2$ receptors may be tetrameric. Lewis et al. (1995) found that coexpression of P2X$_2$ and P2X$_3$ can form a new channel type by subunit heteropolymerization, providing further evidence that the P2X receptors are multimers. Recent experiments with chemical cross-linking of P2X$_1$ and P2X$_3$ receptors (Nicke et al., 1998) indicate that P2X receptor channels are trimeric. Since these results were obtained from native P2X receptors expressed in *Xenopus* oocytes, we expect that they are more representative than the studies on the isolated extracellular domains.

The maximum $P_o$ of -0.6 indicates that the mean opening rates are slower than the closing rates. Our kinetic analysis based on Model 1-4 (Fig. 13) shows that the two opening rates $k_{46}$ and $k_{76}$ are much slower than corresponding closing rates $k_{54}$ and $k_{67}$. The opening rate $k_{46}$ is faster than closing rate $k_{45}$, while the opening rate $k_{75}$ is similar to the closing rate $k_{57}$, so the overall opening rate is slower than the closing rate.

**Affinity of the Pore for Na$^+$**

The theory of independent ion passage predicts that the flux of a permeating ion should increase linearly with the ion concentration (Hille, 1992). However, most channels do not exhibit this behavior due to the competition for binding sites in the channel. Ion flux saturates when the binding–unbinding steps of permeation become rate limiting. This occurs at high ion concentrations when the rate of ion entry approaches the rate of unbinding. Conductances in the P2X$_2$ channels show clear deviations from independence (Fig. 6 A). When the concentration of extracellular NaCl is raised, the single channel conductance saturates. In our data, the mean conductance versus Na$^+$ concentration was well fit by the Michaelis–Menten (MM) equation with one binding site, X, in the pore (Scheme II).

$$
\text{X+Na}^{+\text{out}} \xrightarrow{k_1} \text{X-Na}^{+} \xrightarrow{k_2} \text{X+Na}^{+\text{in}}
$$

**(SCHEME II)**

The rate constants are, in general, dependent on membrane potential. We tried to fit our conductance data with Scheme II, but could not obtain a unique set of rate constants: Scheme II is over determined because the data does not have enough distinguishing features. If we assume that at high potentials the reverse flux is negligible, only the two forward rates are necessary and we can obtain solutions. We found that the equilibrium constant $K_s$ is voltage dependent (Fig. 6 B), with the binding site located -20% of the way through the field relative to the extracellular face. It is interesting to speculate as to where the site may be relative to the primary sequence if one assumes that side chains form the selectivity filter rather than the backbone carbonyls (Doyle et al., 1998). In their study of the ionic pores of P2X$_2$ receptors using the substituted cysteine accessibility method (SCAM), Rassendren et al. (1997a) identified three residues: 1328, N333, and T336 in the M2 domain that were located in the outer
vestibule of the pore. Two of these are polar and might be part of a binding site for Na\(^+\). When the channel was open, D349C could be inhibited only by the small, positively charged MTSEA (2-aminoethyl-methanethiosulfonate), but not by MTSET \(\{2\text{-}(\text{trimethylammonium})\text{-}\text{ethyl}\text{-}\text{methanethiosulfonate}\}\) or MTSES (sodium (2-sulfonatoethyl)methanethiosulfonate), implying that D349 is located near the middle of the channel. D349 is a negatively charged amino acid and is conserved among all seven P2X receptors. It is possible that D349 could be the site of permeant cation binding and is responsible for ionic selectivity.

Cation Selectivity

Our data shows P2X\(_2\) is a nonselective cation channel. The ionic selectivity based on the conductance is: K\(^+\) > Rb\(^+\) > Cs\(^+\) > Na\(^+\) > Li\(^+\), Eisenman sequence IV (Hille, 1992). This sequence is different from free solution mobility and from the sequence of high field sites. This suggests the pore may be smaller than the nicotinic acetylcholine receptor with an interior having little charge in the selectivity filter. This is consistent with the results from substituted cysteine accessibility method experiments (Rassendren et al., 1997a) where only I328C, N333C, T336C, L338C, and D349C in the M2 domain were accessible to MTS reagents. Only D349 is negatively charged among these residues, and it may not be part of the selectivity filter.

Based on whole cell currents of P2X\(_2\) receptors, Brake et al. (1994) reported that the replacement of extracellular Na\(^+\) by K\(^+\) did not affect the reversal potential, suggesting that Na\(^+\) and K\(^+\) have a similar permeability near 0 mV. In our experiments, the currents carried by Na\(^+\) were larger than the currents carried by K\(^+\) at negative potentials. Similar results were reported for PC12 cells (Nakazawa et al., 1990b). The origin of the discrepancy between our results and theirs is masked by the lack of knowledge of the interplay between permeation and gating in the whole cell current. Different ionic environments may change the agonist binding and/or gating. In the nicotinic acetylcholine receptor (Akk and Auerbach, 1996), external monovalent ions compete with agonists for binding, changing the dose-response curves for reasons that have nothing to do with the permeation process itself. If the kinetics of ATP binding and/or gating is different for Na\(^+\) and K\(^+\) in the extracellular solution, the Hill coefficient of 2.5 suggests that the channel has at least three binding sites, which is consistent with the stoichiometry study by Nicke et al. (1998). In related experiments, extracellular protons potentiated adenosine binding to A\(_2A\) receptors, and this effect could be modified by mutagenesis or by chemically altering the strategic residues (Allende et al., 1993). In the extracellular loop of P2X\(_2\) receptors, there are 9 histidine residues interspersed between 10 cysteine residues, the latter being conserved throughout the P2X\(_1\)–7 proteins. Both cysteine and histidine residues have been shown to be important for agonist and antagonist binding at the A\(_1\) receptor, which is pH sensitive (Allende et al., 1993). It is reasonable to speculate that these two amino acids may play a similar role in ATP binding to P2X\(_2\) receptors. Protonation of the histidine residues may account for the increase in P2X\(_2\) current at low pH, but this seems unlikely because diethylpyrocarbonate, which irreversibly denatures histidyl residues, has no effect on the magnitude of the currents (Wildman et al., 1998).

The major effect of pH was on the kinetics of activation. The rate of activation increased as pH decreased (Fig. 8 C), while the deactivation time constant was independent of pH. This suggests that the closing rates and dissociation rates are not affected by protons. The simplest interpretation of the data is that in acidic environments, the binding site becomes more positive, increasing its affinity for ATP. However, since macroscopic kinetics is a function of all of the rate constants, many of which are not associated with binding, such an interpretation is not reliable.

pH Potentiation

The sensitivity of cloned P2X\(_2\) receptors to ATP was affected by extracellular pH. King reported (King et al., 1996) that with acidification, the ATP dose–response curve of whole cell currents shifted to the left without altering the maximal response. The effective receptor affinity for ATP was enhanced 5–10-fold by acidifying the bath solution (to pH 6.5), but was diminished four- to fivefold in an alkaline solution (pH 8.0). Different P2X receptors have different sensitivities to pH. Unlike P2X\(_2\) receptors, P2X\(_3\), P2X\(_5\), and P2X\(_4\) receptors decrease their apparent affinity with acidification (Stoop et al., 1997). Our studies on outside-out patches showed that the mean current increased about an order of magnitude when the extracellular pH changed from 8.3 to 6.8, exhibiting a pK\(_a\) of ~7.9 (Fig. 8 B). The Hill coefficient of 2.5 suggests that the channel has at least three binding sites, which is consistent with the stoichiometry study by Nicke et al. (1998). In related experiments, extracellular protons potentiated adenosine binding to A\(_2A\) receptors, and this effect could be modified by mutagenesis or by chemically altering the strategic residues (Allende et al., 1993). In the extracellular loop of P2X\(_2\) receptors, there are 9 histidine residues interspersed between 10 cysteine residues, the latter being conserved throughout the P2X\(_1\)–7 proteins. Both cysteine and histidine residues have been shown to be important for agonist and antagonist binding at the A\(_1\) receptor, which is pH sensitive (Allende et al., 1993). It is reasonable to speculate that these two amino acids may play a similar role in ATP binding to P2X\(_2\) receptors. Protonation of the histidine residues may account for the increase in P2X\(_2\) current at low pH, but this seems unlikely because diethylpyrocarbonate, which irreversibly denatures histidyl residues, has no effect on the magnitude of the currents (Wildman et al., 1998).

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The single channel current amplitudes at different pH were similar, but the excess open channel noise increased when pH was lowered (Table II). Comparing the single channel currents at different values of pH, as shown in Fig. 9 A, more brief closings can be seen at lower pH. The fluctuations caused by protons are slower than the fluctuations of the intrinsic channel flicker. While these results suggest that protons served as blockers, analysis of the power spectra and single channel kinetic models contradict this interpretation. The blocking and unblocking rates were only weakly dependent on proton concentration (see Eq. 13). Power spectral analysis also showed that the corner frequency decreased with an increase in pH, opposite to the prediction for proton block. It appears that the brief closings at low pH are due to conformational changes produced by the titration of several sites.

**Kinetics**

**Preferred model.** We used the maximum interval likelihood method to statistically compare kinetic models (Vandenberg and Horn, 1984; Horn, 1987; Qin et al., 1996a). The models were built hierarchically, beginning with one that described the transitions near the open states (Vandenberg and Bezanilla, 1991). The first model we chose (Model 1 or, equivalently, Model 2; Figure 11) had three closed states, C1, C2, and C3, representing unliganded, monoliganded, and biliganded closed states, and two open states O5 and O6.

Starting with this model, we expanded to Models 1-2, 1-3, and 1-4 (Fig. 13) to account for the ATP titration data. Comparing Model 1 (Fig. 11) with Model 1-2, and Model 1-3 with Model 1-4, we found that the opening and closing rate constants were surprisingly close (see Tables VI and VII), suggesting that the transitions near the open state were well defined. This result supports our strategy of model development. Among the four models, Model 1-4 had the highest likelihood and AIC rank, and therefore is our preferred model (Table VII). In this model, there are five closed states, C1, C2, C3, C4, and C5, representing unliganded, monoliganded, biliganded, and triliganded closed states, and two open states O5 and O6. The three ATP binding steps require the channel to be at least a trimer. By constraining the ATP association rates to be proportional to concentration, reliable rate constants were obtained from global fitting with this model (Table VII). The predicted probability density functions match reasonably well the open- and closed-time histograms at all concentrations, and the predicted $P_o$ is close to the experimental data so that our final preferred model is Model 1-4.

ATP binding sites and cooperativity. From the results of the basic model (Fig. 11, Model 1) and its expanded versions (Fig. 13, Models 1-2, 1-3, and 1-4), we found that association and dissociation rate constants increased as they approached the open states (Table VI and VII). This means that the subunits are not independent and that the association rate for an incoming ATP is increased by the presence of bound ATPs. This trend was consistent among all models examined.

The increase in association rates with consecutive binding is most surprising when one thinks about the opposing electrostatic factors. Bound ATP with a charge of $-4$ should repel the next incoming ATP (note: the actual valence of bound ATP is not known and is probably less than $-4$). Since the energy is proportional to the product of the charges, if the sites were independent, the second ATP would be repelled by a resident ATP with an energy proportional to $4^2 = 16$, and the third by $8^2 = 32$. We would thus expect the later binding rates to decrease by more than just the number of available sites. The trend of increasing association rates with occupancy must be caused by conformational changes in the channel that mask the electrostatic contribution. While the increasing rates of dissociation with occupancy fit the predicted electrostatic trend, given the conformational changes associated with binding, this trend may be coincidental. The binding of ATP causes the remaining unoccupied sites to open up in such a manner so as to increase the rates of ATP entry and exit without having a large effect on the equilibrium affinity of the site (Fig. 14).

If the channel were a tetramer, as predicted by Kim et al. (1997), we might add one closed state to the left of C1, as in Model 1-5 (Fig. 13). We attempted to fit this model, but could not obtain unique rates. This may be because we had insufficient data at very low ATP concentrations, or because the channel really is a trimer (Nicke et al., 1998). The predicted $P_o$ vs. ATP concentration from Model 1-4 (Fig. 13) fits well the $P_o$, $EC_{50}$, and Hill coefficient obtained from the dose–response curve (Fig. 4 C), supporting the consistency of the model and the necessity of the last ATP-independent closed state.

**Other kinetic models.** There have been only a few studies on the kinetics of P2X receptor ion channels based on single channel currents (Krištál et al., 1988). Kinetic studies using whole cell currents showed that the rise time of current elicited by ATP was strongly concentration dependent, but the decay time was not (Surprenant, 1996). This is in accord with the pH experiments discussed above. There is some variability among reports regarding the rise and fall times. In rat sensory neurons, using fast solution exchange, the rise time was $\sim 10$ ms at saturating ATP and the decay time was $\sim 100$ ms (Bean et al., 1990). In smooth muscle and cloned P2X$_2$ receptors, the rise time was $\sim 5$ ms, while in PC12 cells, cloned P2X$_2$ receptors, and rat superior cervical ganglion (SCG) cells, the rise time was $\sim 25$ ms (Surprenant, 1996). In rat SCG cells, nodose, and guinea-
pig coeliac neurons, the latency to the onset of whole cell currents was estimated to be \( \sim 0.8-4 \) ms, and the 10-90% rise time at high ATP concentrations ranged from 5 to 20 ms (Khakh et al., 1995). Since these experiments were done in the whole cell configuration, the rate of rise was likely limited by the speed of the solution exchange around the cells. Hess (1993) reported that the time resolution for solution exchange around a whole cell of this size is \( 2-10 \) ms under maximal flow velocities and even slower for lower velocities. With outside-out patches, the rise times can be \( \sim 250 \) \( \mu \)s (Colquhoun et al., 1992). Moreover, because in the whole cell experiments there was no marker for the start of solution exchange, the latency to the onset of the currents could not be estimated accurately.

Preliminary kinetic analysis of single channel currents from rat sensory neurons showed that the distribution of the open times could be approximately fit by two exponentials with time constants of 0.35 and 3.4 ms (Krishtal et al., 1988). The ratio between the fast component amplitude and the slow one varied between patches, with ratios of 47.4:1 to 4:8:1.

Using the rise and fall time constants from whole cell currents at different concentrations of ATP, and based on kinetic models of the ACh receptor channel, Bean (1990) proposed a linear kinetic model with independent subunits for ATP activation (Fig. 13, Model 9). The association and dissociation rate constants \( k_+ = 1.2 \times 10^7 \text{M}^{-1}\text{s}^{-1} \) and \( k_- = 4 \text{s}^{-1} \) were chosen so that the simulations mimicked the kinetics seen in the bullfrog sensory neurons. Agreement between the model and the data suggested that the ATP binding sites could be independent.

We applied the independent binding site assumption to Model 1-2; i.e., \( k_{31} = 1/2k_{23} = 1/3k_{13} = 1/3k_{10} = k_+ \), and \( k_{21} = 1/2k_{32} = 1/3k_{34} = k_- \). Although the estimation converged, the likelihood was much lower than our favored models. We also fit our data with Bean’s (1990) model, which has only one open state (Fig. 13, Model 9), and again the fits were poor.

In other ligand-gated channels, such as the Ca2+ activated potassium channel and ACh receptors (Magleby and Song, 1992; Auerbach et al., 1996), partially liganded channels can open. To explore whether these states were visible in our data, we compared the maximum likelihood of Model 8-1 (Fig. 13) that has partially liganded openings with Model 1-2 (Fig. 13). Global fitting of Model 8-1 with data from 5, 10, 15, 20, 30, and 50 \( \mu \)M ATP produced a maximum likelihood 664 units lower than Model 1-2, suggesting the model is \( e^{664} \), less likely to produce the data. Apparently, P2X2 channels do not open for a significant fraction of time in partially liganded states.

In summary, Model 1-4 (Fig. 13) can well represent the channel gating processes. It is adequate to explain the behavior across concentration and is physically reasonable. Fig. 14 shows the calculated free energy barriers and wells at \(-120\) mV referenced to 1 M ATP. The energies were calculated from the rate constants assuming an Eyring model. The free energies of all wells decreased with the reaction coordinate. State \( C_4 \) could go to either open state \( O_5 \) or \( O_6 \) with state \( O_5 \) being slightly more stable. \( O_3 \) and \( O_5 \) go to the same closed state \( C_7 \) that is the most stable in the reaction pathway.

Voltage dependence of \( P_o \) and rate constants. Although \( P_o \) is not strongly dependent on membrane potential, it did decrease with depolarization (Fig. 15 C), indicating that some of the rate constants are voltage dependent. To reduce the free parameters, we limited the voltage sensitivity to only the rate constants in the final loop and ended with very large error limits for \( k_{ij}(0) \) and \( z_{ij} \). We have no confidence in the voltage dependence of any of the individual rate constants. Data from a wider voltage range will be required to adequately address this question.

Despite the wide error limits, by lumping the kinetics into an equilibrium model to predict the probabilities of occupancy, the predicted \( P_o \) and whole cell currents calculated from the mean values of \( k_{ij}(0) \) and \( z_{ij} \) were consistent with those obtained by idealization (Fig. 15 C) and with experimental data (Fig. 3 B). The probability density functions match reasonably well with duration histograms (Fig. 15 D).

In conclusion, the currently optimal model, Model 1-4 (Fig. 13), can be summarized as follows: (a) the channel proceeds through three ATP binding steps before opening; (b) the three ATP binding sites are not independent, but positively cooperative; (c) there are two open states, which connect to a common ATP-independent closed state; (d) activation and deactivation proceed along the same pathway; and (e) channels only open after being fully liganded.

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