

# On the use of time constants as estimates of mean lifetimes in single channel data analysis

R. O. Edeson<sup>1</sup>, R. K. Milne<sup>2</sup>, G. F. Yeo<sup>4,5</sup>, and B. W. Madsen<sup>3</sup>

<sup>1</sup> Department of Anaesthesia, Sir Charles Gairdner Hospital, Nedlands, 6009, Western Australia

<sup>2</sup> Department of Mathematics and <sup>3</sup> Department of Pharmacology, University of Western Australia, Nedlands, 6009, Western Australia

<sup>4</sup> School of Mathematical and Physical Sciences, Murdoch University, Murdoch, 6150, Western Australia

<sup>5</sup> Department of Mathematics and Computer Science, Odense University, DK-5230 Odense M, Denmark

Received July 19, 1989/Accepted in revised form January 2, 1990

**Abstract.** For Markov models of single channel kinetics, a sojourn time in a class of states has a density function which is usually a linear combination of exponential densities. There are many instances in the single channel literature where the time constants of exponentials fitted to sojourn time data have been used as estimated mean sojourn times in individual states, though the two may be very different. In the present study the nature and magnitude of this difference in the case of a two state class is illustrated analytically and numerically. The time constants should be viewed at best as approximations, possibly poor, to the estimated mean sojourn times. Estimates of kinetic parameters cannot in general be obtained explicitly from the fitted parameters of the density alone. However, this is shown to be possible in some special cases and enables direct estimation of, for example, the channel opening rate constant  $\beta$  (or an upper limit to the estimate of  $\beta$  in the case of multiple channels) in standard sequential three or four state models of nicotinic receptor kinetics, using only the fitted parameters of the closed-time density.

**Key words:** Channel kinetics – Markov models – Sojourn times – Approximation – Eigenvalues

## Introduction

In single channel data analysis, sojourn time data are usually fitted by a probability density function of a form determined by model assumptions. If the underlying channel kinetics are modelled by a Markov process satisfying mild conditions (see, for example, Colquhoun and Hawkes 1983 and Appendix 2 of Milne et al. 1988), the sojourn time (lifetime) density for a class of states is a linear combination of exponential densities, with the number of component densities equal to the number of kinetically distinct states in the class. The parameters

(rate constants, reciprocal time constants) of the exponential densities correspond to the eigenvalues of a matrix whose diagonal elements are reciprocal mean sojourn times in individual states. It is well established (as in Colquhoun and Hawkes 1983, p. 174; Fredkin et al. 1985; and others) that the eigenvalues and diagonal elements are usually not identical, but the detailed relationship between the two in the single channel context has received little attention.

In practical single channel inference, the fitting of sojourn time data gives estimates of the rate constants of component exponential densities, whereas it is estimates of the reciprocal mean lifetimes that are usually required for interpretation of the data in terms of a proposed kinetic model. An assumption that fitted rate constants are equal or, implicitly, good approximations to estimated reciprocal mean lifetimes is sufficiently prevalent (see for example Sine and Steinbach 1984, 1986; Auerbach and Lingle 1987; Igusa and Kidokoro 1987; Hestrin et al. 1987; Bormann 1988; Papke et al. 1988; Jaramillo and Schuetze 1988; Sikdar et al. 1989) to warrant examination, since estimation of kinetic parameters based upon this assumption will usually be biased.

Here we discuss the theoretical relationship between eigenvalues and reciprocal mean lifetimes, and the errors that can arise when fitted time constants are used directly as estimates of mean lifetimes. For illustrative purposes, attention is restricted to a class containing just two states, that is, where the corresponding sojourn time density is bi-exponential. We show that although the approximations may often be adequate, they can under some conditions be very poor and lead to substantial errors with possible misinterpretation of data. Consequently, consideration should be given to the adequacy of such approximations if they are used (see for example Colquhoun and Sakmann 1985). In special cases, as we illustrate below, explicit estimation of kinetic parameters is possible, thus obviating the need for approximation. Throughout the following, the unit of time is arbitrary.

### Theory: eigenvalues and reciprocal mean lifetimes

As in Colquhoun and Hawkes (1982); Fredkin and Rice (1986); Milne et al. (1988); or Ball and Sansom (1988), the following development is based upon a finite state continuous-time homogeneous Markov process in equilibrium, where all states communicate. This can be represented by a transition rate matrix  $\mathbf{Q}=[q_{ij}]$ , where  $q_{ij} \geq 0$  ( $i \neq j$ ) is the transition rate from state  $i$  to state  $j$  and  $d_i = -q_{ii} = \sum_{j \neq i} q_{ij}$ .

We are concerned with sojourn times in a class  $\mathcal{C} = \{1, 2\}$  of two neighbouring states for which all possible connections with other states are allowable. This is represented as follows:

$$\rightleftharpoons 1 \rightleftharpoons 2 \rightleftharpoons . \quad (\text{Scheme 1})$$

For  $\mathcal{C}$ , the submatrix of interest is

$$-\mathbf{Q}_{\mathcal{C}\mathcal{C}} = \begin{bmatrix} d_1 & -q_{12} \\ -q_{21} & d_2 \end{bmatrix} \quad (1)$$

with states 1 and 2 having finite mean sojourn times  $d_1^{-1}$  and  $d_2^{-1}$  respectively. Without loss of generality the states are labelled so that  $d_1 \leq d_2$ .

The (real) eigenvalues of  $-\mathbf{Q}_{\mathcal{C}\mathcal{C}}$  are given by

$$\lambda_1, \lambda_2 = \frac{1}{2} \{d_1 + d_2 \mp [(d_1 + d_2)^2 - 4d_1 d_2 s]^{1/2}\} \quad (2)$$

where the subscripts are chosen so that  $\lambda_1 \leq \lambda_2$ , and

$$s = 1 - p_{12} p_{21} , \quad (3)$$

where  $p_{ij} = q_{ij}/d_i$ ,  $j \neq i$ , is the probability that a transition from state  $i$  is to state  $j$ . Then  $s$  can be interpreted as the success probability (that is, the probability of terminating a visit to  $\mathcal{C}$ ) in the geometric distribution of the number of visits to one of the states in  $\mathcal{C}$  during a sojourn in  $\mathcal{C}$  which begins and ends in the other state. Except in the trivial case when  $d_1 = d_2$  and  $s = 1$  (which implies from (2) that  $\lambda_1 = \lambda_2$ ), the density function of a sojourn time in  $\mathcal{C}$  is (from Colquhoun and Hawkes 1982, (1.32) and (1.33))

$$f(t) = C_1 \lambda_1 e^{-\lambda_1 t} + C_2 \lambda_2 e^{-\lambda_2 t} , \quad (4)$$

where

$$C_1 = \{\lambda_2 - \pi_1 (d_1 - q_{12}) - \pi_2 (d_2 - q_{21})\} / (\lambda_2 - \lambda_1) , \quad (5)$$

$C_2 = 1 - C_1$  (not necessarily greater than  $C_1$ ),  $\pi_1$  is the probability that a sojourn in  $\mathcal{C}$  starts in state 1, and  $\pi_2 = 1 - \pi_1$ .

It can be seen from (2) that

$$\lambda_1 + \lambda_2 = d_1 + d_2 , \quad \lambda_1 \lambda_2 = d_1 d_2 s , \quad (6)$$

and consequently

$$0 \leq \lambda_1 \leq d_1 \leq d_2 \leq \lambda_2 \leq d_1 + d_2 . \quad (7)$$

Clearly  $\lambda_1$  and  $\lambda_2$  depend not only on  $d_1$  and  $d_2$  but also on  $q_{12}$  and  $q_{21}$  through their product. Except when  $s = 0$ , in which case exit from  $\mathcal{C}$  is not possible, both  $\lambda_1$  and  $\lambda_2$  are strictly positive. They are distinct whenever  $s < 1$  (that is, when both  $q_{12}$  and  $q_{21}$  are positive). Furthermore,  $\lambda_1 = d_1$  and  $\lambda_2 = d_2$  if and only if  $s = 1$ . (In this last case, at least one of  $p_{12}$  and  $p_{21}$  is zero, and it is not possible to loop between the states during a given sojourn in  $\mathcal{C}$ ;

further, at least one type of sojourn in  $\mathcal{C}$  consists of a single sojourn in one of the two states. If both  $p_{12}$  and  $p_{21}$  are zero, the states cannot be neighbours.) When  $d_1$  is small so is  $\lambda_1$ , and  $\lambda_2$  is then close to  $d_2$ . The difference between an eigenvalue and a reciprocal mean sojourn time is given from (2) by

$$\varepsilon = d_1 - \lambda_1 = \lambda_2 - d_2 = \frac{1}{2} \{[(d_1 + d_2)^2 - 4d_1 d_2 s]^{1/2} - (d_2 - d_1)\} , \quad (8)$$

which decreases monotonically from  $d_1$  to zero as  $s$  increases from zero to 1. The behaviour of  $\lambda_1$  and  $\lambda_2$  as  $s$  varies is illustrated graphically in Fig. 1.

### Inference: approximation errors

In practice, the density (4) is fitted to  $\mathcal{C}$ -sojourn time data giving estimates  $\hat{\lambda}_1, \hat{\lambda}_2$  and  $\hat{C}_1$  of  $\lambda_1, \lambda_2$  and  $C_1$ . A formal approach to inference might involve estimation of the four parameters  $d_1, d_2, q_{12}$  and  $q_{21}$ ; in general four estimating equations would be needed (five if  $\pi_1$  in (5) were also included). A full likelihood method could be used when the successive sojourn times in  $\mathcal{C}$  are independent (and might serve as a useful approximation in the dependent case); though analytically complicated this could be implemented numerically. However, the approach in many published studies is less formal and attempts to recover information on  $d_1$  and  $d_2$  on the basis of the eigenvalue estimates  $\hat{\lambda}_1$  and  $\hat{\lambda}_2$  alone.

Let  $\hat{\lambda}_1 + \hat{\lambda}_2 = b$  and  $\hat{\lambda}_1 \hat{\lambda}_2 = c$ . Then (6) become

$$d_1 + d_2 = b , \quad d_1 d_2 = c/s , \quad (9)$$

and solving these gives estimates

$$\hat{d}_1, \hat{d}_2 = \frac{1}{2} [b \mp (b^2 - 4c/s)^{1/2}] , \quad (\hat{d}_1 \leq \hat{d}_2) \quad (10)$$

provided an estimate  $\hat{s}$  of  $s$  is available. For  $\hat{d}_1, \hat{d}_2$  to be real requires

$$\hat{s} \geq 4c/b^2 , \quad (11)$$

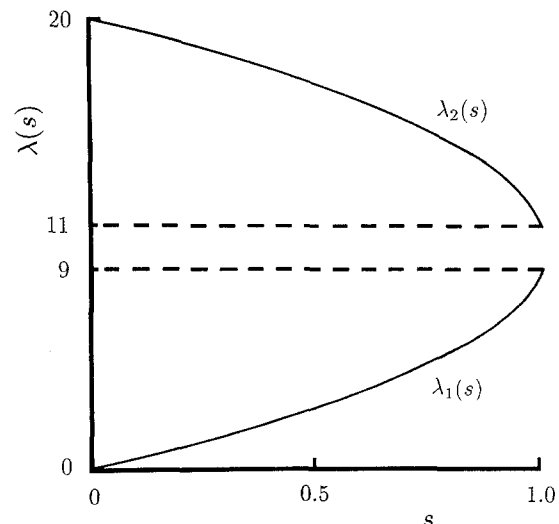


Fig. 1. Eigenvalues of  $-\mathbf{Q}_{\mathcal{C}\mathcal{C}}$ ,  $\lambda_i = \lambda_i(s)$ ,  $i = 1, 2$ , given by (2), are plotted as functions of the probability  $s$  for specified  $d_1, d_2$ . Values used in this case were  $d_1 = 9$  and  $d_2 = 11$ .

an inequality which enters only when the values of  $\hat{\lambda}_1$ ,  $\hat{\lambda}_2$  and  $\hat{s}$  are estimates that may not be constrained to satisfy this by the estimation procedure.

The common assumption that  $\hat{\lambda}_1$  is equal or approximately equal to  $\hat{d}_1$  results in an error of  $\hat{\varepsilon} = \hat{d}_1 - \hat{\lambda}_1 = \hat{\lambda}_2 - \hat{d}_2$ , which can be expressed from (2) or (10) as

$$\hat{\varepsilon} = \frac{1}{2} \{ (\hat{\lambda}_2 - \hat{\lambda}_1) - [(\hat{\lambda}_2 - \hat{\lambda}_1)^2 - 4\hat{\lambda}_1\hat{\lambda}_2(1-\hat{s})/\hat{s}]^{1/2} \}, \quad (12)$$

and has an upper limit of  $(\hat{\lambda}_2 - \hat{\lambda}_1)/2$ . The error  $\hat{\varepsilon}$  decreases (i) as  $\hat{s}$  increases (see Fig. 1), and (ii) as  $\hat{\lambda}_1/\hat{\lambda}_2$  decreases for fixed  $b$  and  $\hat{s}$ . These effects are evident in the numerical examples of Table 1, which shows relative errors of up to 90% in terms of  $d_1$  (or 900% in terms of the observed quantity  $\hat{\lambda}_1$ ). In summary, good estimates of  $d_1$  and  $d_2$  can only be obtained from knowledge of  $s$  as well as  $\lambda_1$  and  $\lambda_2$ ; without some information on the value of  $s$  considerable care should be exercised in using the fitted rate constants  $\hat{\lambda}_i$  as estimates of the reciprocal mean lifetimes  $d_i$ .

The value of  $\hat{C}_1$  places some restriction on the value of  $\hat{s}$  and hence on  $\hat{\varepsilon}$ ; for example, if  $\hat{\lambda}_1 = 1$ ,  $\hat{\lambda}_2 = 19$  and information on the value of  $\hat{C}_1$  is not used, the estimated error for  $\hat{d}_1$  and  $\hat{d}_2$  may be up to  $\varepsilon_1 = \varepsilon_2 = 90\%$  (see Table 1), but if  $\hat{C}_1 = 0.9$ ,  $\hat{s} = 0.28$  and the estimated probability  $\hat{\pi}_1$  of a sojourn in  $\mathcal{C}$  starting in state 1 is 0.125, we have  $\hat{\varepsilon} = 3.33$ ,  $\varepsilon_1 = 77\%$ , and  $\varepsilon_2 = 21\%$ . Furthermore, whenever  $\hat{C}_1 < 0.5$  a simple restriction on  $\hat{\varepsilon}$  can be demonstrated. Letting  $u = \hat{C}_1\hat{\lambda}_1 + \hat{C}_2\hat{\lambda}_2$  we have, corresponding to a rearrangement of (5)

$$\hat{\pi}_1\hat{q}_{1*} + \hat{\pi}_2\hat{q}_{2*} = u \quad (13)$$

where  $q_{1*} = d_1 - q_{12}$ ,  $q_{2*} = d_2 - q_{21}$ . Thus  $u \leq \hat{d}_2$  and it follows that

$$\hat{\varepsilon} = \hat{\lambda}_2 - \hat{d}_2 \leq \hat{\lambda}_2 - u < (\hat{\lambda}_2 - \hat{\lambda}_1)/2$$

whenever  $\hat{C}_1 < 0.5$  (that is  $u > b/2$ ), in which case  $\hat{s} \geq c/[u(b-u)]$ . Again, if  $\hat{\lambda}_1 = 1$  and  $\hat{\lambda}_2 = 19$ , and it is known that  $\hat{C}_1 = 0.05$ , we have  $u = 18.1$  and  $\hat{s} \geq 0.5525$ , with  $\hat{\varepsilon} \leq 0.9$ ,  $\varepsilon_1 \leq 47\%$  and  $\varepsilon_2 \leq 5\%$ .

### Explicit parameter estimation

As remarked above, information on the error  $\hat{\varepsilon}$  depends on the value of  $\hat{s}$ ; even a rough estimate of  $s$  should yield estimates of  $d_1$  and  $d_2$  better than  $\hat{d}_1 = \hat{\lambda}_1$ ,  $\hat{d}_2 = \hat{\lambda}_2$ . The parameter  $s$  cannot be estimated from  $\hat{\lambda}_1$  and  $\hat{\lambda}_2$ , nor from these together with  $\hat{C}_1$  (that is, from the fitted parameters of the density function) except in special cases. One such case is

$$\rightleftharpoons 1 \rightleftharpoons 2, \quad (\text{Scheme 2})$$

a reduction of Scheme 1 where exit from and entry into  $\mathcal{C}$  via state 2 are disallowed (the labelling no longer implying that  $d_1 \leq d_2$ ), so  $q_{21} = d_2$  and  $\pi_1 = 1$ . Then the fitted parameters provide explicit estimates for  $d_1$ ,  $d_2$  and  $s$ , it following from (3), (5) and (9) that

$$\begin{aligned} \hat{d}_2 &= \hat{\lambda}_1\hat{\lambda}_2/(\hat{C}_1\hat{\lambda}_1 + \hat{C}_2\hat{\lambda}_2), & \hat{d}_1 &= \hat{\lambda}_1 + \hat{\lambda}_2 - \hat{d}_2, \\ \hat{s} &= \hat{\lambda}_1\hat{\lambda}_2/(\hat{d}_1\hat{d}_2). \end{aligned} \quad (14)$$

**Table 1.** Errors in estimating  $d_1$  and  $d_2$  from  $\lambda_1$  and  $\lambda_2$  for Scheme 1. Effect of the estimates  $\hat{\lambda}_1$ ,  $\hat{\lambda}_2$  and  $\hat{s}$  on the error  $\hat{\varepsilon} = \hat{d}_1 - \hat{\lambda}_1 (= \hat{\lambda}_2 - \hat{d}_2)$  determined from (12). An asterisk indicates the minimum value (11) which  $\hat{s}$  can take if only  $\hat{\lambda}_1$  and  $\hat{\lambda}_2$  are specified. The effects of holding any two of the sum  $\hat{\lambda}_1 + \hat{\lambda}_2$ , product  $\hat{\lambda}_1\hat{\lambda}_2$  and  $\hat{s}$  constant as the other varies can be seen. Relative errors (with respect to the desired value) were calculated as  $\varepsilon_1 = 100\hat{\varepsilon}/(\hat{\lambda}_1 + \hat{\varepsilon})$ , and  $\varepsilon_2 = 100\hat{\varepsilon}/(\hat{\lambda}_2 - \hat{\varepsilon})$

$\hat{\lambda}_1$	$\hat{\lambda}_2$	$\hat{s}$	$\hat{\varepsilon}$	$\varepsilon_1$ (%)	$\varepsilon_2$ (%)
1	19	0.19*	9.000	90.0	90.0
		0.39	1.839	64.8	10.7
		0.59	0.766	43.4	4.2
		0.79	0.285	22.2	1.5
		0.99	0.011	1.1	0.1
5	15	0.75*	5.000	50.0	50.0
		0.79	2.750	35.5	22.4
		0.89	1.034	17.1	7.4
		0.99	0.076	1.5	0.5
9	11	0.99*	1.000	10.0	10.0
		0.994	0.366	3.9	3.4
		0.999	0.051	0.6	0.5
1	20	0.99	0.011	1.1	0.1
2	10	0.99	0.025	1.3	0.3
4	5	0.99	0.281	6.6	6.0

**Table 2.** Estimation of  $d_1$ ,  $d_2$  and  $s$  for Scheme 2. Estimates of  $d_1$ ,  $d_2$  and  $s$  were obtained using (14) for a range of fitted density functions based on (4). Note that in Scheme 2 the labelling of states does not imply that  $d_1 \leq d_2$

$\hat{\lambda}_1$	$\hat{\lambda}_2$	$\hat{C}_1$	$\hat{d}_1$	$\hat{d}_2$	$\hat{s}$
1	19	0.1	18.90	1.10	0.9103
		0.5	18.10	1.90	0.5525
		0.9	13.21	6.79	0.2119
		0.95	10.00	10.00	0.1900
		0.99	3.90	16.10	0.3027
		0.999	1.34	18.66	0.7620
5	15	0.1	14.64	5.36	0.9561
		0.5	12.50	7.50	0.8000
		0.9	7.50	12.50	0.8000
		0.99	5.29	14.71	0.9633
9	11	0.01	10.98	9.02	0.9997
		0.5	10.10	9.90	0.9901
		0.99	9.02	10.98	0.9995

Table 2 presents solutions for  $\hat{d}_1$ ,  $\hat{d}_2$  and  $\hat{s}$  from (14) based on a range of fitted density functions. Even modest changes in the coefficients of the bi-exponential density result in marked shifts in the values of  $\hat{d}_1$  and  $\hat{d}_2$ . In particular, the finding that sojourn times can be fitted with a bi-exponential density having well separated parameters ( $\hat{\lambda}_1 \ll \hat{\lambda}_2$ ) does not necessarily indicate that the two underlying kinetic states have commensurately unequal mean lifetimes; they may even be equal. Nor can it be assumed that if one state really is short-lived, it must be the first visited (that is state 1 of Scheme 2). Though this conclusion may be suggested by other evidence, such as bursting behaviour, it can be confirmed by solution of (14) using the estimate  $\hat{C}_1$ .

It is worth noting that the transition rate  $q_{1*} = d_1 - q_{12} = d_1$  s out of class  $\mathcal{C}$  is, from (13) or (14), simply estimated by  $\hat{q}_{1*} = u$ , a result which extends naturally to a class of more than two states. That is, for any class  $\mathcal{C}$  containing  $n$  states where entry to and exit from  $\mathcal{C}$  can occur via only one state (say state 1), the transition rate out of  $\mathcal{C}$  can be estimated by  $\hat{q}_{1*} = \sum_{i=1}^n \hat{C}_i \hat{\lambda}_i$ . Thus, in the four state sequential model of nicotinic or GABA receptor agonist binding and isomerization,



where  $\mathcal{C} = \{1, 2, 3\}$  is a class of closed states and state 4 is open, the channel opening rate constant  $\beta = q_{14}$  can be estimated explicitly from the closed-time density (see Madsen and Edeson 1988, for a survey of  $\beta$  estimates). This may be a useful alternative, particularly in systems where bursting is not evident, to the widely applied method based on the eigenvalue approximation for  $d_1$  and an estimate of the number of gaps per burst (represented, for example, by (12) and (13) of Colquhoun and Sakmann 1985). Moreover,  $\sum_{i=1}^3 C_i \lambda_i$  is independent of agonist concentration (as can be seen from (3.75) of Colquhoun and Hawkes 1981, where  $f_{34}(0) = \beta$ ) and so in principle is  $\hat{q}_{14}$ , whereas the standard estimation procedure assumes zero concentration (Colquhoun and Hawkes 1981, (3.9)).

As always, the quality of the estimation (for both methods) is dependent on the appropriateness of the model; thus it is assumed that the experimental record arose from a patch containing only one channel and that desensitization is not present. If the number of channels in the patch is unknown,  $v = \sum_{i=1}^3 \hat{C}_i \hat{\lambda}_i$  represents (again in principle) an upper limit to  $\beta$ . In the case of desensitization, if the real underlying mechanism is described by Scheme 3 with states 2 and 3 normal closed, state 1 open, and state 4 desensitized (or blocked), the closed-time density will reflect sojourns in state 4 as well as states 2 and 3. One eigenvalue (say  $\lambda_3$ ) will equal  $d_4$  and  $C_3 = p_{14}$ . Then  $\beta = q_{21} = \left[ \sum_{i=1}^2 C_i \lambda_i \right] / (1 - C_3)$  and the estimation error in assuming desensitization is not present will be  $v - \hat{\beta} = \hat{C}_3 (\hat{\lambda}_3 - v) / (1 - \hat{C}_3)$ . Thus  $v$  will be greater than (less than)  $\hat{\beta}$  whenever  $\hat{d}_4 = \hat{\lambda}_3$  is greater than (less than)  $v$ , for example when  $\hat{\lambda}_3$  is greater than (less than) both  $\hat{\lambda}_1$  and  $\hat{\lambda}_2$ . If the probability that an open channel closes to the desensitized state is low this error will tend to be small.

Some comparisons of estimates given by the two methods are possible where published studies provide sufficient information to calculate  $v$ . These suggest that differences may often be relatively small, but can also be appreciable. For example, Hestrin et al. (1987) give a value for  $\hat{\beta}$  of  $1587 \text{ s}^{-1}$ , 16% higher than  $v$  calculated from their data on p 450; Sine and Steinbach (1986) estimate  $\beta$  to be  $14\,764 \text{ s}^{-1}$  (suberyldicholine data, brief closure assumption), which is 51% higher than  $v$  obtained from their data of Fig. 2. How these differences should be interpreted, and the relative merits of the two estimation

methods in terms of approximation errors, selection of bursts from the experimental record, unknown number of channels, possible desensitization, limited time resolution and digitization rate remain to be studied.

**Acknowledgement.** Supported by the Faculty of Anaesthetists of the Royal Australasian College of Surgeons through a Harry Daly Research Fellowship (ROE).

## References

- Auerbach A, Lingle CJ (1987) Activation of the primary kinetic modes of large- and small-conductance cholinergic ion channels in *Xenopus* myocytes. *J Physiol (London)* 393:437–466
- Ball FG, Sansom MSP (1988) Single channel autocorrelation functions – the effects of time interval omission. *Biophys J* 53:819–832
- Bormann J (1988) Patch-clamp analysis of GABA- and glycine-gated chloride channels. In: Biggio G, Costa E (eds) *Chloride channels and their modulation by neurotransmitters and drugs*. Raven Press, New York, pp 47–60
- Colquhoun D, Hawkes AG (1981) On the stochastic properties of single ion channels. *Proc R Soc London B* 211:205–235
- Colquhoun D, Hawkes AG (1982) On the stochastic properties of bursts of single ion channel openings and of clusters of bursts. *Phil Trans R Soc London B* 300:1–59
- Colquhoun D, Hawkes AG (1983) The principles of the stochastic interpretation of ion-channel mechanisms. In: Sakmann B, Neher E (eds) *Single-channel recording*. Plenum Press, New York, pp 135–175
- Colquhoun D, Sakmann B (1985) Fast events in single-channel currents activated by acetylcholine and its analogues at the frog muscle end-plate. *J Physiol (London)* 369:501–557
- Fredkin D, Rice JA (1986) On aggregated Markov processes. *J Appl Probab* 23:208–214
- Fredkin DR, Montal M, Rice JA (1985) Identification of aggregated Markovian models: application to the nicotinic acetylcholine receptor. In: Le Cam L, Olshen R (eds) *Proceedings of the Berkeley conference in honor of Jerzy Neyman and Jack Keifer*, vol 1, Wadsworth, Monterey, pp 269–289
- Hestrin S, Korenbrot JJ, Maricq AV (1987) Kinetics of activation of acetylcholine receptors in a mouse muscle cell line under a range of acetylcholine concentrations. *Biophys J* 51:449–455
- Igusa Y, Kidokoro Y (1987) Two types of acetylcholine receptor channels in developing *Xenopus* muscle cells in culture: further kinetic analyses. *J Physiol (London)* 389:271–300
- Jaramillo F, Schuetze SM (1988) Kinetic differences between embryonic- and adult-type acetylcholine receptors in rat myotubes. *J Physiol (London)* 396:267–296
- Madsen BW, Edeson RO (1988) Nicotinic receptors and the elusive  $\beta$ . *Trends Pharmacol Sci* 9:315–316
- Milne RK, Yeo GF, Edeson RO, Madsen BW (1988) Stochastic modelling of a single ion channel: an alternating renewal approach with application to limited time resolution. *Proc R Soc London B* 233:247–292
- Papke RL, Millhauser G, Lieberman Z, Oswald RE (1988) Relationships of agonist properties to the single channel kinetics of nicotinic acetylcholine receptors. *Biophys J* 53:1–10
- Sikdar SK, McIntosh RP, Mason WT (1989) Differential modulation of  $\text{Ca}^{2+}$ -activated  $\text{K}^+$  channels in ovine pituitary gonadotrophs by GnRH,  $\text{Ca}^{2+}$  and cyclic AMP. *Brain Res* 496:113–123
- Sine SM, Steinbach JH (1984) Activation of a nicotinic acetylcholine receptor. *Biophys J* 45:175–185
- Sine SM, Steinbach JH (1986) Activation of acetylcholine receptors on clonal mammalian BC3H-1 cells by low concentrations of agonist. *J Physiol (London)* 373:129–162