

TMA4195 Mathematical Modelling Project

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Deriving the modelling equations

Diffusion equation

Flux J :

$$J = -D\nabla c$$

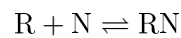
$$\begin{aligned}c_t &= \kappa \Delta c \\ \frac{dc}{dt} &= \kappa \nabla^2 c\end{aligned}$$

Neumann BC:

$$\nabla c \cdot n = g(t, c)$$

The binding process

First we look at the reversible chemical reaction



with reaction rate k_1^* to the right and k_2^* to the left, being respectively the probability for the reactions to occur in their direction. We get 3 ODE's from this chemical reaction:

$$\begin{aligned}\frac{d[\text{R}]}{dt} &= -k_1^*[\text{R}][\text{N}] + k_2^*[\text{RN}], \\ \frac{d[\text{N}]}{dt} &= -k_1^*[\text{R}][\text{N}] + k_2^*[\text{RN}], \\ \frac{d[\text{RN}]}{dt} &= k_1^*[\text{R}][\text{N}] - k_2^*[\text{RN}],\end{aligned}$$

where $[R]$, $[N]$ and $[RN]$ are the concentrations of the receptors, neurotransmitters and the bound product of them. We may consider $[N][R]$ the probability of a neurotransmitter meeting an unoccupied receptor, and \bar{k}_1^* the probability of the binding reaction happening. Likewise for \bar{k}_2^* . P^R is the probability of a receptor being unoccupied, $(1 - P^R)$ the probability that the neurotransmitter is attached to the receptor, leads to the following simplification of the above ODE's:

$$\begin{aligned}\frac{d[N]}{dt} &= -\bar{k}_1^*[N]P^R + \bar{k}_2^*(1 - P^R), \\ \frac{dP^R}{dt} &= -\bar{k}_1^*[N]P^R + \bar{k}_2^*(1 - P^R).\end{aligned}$$

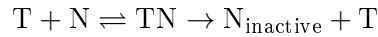
If we assume that the receptors are not uniformly distributed, we need to introduce a $\gamma(x)$ to describe the density of receptors. At the boundary:

$$\begin{aligned}\frac{d[N]}{dt} &= -\bar{k}_1^*[N]\gamma P^R + \bar{k}_2^*\gamma(1 - P^R), \\ \frac{dP^R}{dt} &= -\bar{k}_1^*[N]\gamma P^R + \bar{k}_2^*\gamma(1 - P^R),\end{aligned}$$

which are Neumann boundary conditions (inserting c for $[N]$)

$$\begin{aligned}\kappa \nabla c \cdot n &= -\bar{k}_1^*c\gamma P^R + \bar{k}_2^*\gamma(1 - P^R), \\ \frac{dP^R}{dt} &= -\bar{k}_1^*cP^R + \bar{k}_2^*(1 - P^R),\end{aligned}$$

Glia cells



Define k_3, k_4, k_5 as the reaction rates of first rightward, first leftward, second rightward equation.

Similarly to the binding process, we get the following sets of equations:

$$\begin{aligned}\kappa \nabla c \cdot n &= -\bar{k}_3c\gamma^T P^T + \bar{k}_4\gamma^T(1 - P^T), \\ \frac{dP^T}{dt} &= -\bar{k}_3cP^T + (1 - P^T)(\bar{k}_4 + \bar{k}_5),\end{aligned}$$

Combining these equations, we get

$$\begin{aligned}\kappa \nabla c \cdot n &= -c(\bar{k}_1^* \gamma^R P^R + \bar{k}_3 \gamma^T P^T) + \bar{k}_2^* \gamma^R (1 - P^R) + \bar{k}_4 \gamma^T (1 - P^T), \\ \frac{dP^R}{dt} &= -c \bar{k}_1^* P^R + \bar{k}_2^* (1 - P^R), \\ \frac{dP^T}{dt} &= -c \bar{k}_3 P^T + (\bar{k}_4 + \bar{k}_5)(1 - P^T),\end{aligned}$$

1D solution using solver

Modelling the equation in one dimension is done by considering the points a and b , and the line between them. In this case, a is on one side of the synaptic cleft, and b is on the other side. Due to this, the boundary conditions for a and b differ. For a , we have

$$\kappa \nabla c = -k_3 c P^T + k_4 (1 - P^T),$$

and for b we have

$$\kappa \nabla c = -k_1 c P^R + k_2 (1 - P^R).$$

The next step is to combine these boundary conditions with the modelling equation to form a matrix equation.

The final equation was

$$\begin{aligned}\hat{M} \dot{X}(t) &= -\kappa \hat{K} X(t) - k_3 \hat{Q}^a(X(t)) X(t) + k_4 (1 - X_{N+1}(t)) \hat{d}^a \\ &+ (k_4 + k_5)(1 - X_{N+1}(t)) \hat{e}^a - k_1 \hat{Q}^b(X(t)) X(t) + k_2 (1 - X_{N-2}(t)) \hat{d}^b,\end{aligned}$$

and was found by Jorg Henrik Holstad ¹ Here, X is a vector of length $N + 2$, consisting of the concentrations at the nodes, as well as the probabilities

$$P_a^T \text{ and } P_b^R: X = \begin{bmatrix} C \\ P_a^T \\ P_b^R \end{bmatrix}.$$

A plot using $N = 9$ internal nodes is shown below:

¹Jorg Henrik Holstad. (2011). Modelling av Diffusjon av Nevrotransmittere i den Ekstracellelaere Vaesken. Retrieved from <https://www.duo.uio.no/bitstream/handle/10852/10871/MasteroppgaveHenrikHolstad.pdf>

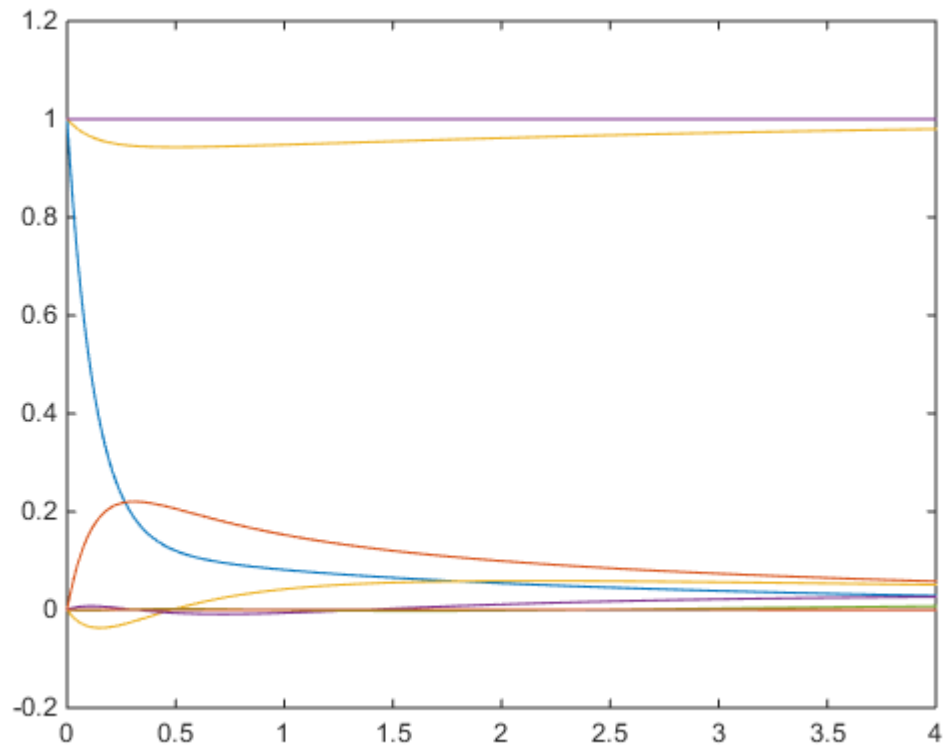


Figure 1: $a = 0$, $b = 8$, $N = 9$, $k_1 = k_2 = k_3 = k_4 = k_5 = 0.5$, $P_a^T(0) = P_b^R(0) = 1$