

Local Asymptotic Normality for Stochastic Hodgkin-Huxley-Systems

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Classical Hodgkin-Huxley-System

$$dV_t = -F(V_t, n_t, m_t, h_t)dt + S(t)dt$$

$$dn_t = [\alpha_n(V_t)(1 - n_t) - \beta_n(V_t)n_t]dt$$

$$dm_t = [\alpha_m(V_t)(1 - m_t) - \beta_m(V_t)m_t]dt$$

$$dh_t = [\alpha_h(V_t)(1 - h_t) - \beta_h(V_t)h_t]dt$$

with

- membrane potential V of a neuron
- deterministic periodic external input signal S
- internal gating variables n, m, h modeling activation of ion channels
- (explicit) smooth coefficient functions F, α_i, β_i

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Stochastic Hodgkin-Huxley-System

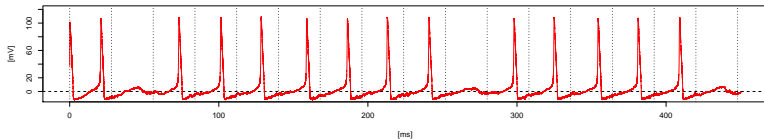
$$\begin{aligned}dV_t &= -F(V_t, n_t, m_t, h_t)dt + d\xi_t \\dn_t &= [\alpha_n(V_t)(1 - n_t) - \beta_n(V_t)n_t] dt \\dm_t &= [\alpha_m(V_t)(1 - m_t) - \beta_m(V_t)m_t] dt \\dh_t &= [\alpha_h(V_t)(1 - h_t) - \beta_h(V_t)h_t] dt \\d\xi_t &= \gamma(S(t) - \xi_t)dt + \sigma(\xi_t)dW_t\end{aligned}$$

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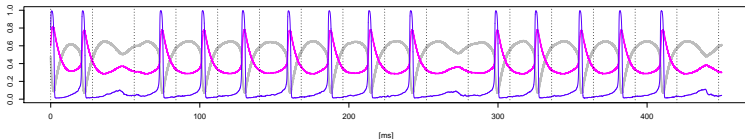
- membrane potential V of a neuron
- deterministic periodic external input signal S
- internal gating variables n, m, h modeling activation of ion channels
- (explicit) smooth coefficient functions F, α_i, β_i
- $\gamma > 0$ and $\sigma \in C_b^3$ bounded away from 0, W 1D Brownian Motion

Some Sample Paths

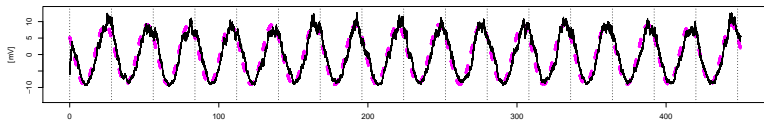
stochastic HH with periodic signal: voltage $v(t)$ function of t ; black dotted line indicating periodicity of the semigroup



stochastic HH with periodic signal: gating variables $n(t)$ (violet), $m(t)$ (blue), $h(t)$ (grey) functions of t



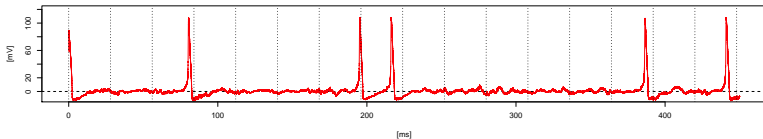
stochastic HH with periodic signal: periodic signal and driving noisy input (mean reverting CIR type diffusion)



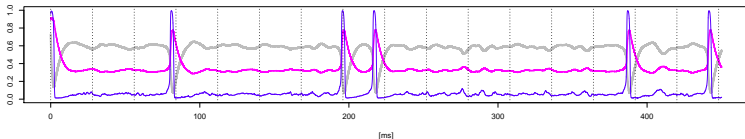
the following parameters were used for signal and CIR : period = 28 , amplitude = 9 , sigma = 0.5 , tau = 0.75 , K = 30

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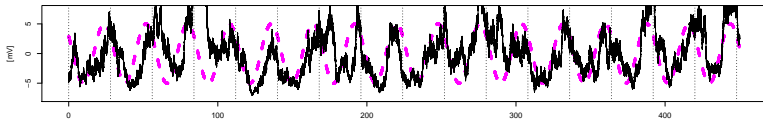
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stochastic HH with periodic signal: periodic signal and driving noisy input (mean reverting CIR type diffusion)



the following parameters were used for signal and CIR : period = 28 , amplitude = 5 , sigma = 1.5 , tau = 0.25 , K = 30

Introduce Parametrized Signal

Let $S = S_{(\vartheta, T)}$ depend on an unknown shape parameter $\vartheta \in \Theta \subset \mathbb{R}^d$ and let it be periodic with unknown periodicity $T \in (0, \infty)$.

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Then the equation for $X = (V, n, m, h, \xi)$ is of the form

$$dX_t = B_{(\vartheta, T)}(t, X_t)dt + \Sigma(X_t)dW_t$$

and its solution lives on $\mathbb{R} \times [0, 1]^3 \times \mathbb{R}$ (if started there).

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Long-term goal:

Estimate (ϑ, T) from continuous observation not of X , but only of the membrane potential V .

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Immediate goal:

Prove LAN for the corresponding sequence of statistical experiments.

Likelihood-Ratios for Observation of X , V or ξ

Write $\mathbb{P}^{(\vartheta, T)}$ for the law on $C([0, \infty); \mathbb{R}^5)$ under which the canonical process $(\eta_t)_{t \geq 0}$ solves the stochastic Hodgkin-Huxley system starting from a fixed and deterministic $x_0 = (v_0, n_0, m_0, h_0, \xi_0)$.

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These lead to three different sequences of experiments

$$\left(C([0, \infty); \mathbb{R}^5), \mathcal{F}_n^i, \left\{ \mathbb{P}^{(\vartheta, T)} \Big|_{\mathcal{F}_n^i} \mid (\vartheta, T) \in \Theta \times (0, \infty) \right\} \right), \quad n \in \mathbb{N}$$

where $i = 0, 1, 5$.

Likelihood-Ratios for Observation of X , V or ξ

However, as the parameters are only present in the drift term for the fifth (and thus also the first) equation and the local martingale part of X under $\mathbb{P}^{(\vartheta, T)}|_{\mathcal{F}_n^0}$ is given by

$$\left(\int_0^\cdot \sigma(\xi_t) dW_t, 0, 0, 0, \int_0^\cdot \sigma(\xi_t) dW_t \right)^\top,$$

we can conclude that for all $i \in \{0, 1, 5\}$ under $\mathbb{P}^{(\vartheta, T)}|_{\mathcal{F}_t^i}$

$$\begin{aligned} \log \left(\frac{d\mathbb{P}^{(\vartheta', T')}|_{\mathcal{F}_t^i}}{d\mathbb{P}^{(\vartheta, T)}|_{\mathcal{F}_t^i}} \right) &\stackrel{d}{=} \gamma \int_0^t \frac{S_{(\vartheta', T')}(s) - S_{(\vartheta, T)}(s)}{\sigma(\xi_s)} dW_s \\ &\quad - \frac{\gamma^2}{2} \int_0^t \left(\frac{S_{(\vartheta', T')}(s) - S_{(\vartheta, T)}(s)}{\sigma(\xi_s)} \right)^2 ds \\ &=: \Lambda_t^{(\vartheta', T')/(\vartheta, T)}. \end{aligned}$$

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\Rightarrow If LAN holds for *any* of these sequences, it holds for *all* of them.

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- 4 $\dot{S}: (\vartheta, T) \mapsto \dot{S}_{(\vartheta, T)}$ is $\mathbb{L}_{\text{loc}}^2$ -continuous.
- 5 For each $(\vartheta, T) \in \Theta \times (0, \infty)$ there are $\alpha \in (0, 1]$ and $\beta \in [0, (1 + 3\alpha)/2)$ such that for suitable $\varepsilon > 0$

$$\|\nabla_\vartheta S_{(\vartheta, T)} - \nabla_\vartheta S_{(\vartheta, T')}\|_{\mathbb{L}^2(0, t)} \leq Ct^\beta |T - T'|^\alpha$$

for all $t > 0$, $T' \in (T - \varepsilon, T + \varepsilon)$ and some constant C that does not depend on T' or t .

Local Asymptotic normality

Fix $(\vartheta, T) \in \Theta \times (0, \infty)$. Suppose that for each $t > 0$ the matrix

$$J^{(\vartheta, T)}(\mathbf{t}) = \gamma^2 \nu \left[\begin{pmatrix} \nabla_{\vartheta} S_{\vartheta} \\ -tT^{-2}S'_{\vartheta} \end{pmatrix} \begin{pmatrix} \nabla_{\vartheta} S_{\vartheta} \\ -tT^{-2}S'_{\vartheta} \end{pmatrix}^{\top} \right] \in \mathbb{R}^{(d+1) \times (d+1)}$$

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$$\delta_n := \text{diag} \left(n^{-1/2}, \dots, n^{-1/2}, n^{-3/2} \right) \in \mathbb{R}^{(d+1) \times (d+1)}.$$

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Theorem (Local Asymptotic Normality)

With

$$I^{(\vartheta, T)} := \int_0^1 J^{(\vartheta, T)}(s) ds \quad \text{and} \quad \Delta_n^{(\vartheta, T)} := \gamma \delta_n \int_0^1 \frac{\dot{S}^{(\vartheta, T)}(s)}{\sigma(\xi_s)} dW_s$$

we have $\Delta_n^{(\vartheta, T)} \xrightarrow{\mathcal{L}} \mathcal{N}(0, I^{(\vartheta, T)})$ and

$$\Lambda_n^{(\vartheta_n, T_n)/(\vartheta, T)} = h_n^{\top} \Delta_n^{(\vartheta, T)} - \frac{1}{2} h_n^{\top} I^{(\vartheta, T)} h_n + o_{\mathbb{P}^{(\vartheta, T)}}(1).$$

Tools for the Proof

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- The *path segment chain* $(\Xi_k)_{k \in \mathbb{N}_0}$ with

$$\begin{aligned}\Xi_k &:= (\xi_{(k-1)T+s})_{s \in [0, T]}, \quad k \in \mathbb{N}, \\ \Xi_0 &\in C[0, T] \text{ arbitrary with } \Xi_0(T) = \xi_0,\end{aligned}$$

is a $C[0, T]$ -valued time homogeneous Markov chain. It inherits positive Harris-recurrence from the grid chain and we denote its invariant probability measure by m .

Tools for the Proof

Strong Law of Large Numbers for Ξ (Höpfner, Kutoyants, 2010)

$(A_t)_{t \geq 0}$ increasing process, $F \in \mathbb{L}^1(m)$ nonnegative with

$$A_{kT} = \sum_{j=1}^k F(\Xi_j) \quad \text{for all } k \in \mathbb{N}.$$

Then

$$\frac{1}{t} A_t \xrightarrow{t \rightarrow \infty} \frac{1}{T} m(F) \quad \mathbb{P}^{(\vartheta, T)\text{-a.s.}}$$

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Apply this to show that for 1-periodic bounded measurable f and $m \in \mathbb{N}_0$

$$\frac{1}{t} \int_0^t \frac{f(s/T)}{\sigma^2(\xi_s)} ds \xrightarrow{t \rightarrow \infty} \int_0^1 f(s) \underbrace{\mu P_{0,sT}(\sigma^{-2})}_{=: \nu(ds)} ds \quad \mathbb{P}^{(\vartheta, T)\text{-a.s.}}$$

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$$\frac{1}{n^{m+1}} \int_0^{tn} s^m \frac{f(s/T)}{\sigma^2(\xi_s)} ds \xrightarrow{n \rightarrow \infty} \frac{t^{m+1}}{m+1} \int_0^1 f(s) \underbrace{\mu P_{0,sT}(\sigma^{-2}) ds}_{=: \nu(ds)} \quad \mathbb{P}^{(\vartheta, T)\text{-a.s.}}$$

Main Step of the Proof

$$\begin{aligned}
 & \left\langle \delta_n \int_0^{\cdot n} \frac{\dot{S}_{(\vartheta, T)}(s)}{\sigma(\xi_s)} dW_s \right\rangle_t = \delta_n^2 \int_0^{tn} \frac{\dot{S}_{(\vartheta, T)}(s) \dot{S}_{(\vartheta, T)}(s)^\top}{\sigma^2(\xi_s)} ds \\
 & = \int_0^{tn} \left(\begin{array}{cc} n^{-1} \nabla_{\vartheta} S_{\vartheta}(\frac{s}{T}) \nabla_{\vartheta} S_{\vartheta}(\frac{s}{T})^\top & n^{-2} (-s T^{-2} S'_{\vartheta}(\frac{s}{T}) \nabla_{\vartheta} S_{\vartheta}(\frac{s}{T})) \\ n^{-2} (-s T^{-2} S'_{\vartheta}(\frac{s}{T}) \nabla_{\vartheta} S_{\vartheta}(\frac{s}{T}))^\top & n^{-3} s^2 T^{-4} (S'_{\vartheta}(\frac{s}{T}))^2 \end{array} \right) \sigma^{-2}(\xi_s) ds \\
 & \xrightarrow{n \rightarrow \infty} \nu \left[\left(\begin{array}{cc} t \nabla_{\vartheta} S_{\vartheta} \nabla_{\vartheta} S_{\vartheta}^\top & - \left(\frac{t^2}{2} T^{-2} S'_{\vartheta} \nabla_{\vartheta} S_{\vartheta} \right) \\ - \left(\frac{t^2}{2} T^{-2} S'_{\vartheta} \nabla_{\vartheta} S_{\vartheta} \right)^\top & \frac{t^3}{3} T^{-4} (S'_{\vartheta})^2 \end{array} \right) \right] \\
 & = \int_0^t \nu \left[\left(\begin{array}{c} \nabla_{\vartheta} S_{\vartheta} \\ -s T^{-2} S'_{\vartheta} \end{array} \right) \left(\begin{array}{c} \nabla_{\vartheta} S_{\vartheta} \\ -s T^{-2} S'_{\vartheta} \end{array} \right)^\top \right] ds
 \end{aligned}$$

Example

- A simple example for a signal that satisfies the regularity assumptions is

$$S_{(\vartheta, T)}(s) = \sum_{k=0}^l \left(g_k(\vartheta) \sin\left(\frac{2k\pi s}{T}\right) + h_k(\vartheta) \cos\left(\frac{2k\pi s}{T}\right) \right)$$

with $l \in \mathbb{N}_0$ and $g_k, h_k \in C^1(\Theta)$ for all $k \in \{0, \dots, n\}$.

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with $l \in \mathbb{N}_0$ and $g_k, h_k \in C^1(\Theta)$ for all $k \in \{0, \dots, n\}$.

- For $\sigma \equiv 1$ and the above signal with $l = d$, $h_k \equiv 0$ and g_k depending only on ϑ_k , the invertibility condition for $J^{(\vartheta, T)}(t)$ also holds.

Next Step

Construct estimator(s) for (ϑ, T) involving only V .

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[to appear on arXiv soon\(ish\)](#)



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SISP No. 13



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