

MLE for Poisson Process with Increasing Rate

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Observe time range $[0, T]$, obtain events at times $\mathbf{t} = (t_1, \dots, t_n)$, where $n \geq 1$, $0 < t_1 < \dots < t_n < T$
 Events generated by Poisson process with rate function $\lambda : [0, T] \rightarrow [0, \infty)$

Log-likelihood of observation: $\ell(\lambda; \mathbf{t}) = \sum_{i=1}^n \ln(\lambda(t_i)) - \int_0^T \lambda(t) dt$

If λ unrestricted, can obtain arbitrarily high $\ell(\lambda; \mathbf{t})$ by concentrating value around observations

Instead, assume λ (non-strictly) increasing: how to maximize $\ell(\lambda; \mathbf{t})$ (equivalently, minimize $-\ell(\lambda; \mathbf{t})$)?

Given increasing λ , may define increasing λ' as follows:

$$\lambda'(t) = \begin{cases} 0 & t \in [0, t_1) \\ \lambda(t_i) & t \in [t_i, t_{i+1}) \text{ for } 1 \leq i \leq n \text{ (convention: } t_{n+1} = T) \end{cases}$$

Intuition: replace rate before first event with 0, extend rate at each event to the right until next event

Then $\lambda'(t) \leq \lambda(t)$ for all t , so $\int_0^T \lambda'(t) dt \leq \int_0^T \lambda(t) dt$,

and $\lambda'(t_i) = \lambda(t_i)$ for all i , so $\sum_{i=1}^n \ln(\lambda'(t_i)) = \sum_{i=1}^n \ln(\lambda(t_i))$

Thus $\ell(\lambda'; \mathbf{t}) \geq \ell(\lambda; \mathbf{t})$, so search space may be restricted to all such λ' :

if MLE among λ' does not exist, it does not exist in general, and if it does, it is MLE in general

From here, λ assumed to be of this form

λ may be parameterized by $\lambda_1 = \lambda(t_1), \dots, \lambda_n = \lambda(t_n)$

Then $\ell(\lambda; \mathbf{t}) = \sum_{i=1}^n \ln(\lambda_i) - \sum_{i=1}^n s_i \lambda_i$, where $s_i = t_{i+1} - t_i > 0$

λ increasing, so $0 < \lambda_1 \leq \dots \leq \lambda_n$

Lemma: Let f be a convex function over convex $C \subseteq \mathbb{R}^n$, let $D \subseteq \mathbb{R}^n$ be closed and convex, and suppose $C \cap D$ is nonempty. If f attains min over C at $\mathbf{x} \notin D$, then f attains min over $C \cap D$ at some $\mathbf{w} \in \partial D$.

Proof: Let \mathbf{y} be a point in $C \cap D$ where f is minimized. Then the segment between \mathbf{x} and \mathbf{y} intersects ∂D (if $\mathbf{y} \in \partial D$, at \mathbf{y} itself). Call this intersection \mathbf{z} . Then $f(\mathbf{z}) \leq f(\mathbf{y})$ by convexity.

Apply lemma with $C = (\mathbb{R}^+)^n$, $D = \{\boldsymbol{\lambda} = (\lambda_1, \dots, \lambda_n) \in \mathbb{R}^n \mid \lambda_1 \leq \dots \leq \lambda_n\}$

WTS: if unconstrained problem has $\lambda_i > \lambda_{i+1}$ for some i , then constrained problem has $\lambda_i = \lambda_{i+1}$ (for one of those i , not an arbitrary one at first)

Next, WTS

Integral of MLE is underside of convex hull of $\{(t_i, i) \mid 0 \leq i \leq n+1\}$ ($t_0 = 0$)
May help find bias/variance