
A note on the L2-convergence rates of derivatives

Johan Sebastian Ohlendorff

johan.ohlendorff@sund.ku.dk

University of Copenhagen

ABSTRACT

In this brief note, ...

1 Main section

Let the $L_2(\nu)$ -norm of a function $f \in L_2(\nu)$ be defined as

$$\|f\|_\nu = \sqrt{\int f^2 d\nu}. \quad (1)$$

Consider a sequence of estimators $\hat{P}_n(t | x)$ of $P(t | x)$ which are defined on $[0, \tau]$. We assume that $\hat{P}_n(0 | x) = P(0 | x) = 0$. We let μ_0 denote an appropriate measure for the covariates x . These are assumed to have the $L_2(\mu_0)$ -convergence rate $n^{-\frac{1}{4}-\varepsilon}$ for Lebesgue almost all $t \in [0, \tau]$ for $\varepsilon > \frac{1}{12}$. This corresponds to a convergence rate of slightly better than $n^{-\frac{1}{3}}$. We are interested in constructing an estimator $p(t | x) = P'(t | x)$ of the derivative of $P(t | x)$ which also has the $L_2(\mu_0 \otimes m)$ -convergence rate $n^{-\frac{1}{4}}$, where m is the Lebesgue measure on $[0, \tau]$. The precise statement is given in Theorem 1.1. This is useful if one wishes to obtain convergence rates for a hazard function which one has not explicitly considered, but only the cumulative hazard function, such as in a Cox regression.

Theorem 1.1: Let $\hat{P}_n(t | x)$ be a sequence of estimators of $P(t | x)$ defined on $[0, \tau]$ fulfilling that $\hat{P}_n(0 | x) = P(0 | x) = 0$. Suppose that $P(t | x) \in C^2([0, \tau])$ for μ_0 -almost all x and that there exists a constant $K > 0$ such that $p'(t | x) \leq K$ for μ_0 -almost all x and $t \in [0, \tau]$. If $\|\hat{P}_n(t | \cdot) - P(t | \cdot)\|_{\mu_0} = o_P(n^{-\frac{1}{4}-\varepsilon})$ for Lebesgue almost all $t \in [0, \tau]$ for $\varepsilon > \frac{1}{12}$, then there exists an estimator $\hat{p}_n(t | x)$ of $p(t | x) = P'(t | x)$ such that

$$\|\hat{p}_n - p\|_{\mu_0 \otimes m} = o_P(n^{-\frac{1}{4}}). \quad (2)$$

The estimator $\hat{p}_n(t | x)$ fulfills on a grid $0 = t_1 < \dots < t_{K_n} = \tau$ with some mesh $b(n) = \max_{1 \leq k \leq K_n} (t_k - t_{k-1}) \rightarrow 0$ as $n \rightarrow \infty$ and $K_n \rightarrow \infty$ as $n \rightarrow \infty$ determined by ε such

$$\int_0^{t_k} \hat{p}_n(s | x) ds = \hat{P}_n(t_k | x). \quad (3)$$

Proof: Consider a partition $0 = t_1 < \dots < t_{K_n} = t$ of $[0, t]$ with mesh $b(K_n) = \max_{1 \leq k \leq K_n} (t_k - t_{k-1})$. Choose $x_1 > 0$ and $x_2 < 0$ such that $0 < x_1 < \frac{3}{4}\varepsilon - \frac{1}{16}$ and $2(x_1 - \varepsilon) < x_2 < -\frac{2}{3}\varepsilon - \frac{1}{6} < 0$. Here, we will let $K_n = \lfloor n^{x_1} \rfloor$ and $b(K_n) = \lfloor n^{x_2} \rfloor$. Then $K_n \rightarrow \infty$ as $n \rightarrow \infty$ because $\varepsilon > \frac{1}{12}$ and $b(K_n) \rightarrow 0$ as $n \rightarrow \infty$ because $\varepsilon > 0$. We will show the

theorem by constructing an explicit estimator $\hat{p}_n(t | x)$ by approximating the derivative via a secant. Let

$$\hat{p}_n(t | x) = \sum_{k=1}^{K_n} \mathbb{1}\{t \in (t_k, t_{k+1}]\} \frac{\hat{P}_n(t_{k+1} | x) - \hat{P}_n(t_k | x)}{t_{k+1} - t_k} \quad (4)$$

Then evidently, we have

$$\int_0^{t_k} \hat{p}_n(s | x) ds = \sum_{j=1}^{k-1} \frac{\hat{P}_n(t_{k+1} | x) - \hat{P}_n(t_k | x)}{t_{k+1} - t_k} (t_{k+1} - t_k) = \hat{P}_n(t_k | x). \quad (5)$$

Furthermore, let

$$\tilde{p}_n(t | x) = \sum_{k=1}^{K_n} \mathbb{1}\{t \in (t_k, t_{k+1}]\} \frac{P(t_{k+1} | x) - P(t_k | x)}{t_{k+1} - t_k}. \quad (6)$$

By the triangle inequality, we have

$$\|\hat{p}_n - p\|_{\mu_0 \otimes m} \leq \|\hat{p}_n - \tilde{p}_n\|_{\mu_0 \otimes m} + \|\tilde{p}_n - p\|_{\mu_0 \otimes m}. \quad (7)$$

We start with the first term on the right-hand side.

$$\begin{aligned} \|\hat{p}_n - \tilde{p}_n\|_{\mu_0 \otimes m} &= \left\| \sum_{k=1}^{K_n} \mathbb{1}\{\cdot \in (t_k, t_{k+1}]\} \frac{(\hat{P}_n(t_{k+1} | \cdot) - P(t_{k+1} | \cdot)) - (\hat{P}_n(t_k | \cdot) - P(t_k | \cdot))}{t_{k+1} - t_k} \right\|_{\mu_0 \otimes m} \\ &\leq \sum_{k=1}^{K_n} \left\| \mathbb{1}\{\cdot \in (t_k, t_{k+1}]\} \frac{\hat{P}_n(t_{k+1} | \cdot) - P(t_{k+1} | \cdot) - (\hat{P}_n(t_k | \cdot) - P(t_k | \cdot))}{t_{k+1} - t_k} \right\|_{\mu_0 \otimes m} \quad (8) \\ &\leq \sum_{k=1}^{K_n} \frac{1}{\sqrt{t_{k+1} - t_k}} \left(\|\hat{P}_n(t_{k+1} | \cdot) - P(t_{k+1} | \cdot)\|_{\mu_0} + \|\hat{P}_n(t_k | \cdot) - P(t_k | \cdot)\|_{\mu_0} \right) \\ &= o(n^{x_1 - \frac{1}{2}x_2}) o_P(n^{-\frac{1}{4} - \varepsilon}) = o_P(n^{-\frac{1}{4} - \varepsilon}) = o_P(n^{-\frac{1}{4}}). \end{aligned}$$

There exists by the mean value theorem a $\xi_{k,x} \in (t_k, t_{k+1})$ such that $\frac{P(t_{k+1} | x) - P(t_k | x)}{t_{k+1} - t_k} = p(\xi_{k,x} | x)$ for μ_0 -almost all x . Furthermore, there exists also a $\xi'_{k,t,x}$ between t and $\xi_{k,x}$ such that $p(t | x) - p(\xi_{k,x} | x) = (t - \xi_{k,x})p'(\xi'_{k,t,x} | x)$. This implies that we can bound the second term on the right-hand side as

$$\begin{aligned} \|\tilde{p}_n - p\|_{\mu_0 \otimes m} &= \left\| \sum_{k=1}^{K_n} \mathbb{1}\{\cdot \in (t_k, t_{k+1}]\} p(\xi_{k,\cdot} | \cdot) - p(\cdot | \cdot) \right\|_{\mu_0 \otimes m} \\ &= \left\| \sum_{k=1}^{K_n} \mathbb{1}\{\cdot \in (t_k, t_{k+1}]\} (p(\xi_{k,\cdot} | \cdot) - p(\cdot | \cdot)) \right\|_{\mu_0 \otimes m} \\ &= \left\| \sum_{k=1}^{K_n} \mathbb{1}\{\cdot \in (t_k, t_{k+1}]\} (\cdot - \xi_{k,\cdot}) p'(\xi'_{k,\cdot} | \cdot) \right\|_{\mu_0 \otimes m} \quad (9) \\ &\leq K \sum_{k=1}^{K_n} (t_{k+1} - t_k) \sqrt{t_{k+1} - t_k} \\ &= K \sum_{k=1}^{K_n} (b(k))^{\frac{3}{2}} = o(n^{x_1 + \frac{3}{2}x_2}) = o(n^{-\frac{1}{4}}). \end{aligned}$$

so that we have

$$\|\hat{p}_n - p\|_{\mu_0 \otimes m} = o_P(n^{-\frac{1}{4}}). \quad (10)$$

□