

Lecture 3

distribution

$X_1, \dots, X_n \sim P_\eta, \eta \in H \leftarrow$ subset of a normed ~~line~~ space.
 $P_\eta \ll \mu$ having density p_η .

First

~~second~~ order Fréchet Taylor expansion:

$$x(\eta) = x(\hat{\eta}) + x'_{\hat{\eta}}(\eta - \hat{\eta}) + o_p(\|\eta - \hat{\eta}\|^2)$$

$$\begin{aligned} x'_{\hat{\eta}}(\eta - \hat{\eta}) &= \int x'_{\hat{\eta}}^{(1)}(dP_\eta - dP_{\hat{\eta}}) \\ &= \int x_{\hat{\eta}}^{(1)} dP_\eta \rightarrow \text{estimate with } P_n x_{\hat{\eta}}^{(1)} \end{aligned}$$

Alternative ~~the~~ way of looking at $\int x_{\hat{\eta}}^{(1)}$ is the following.

Consider the model $\mathcal{P} = \{P_\eta; \eta \in H\}$. A tangent set at P_η is the set of all \dot{g}_η s.t.

$$\dot{g}_\eta = \left. \frac{\partial}{\partial t} P_{\eta_t} \right|_{t=0} / P_\eta \quad (\text{scores at } P_\eta \text{ corresponding to the paths})$$

for some $\{P_{\eta_t}; t \in (-1, 1)\} \subset \mathcal{P}$ & $P_{\eta_0} = P_\eta$.

then, $x_{\hat{\eta}}^{(1)}$ satisfies $\in L_2^0(P_\eta)$ satisfies

$$\left. \frac{\partial}{\partial t} x(\eta_t) \right|_{t=0} = \int x_{\hat{\eta}}^{(1)} \dot{g}_\eta dP_\eta \quad \text{for any such paths \& corresponding}$$

(Actually all you need is

paths for which $\int \left(\frac{\sqrt{P_{\eta_t}} - \sqrt{P}}{t} - \frac{1}{2} \dot{g}_\eta \sqrt{P_\eta} \right)^2 d\mu \rightarrow 0$

$\Rightarrow \dot{g}_\eta \in L_2^0(P_\eta)$)

We now want carry forward this routine to a second order. #

second order Fréchet Representation,

$$x(\eta) = x(\hat{\eta}) + x_{\hat{\eta}}'(\eta - \hat{\eta}) + \frac{1}{2} x_{\hat{\eta}}''(\eta - \hat{\eta}, \eta - \hat{\eta}) + o_p(\|\eta - \hat{\eta}\|^3)$$

2nd order Von Mises Representation:

$$x_{\hat{\eta}}''(\eta - \hat{\eta}, \eta - \hat{\eta}) = \iint x_{\hat{\eta}}^{(2)}(x_1, x_2) dP_{\hat{\eta}}(x_1) dP_{\hat{\eta}}(x_2)$$

→ estimate by $U_n x_{\hat{\eta}}^{(2)}$ (will come back to what we mean by this later)

First: what does this mean in terms of 1-dimensional differentiable submodels

$\{P_{\eta_t}; t \in (-1, 1)\} \subseteq \mathcal{P}$ at P_{η} ?

Von-Mises Expansion implies

$$x(\eta_t) \approx x(\eta) + \int x_{\eta}^{(1)}(x_1) P_{\eta_t}(x_1) d\mu(x_1) + \frac{1}{2} \iint x_{\eta}^{(2)}(x_1, x_2) P_{\eta_t}(x_1) P_{\eta_t}(x_2) d\mu(x_1) d\mu(x_2)$$

$$= x(\eta) + \iint x_{\eta}^{(1)}(x_1) P_{\eta_t}(x_1) P_{\eta_t}(x_2) d\mu(x_1) d\mu(x_2)$$

$$+ \frac{1}{2} \iint x_{\eta}^{(2)}(x_1, x_2) P_{\eta_t}(x_1) P_{\eta_t}(x_2) d\mu(x_1) d\mu(x_2)$$

$$= x(\eta) + t \iint \left(x_{\eta}^{(1)}(x_1) + \frac{1}{2} x_{\eta}^{(2)}(x_1, x_2) \right) \frac{\frac{\partial}{\partial t} \prod_{i=1}^2 P_{\eta_t}(x_i)}{\prod_{i=1}^2 P_{\eta}(x_i)} dP_{\eta}(x_1) dP_{\eta}(x_2)$$

$$+ \frac{t^2}{2} \iint \left(x_{\eta}^{(1)}(x_1) + \frac{1}{2} x_{\eta}^{(2)}(x_1, x_2) \right) \frac{\frac{\partial^2}{\partial t^2} \prod_{i=1}^2 P_{\eta_t}(x_i)}{\prod_{i=1}^2 P_{\eta}(x_i)} dP_{\eta}(x_1) dP_{\eta}(x_2)$$

⇒ If we define a tangent set of order 2 as the set of all derivatives

$$\dot{g}_\eta(x_1, x_2) = \left. \frac{\partial}{\partial t} \prod_{i=1}^2 p_{\eta_t}(x_i) \right|_{t=0} / \prod_{i=1}^2 p_\eta(x_i)$$

$$\ddot{g}_\eta(x_1, x_2) = \left. \frac{\partial^2}{\partial t^2} \prod_{i=1}^2 p_{\eta_t}(x_i) \right|_{t=0} / \prod_{i=1}^2 p_\eta(x_i)$$

which arise from differentiable submodels

$(p_{\eta_t}; t \in (-1, 1))$, then $x_\eta^{(1)} + \frac{1}{2} x_\eta^{(2)}$ satisfies

$$\left. \frac{d}{dt} x(\eta_t) \right|_{t=0} = \left. \frac{d}{dt} p_{\eta_t}^2 \left(x_\eta^{(1)} + \frac{1}{2} x_\eta^{(2)} \right) \right|_{t=0} = p_\eta^2 \left(x_\eta^{(1)} + \frac{1}{2} x_\eta^{(2)} \right) \dot{g}_\eta$$

$$\left. \frac{d^2}{dt^2} x(\eta_t) \right|_{t=0} = \left. \frac{d^2}{dt^2} p_{\eta_t}^2 \left(x_\eta^{(1)} + \frac{1}{2} x_\eta^{(2)} \right) \right|_{t=0}$$

$$= p_\eta^2 \left(x_\eta^{(1)} + \frac{1}{2} x_\eta^{(2)} \right) \ddot{g}_\eta$$

Going out in this direction, we can think about going on doing such an expansion & provided we have von-Mises derivatives representation of the derivatives we would expect for $j=1, \dots, m$

$$\left. \frac{d^j}{dt^j} x(\eta_t) \right|_{t=0} = \left. \frac{d^j}{dt^j} p_{\eta_t}^m \left(x_\eta^{(1)} + \frac{1}{2} x_\eta^{(2)} + \dots + \frac{1}{m} x_\eta^{(m)} \right) \right|_{t=0}$$

$$= p_\eta^m \left(x_\eta^{(1)} + \frac{1}{2} x_\eta^{(2)} + \dots + \frac{1}{m} x_\eta^{(m)} \right) \overset{\uparrow}{g}_\eta^{(j \text{ dots})}$$

Also whenever

$\{\dot{g}_\eta, \ddot{g}_\eta, \dots, g_\eta^{(j)}\} = L_2^0(p_\eta^m)$ (which happens

$$\left. \frac{\partial^j}{\partial t^j} \prod_{i=1}^m p_{\eta_t}(x_i) \right|_{t=0}$$

$x_\eta^{(j)}(x_1, \dots, x_j)$ is the first order IF of

for models

$$\prod_{i=1}^m p_\eta(x_i)$$

restricted by smoothness/qualitative

$x_\eta^{(j-1)}(x_1, \dots, x_{j-1})$ at x_j i.e.

assumptions)

$$\frac{\partial}{\partial t} x_\eta^{(j-1)}(x_1, \dots, x_{j-1}) = \int x_\eta^{(j)}(x_1, \dots, x_{j-1}, x_j) \dot{g}_\eta(x_j) dp_\eta(x_j)$$

This will allow us to have an estimator

$$\hat{x}_n = x(\hat{\eta}) + O_n \left(x \hat{\eta}^{(1)} + \frac{1}{2} x \hat{\eta}^{(2)} + \dots + \frac{1}{m!} x \hat{\eta}^{(m)} \right)$$

having error of the order $O_p(\|\eta - \hat{\eta}\|^{m+1})$
 $+ O_p(1/\sqrt{n})$

Unfortunately there is no free lunch. We will try to understand through $x_1 \dots x_n \sim \eta$ on $[0,1]$ & $x(\eta) = \int \eta^2 d\mu$.

In this case the m th order tangent set at η is $L_2^0(\eta^m)$ (not too difficult to verify) ~~by fact~~

We will consider $m=2$ to begin with ~~and~~ and try to argue heuristically that ~~is~~ even a second order von Mises expansion fails. Suppose not. Last time we argued that

$$x_{\eta}^{(1)}(x_1) = 2(\eta(x_1) - x(\eta))$$

\uparrow this was used for centering
 actually $x_{\eta}^{(1)} = 2\eta$ works.

so ~~the~~ finding $x_{\eta}^{(2)}$ relies on finding the first order ~~the~~ influence function of $\eta \rightarrow 2\eta(x_2)$ at x_2 . This implies that

$$\frac{\partial}{\partial t} \eta_t(x_1) \Big|_{t=0} = \int x_{\eta}^{(2)}(x_1, x_2) \dot{g}_{\eta}(x_2) d\mu(x_2)$$

$\parallel \frac{\partial}{\partial t} \eta_t(x_2) \Big|_{t=0} = \frac{\dot{g}_{\eta}(x_2)}{\eta(x_2)}$
 for

$$\Rightarrow \dot{g}_{\eta}(x_1) \eta(x_1) = \int x_{\eta}^{(2)}(x_1, x_2) \dot{g}_{\eta}(x_2) \eta(x_2) d\mu(x_2) \quad \forall \dot{g}_{\eta}$$

~~$\dot{g}_{\eta} \in L_2(\eta)$~~ $\dot{g}_{\eta} \in L_2(\eta)$ $\dot{g}_{\eta} \sqrt{\eta} \in L_2(\mu)$ $\dot{g}_{\eta} \eta \in L_1(\mu)$
 $\sqrt{\eta} \in L_2(\mu)$

$$\dot{g}_\eta \in L_2(\eta)$$

$$\dot{g}_\eta \rightarrow \dot{g}_\eta(x_1) \eta(x_1)$$

$$L_2(\eta) \rightarrow \mathbb{R}$$

can be written as, $\int \chi_\eta^{(2)}(x_1, x_2) \dot{g}_\eta(x_2) d\mu_\eta(x_2)$

$$\langle \chi_\eta^{(2)}(x_1, \cdot) \dot{g}_\eta \rangle_{\mathcal{P}_\eta} \leftarrow \text{bounded linear functional}$$

However $\dot{g}_\eta \rightarrow \dot{g}_\eta(x_1) \eta(x_1)$ is not even continuous. functional over $L_2(\mathcal{P}_\eta)$

so such a representation does not exist. Do not seek such a representation for all \dot{g}_η indeed

~~$\dot{g}_\eta(x_1)$~~ if $L \subset L_2(\mu)$ is finite dimensional, $\exists \chi_L(x_1, x_2)$ s.t. $f(x_1) = \int \chi_L(x_1, x_2) f(x_2) d\mu(x_2)$ for all $f \in L$.

so if $\dot{g}_\eta \eta \in L$ then

$$\dot{g}_\eta(x_1) \eta(x_1) = \int \chi_L(x_1, x_2) \dot{g}_\eta(x_2) \eta(x_2) d\mu(x_2)$$

$\Rightarrow \chi_L(x_1, x_2)$ works in L .

let e_1, e_2, \dots be an o.n.b of $L_2(\mu)$ let

$$\Pi(x_1, x_2) = \sum_{l=1}^k e_l(x_1) e_l(x_2)$$

\Rightarrow For all $f \in \text{lin}\{e_1, \dots, e_k\}$, $f(x_1) = \int \Pi(x_1, x_2) f(x_2) d\mu(x_2)$

$$\text{let } \chi_{\eta*}^{(2)}(x_1, x_2) = 2\Pi(x_1, x_2)$$

$$\chi_\eta^{(2)}(x_1, x_2) = 2\Pi(x_1, x_2) - 2\Pi_{\mathcal{P}}(x_1) - 2\Pi_{\mathcal{P}}(x_2) + 2\left(\Pi_{\mathcal{P}}\right)^2$$

$$\hat{\chi}_n = \chi(\hat{\eta}) + \frac{1}{n} \sum_{i=1}^n \chi_{\hat{\eta}}^{(1)} + \frac{1}{2} \sum_{i=1}^n \chi_{\hat{\eta}}^{(2)}$$

$$= \chi(\hat{\eta}) + \frac{1}{n} \sum_{i=1}^n 2(\hat{\eta}(x_i) - \chi(\hat{\eta}))$$

$$+ \frac{1}{2} \frac{1}{n(n-1)} \sum_{i \neq j} 2 \left[\Pi(x_i, x_j) - 2 \int \Pi(x_i, x) \hat{\eta}(x) d\mu(x) - 2 \int \Pi(x_j, x) \hat{\eta}(x) d\mu(x) + 2 \int (\Pi \hat{\eta})^2 d\mu(x) \right] *$$

$$= \chi(\hat{\eta}) + \frac{1}{n} \sum_{i=1}^n 2(\hat{\eta}(x_i) - \chi(\hat{\eta}))$$

$$\equiv \int \Pi(x_1, x_2) \hat{\eta}(x_2) d\mu(x_2)$$

$$= \int \sum_{\ell=1}^k e_{\ell}(x) e_{\ell}(x_2) \hat{\eta}(x_2) d\mu(x_2)$$

$$= \sum_{\ell=1}^k \int e_{\ell}(x) e_{\ell}(x_2) \hat{\eta}(x_2) d\mu(x_2)$$

$$\equiv \hat{\eta} = \sum_{\ell=1}^k e_{\ell}(x) \left(\frac{1}{n} \sum_{i=1}^n e_{\ell}(x_i) \right)$$

$$\Rightarrow \hat{\eta} \in \text{lin} \{e_1, \dots, e_k\}$$

$$(*) \hat{\chi}_n = \frac{1}{n} \sum_{i \neq j} \Pi(x_i, x_j) = \frac{1}{n} \sum_{i \neq j} \left(\sum_{\ell=1}^k e_{\ell}(x_i) e_{\ell}(x_j) \right)$$

We shall now study properties of this estimator.