

Posterior consistency:

$$D^n \sim p_n(d^n | \theta), \theta \in \Theta, \theta \sim \pi_n$$

$$\pi_n(B | D^n) = \frac{\int_B p_n(D^n | \theta) d\pi_n(\theta)}{\int_{\Theta} p_n(D^n | \theta) d\pi_n(\theta)}, B \subset \Theta$$

$$p_n(d^n | \theta) = \prod_{i=1}^n f(x_i | \theta)$$

$$\text{Expect: } \pi_n(\cdot | D^n) \xrightarrow{\text{weakly}} \delta_{\theta_0}$$

$$f_n(\theta) := d_w(\pi_n(\cdot | D^n), \delta_{\theta_0})$$

d_w = Wasserstein metric

Def. 1: The posterior sequence is consistent at θ_0 if $f_n(\theta_0) \rightarrow 0$ a.s. or in prob. when $D^n \sim p_n(\cdot | \theta_0)$.

Lemma 1: If (Θ, d) is separable, $f_n(\theta_0) \rightarrow 0$ a.s. or i.p. iff \forall open neighborhood U of θ_0 , $\pi_n(U^c | D^n) \rightarrow 0$ a.s./i.p.

The Schwartz theorem:

$$D^n = (X_1, \dots, X_n); X_i \stackrel{iid}{\sim} f, f \in \Pi$$
$$d_{KL}(p, q) = \int p(x) \log\left(\frac{p(x)}{q(x)}\right) dx$$

Test fcn: $\Phi_n: \mathcal{X}^n \rightarrow [0, 1]$.

$f = f_0$ is true density. p_0^∞ joint density of (X_1, \dots, X_n) . f_0 belongs to the KL support of \mathbb{P} if

$$\forall \varepsilon > 0, \mathbb{P}(f: d_{KL}(f_0, f) < \varepsilon) > 0.$$

Theorem: If f_0 belongs to the KL support of \mathbb{P} + $U_n \subset \mathcal{F}$ are neighborhoods of f_0 s.t. Φ_n with

$$\mathbb{E}_{f_0} \Phi_n \leq B e^{-bn} \quad \text{Type I error}$$

$$\sup_{f \in U_n^c} \mathbb{E}_f (1 - \Phi_n) \leq B e^{-bn}, \quad \text{Type II error}$$

for $b, B > 0$ then $\mathbb{P}(U_n^c | P_n) \rightarrow 0$

a.s. p_0^∞ ;

Technical extension (for infinite dim models)

A sequence of compact subsets $F_n \subset F$ are called sieves if:

The metric d on F is bounded from above by d_H . Fix $\epsilon > 0$, $F_n \subset F$ and $\delta > 0$, $0 < \xi < \frac{1}{2}$, $C > 0$ s.t.

$$1) \log(N(\delta, F_n, d)) \leq \xi n \epsilon^2$$

$$2) \Pi(F_n^c) \leq e^{-Cn}.$$

What about non-iid, dynamical systems?

There is a variational form of posterior consistency based on large deviation theory.

non-iid models:

$$\text{Markov: } X_{t+1} = f(X_t; \theta)$$

$$\begin{aligned} \text{HMM: } X_{t+1} &= f(X_t; \theta) && : \text{state process} \\ Y_{t+1} &= g(X_{t+1}; \theta) && : \text{obs. process} \end{aligned}$$

Deterministic dynamics - $\theta \in \Theta$ we have $(X, \mathcal{X}, T_\theta, \mu_\theta)$

- X is complete separable metric space with Borel σ -algebra \mathcal{X}
- $T_\theta: X \rightarrow X$ is a measurable map
- μ_θ on (X, \mathcal{X}) is T_θ -invariant if $\mu_\theta(T_\theta^{-1}A) = \mu_\theta(A) \quad \forall A \in \mathcal{X}$
- the system is ergodic: $T_\theta^{-1}A = A \Rightarrow \mu(A) = \{0, 1\}$

$$(X, \mathcal{X}, T_\theta, \mu_\theta)_{\theta \in \Theta} = (T_\theta, \mu_\theta)_{\theta \in \Theta}$$

Observation process:

$$\int g_{\theta}(y|x) d\nu(y) = 1$$

$$g: \Theta \times X \times Y \rightarrow \mathbb{R}_+$$

$$p_{\theta}(y_0^n | x) = \prod_{k=0}^n g_{\theta}(y_k | T_{\theta}^k(x))$$

$$p_{\theta}(y_0^n) = \int p_{\theta}(y_0^n | x) d\mu_{\theta}(x)$$

$$X_0 \sim U[0,1]$$

$$x_{t+1} = \theta x_t (1 - x_t)$$

$$Y_{t+1} = N(x_{t+1}, \sigma^2)$$

Generalized Baye set-up

a) Observation system (Y, T, ν) ,
 $T: Y \rightarrow Y$

b) Tracking systems: $X := X \times \Theta$,
 $S: \Theta \times X \rightarrow X$, $S_{\theta}: X \rightarrow X$

c) loss: $X \times Y \rightarrow \mathbb{R}_+$

$$l_n(x, y; \theta) := l_n(x_0^{n-1}, y_0^{n-1}) \\ = \sum_{k=0}^{n-1} l(x_k, y_k)$$

$$x_0^{n-1} = (x, S_\theta x, \dots, S_\theta^{n-1}(x))$$

$$y_0^{n-1} = (y, T y, \dots, T^{n-1} y)$$

Gibbs posterior -

$$\Pi_n(A | y) = \frac{\int_A \exp(-l_n(x, y; \theta)) d\pi(x)}{Z_n(y)}$$

$$Z_n(y) = \int_X \exp(-l_n(x, y; \theta)) d\pi(x)$$

(1) is $\lim_{n \rightarrow \infty} \Pi_n(\cdot | y)$ unique

(2) does it concentrate around T

$$\mathbb{P}_n(A|y) = \frac{\int_A \exp(-\psi \ell_n(x, y; \theta)) d\theta(y)}{Z_n(y)}$$

Sequence space model

A is a finite set ($|A| = N$), $\Sigma = A^{\mathbb{Z}}$

Left shift: $\sigma: A^{\mathbb{Z}} \rightarrow A^{\mathbb{Z}}$ + $(\sigma x)_i = x_{i+1}$

$$\Sigma_{\mathcal{F}} = \left\{ x \in A^{\mathbb{Z}} \mid x_{[i, j]} \neq u \quad \forall i, j \in \mathbb{Z}, \right. \\ \left. u \in \mathcal{F} \right\}$$

forbids a finite # of words \mathcal{F}

Shift of finite type (SFT)

Note: model system to study systems like axiom A

Can encode $\Sigma_{\mathcal{F}}$ by a matrix A :

$$\Sigma_A = \left\{ (a_i)_{i=-\infty}^{\infty} \in \Sigma_{\mathcal{F}}, A_{a_i, a_{i+1}} = 1 \quad \forall i \in \mathbb{Z} \right\}$$

a topological Markov chain or

1-step SFT

Σ_A is mixing iff $\exists n \geq 1$ s.t.
 A^n consists of all positive entries.

Gibbs measure:

Let $f: \Sigma_f \rightarrow \mathbb{R}$ be continuous.

A measure μ on Σ_f has the Gibbs property for f if there exists $K > 1$ and $P \in \mathbb{R}$ s.t. \forall

$x \in A^{\mathbb{Z}}$ and $m \geq 1$

$$K^{-1} \leq \frac{\mu(x[0, m-1])}{\exp(-Pm + \sum_{k=0}^{m-1} f(\sigma^k(x)))} \leq K.$$

Theorem (Bowen):

If Σ_f is a mixing SFT and $f: \Sigma_f \rightarrow \mathbb{R}$ is Hölder continuous then there exists a unique Gibbs measure on Σ_f .

$f: \Sigma_f \rightarrow \mathbb{R}$ is called the potential

$P = P(f)$ is its pressure.

Model class:

Let Θ be a compact metric space.

Let $\{f_\theta : \theta \in \Theta\}$ be a continuous parametrized family of Hölder continuous potential fcn's.

Let $\{\mu_\theta : \theta \in \Theta\}$ be the corresponding family of Gibbs measures.

Markov chains are an example.

Observation process: Let λ be a Borel measure on γ

Let $g : \Theta \times X \times \gamma \rightarrow [0, \infty)$ be a measurable fcn. s.t. $\forall \theta \in \Theta$ and $x \in X$

$$\int g(\theta, x, \gamma) \lambda(d\gamma) = 1$$

$$g_\theta(\cdot | x) = g(\theta, x, \cdot)$$

Hidden Gibbs process:

$$P_\theta(\gamma_0^{n-1}) = \int \prod_{k=0}^{n-1} g_\theta(\gamma_k | \sigma^k(x)) \mu_\theta(dx)$$

$$X_0 \sim \mu_\theta, X_{n+1} = \sigma(X_n), \gamma_n \sim g_\theta(\gamma | X_n) \lambda(d\gamma)$$

P_θ^Y denote the distribution of $\{Y_n\}_{n \geq 0}$ under θ .

For $\theta \in \Theta$, let $[\theta] = \{ \theta' \in \Theta : P_\theta^Y = P_{\theta'}^Y \}$

Theorem: Suppose π is fully supported on Θ , let $\theta_0 \in \Theta$. Then for any neighborhood U of $[\theta_0]$

$$\pi_n(\Theta \setminus U | Y_0^{n-1}) \rightarrow 0, P_{\theta_0}^Y \text{-a.s.}$$

$$\hat{\pi}_n(A | Y_0^{n-1}) = \frac{\int_A \exp(-\ell(\theta, x; Y_0^{n-1})) P_0(d\tau, dx)}{Z_n(Y_0^{n-1})}$$

Questions:

1. What is $\lim_n \frac{1}{n} \log Z_n(Y_0^{n-1})$
2. Convergence of $\{\hat{\pi}_n\}$

Coupling: A coupling of two r.v.'s X, X' taking values in (E, \mathcal{E}) is any pair of r.v.'s (Y, Y') in $(E \times E, \mathcal{E} \times \mathcal{E})$ with $X \stackrel{D}{=} Y$, $X' \stackrel{D}{=} Y'$

Joining: Let (X, A, μ, T) and (Y, B, ν, S) be two dynamical systems. A joining of S, T is a prob. measure λ on $X \times Y$ with marginals μ, ν , and invariant to the product map $T \times S$.

A stationary X -valued process $\{X_n\}_{n \geq 0}$ is in $\mathcal{P}(X, \sigma)$ if

$$X_{n+1} = \sigma(X_n), \quad \forall n \text{ w.p. } 1.$$

A joining of (X, σ) with $\{Y_n\}_{n \geq 0}$ is a stationary bi-variate process $(U, V) = \{(U_n, V_n)\}_{n \geq 0}$ on $X \times Y$ s.t.

$$U = \{U_n\}_{n \geq 0} \text{ is in } \mathcal{P}(X, \sigma)$$

$$V = \{U_n\}_{n \geq 0} \stackrel{d}{=} \{Y_n\}_{n \geq 0}$$

We denote the set of joinings of (X, σ) with $\{Y_n\}$ as \mathcal{J}

Theorem. There exists a lower semicontinuous fcn. $\mathcal{G}: \Theta \rightarrow \mathbb{R}$ s.t. w.p. 1

$$\lim_n -\frac{1}{n} \log Z_n(\gamma) = \inf_{\theta \in \Theta} \mathcal{G}(\theta)$$

The r.h.s. is the rate fcn. in the large deviation sense.

Limiting average cost:

$$\lim_{n \rightarrow \infty} \frac{1}{n} \int_X \ell_n(x, \gamma) d\lambda_\gamma(x) = \int \ell d\lambda$$

Entropy \doteq measures π, μ on X , finite measurable partition \mathcal{I}

$$\mu \ll_{\mathcal{I}} \pi \Rightarrow \pi(c) = 0 \Rightarrow \mu(c) = 0 \text{ for } c \in \mathcal{I}$$

$$L(\mu || \pi, \mathcal{I}) = \begin{cases} \sum_{c \in \mathcal{I}} \mu(c) \log \pi(c) \\ -\infty \end{cases}$$

$$F(\mu, \pi) = \sup_{\mathcal{I}} L(\mu || \pi, \mathcal{I})$$

Theorem. Given a Gibbs prior

$$\lim_{n \rightarrow \infty} -\frac{1}{n} \log Z_n(\gamma) = \inf_{\lambda \in \mathcal{T}} \left\{ \int \ell d\lambda + F(\lambda, \mu_0) \right\}$$

recall

$$\pi^n(\theta | x) = \arg \min_{\mu} \left\{ \int_{\mathcal{D}} \ell(\theta, \alpha) d\mu(\theta) + d_{KL}(\mu, \pi) \right\}$$

Prop. : Suppose a Gibbs prior and pressure

$$P = \inf_{\lambda \in \mathcal{T}} \left\{ \int \ell d\lambda + F(\lambda, \mu_0) \right\}$$

$$\Theta_n = \arg \min_{\mathcal{D} \in \mathcal{D}} P$$

For all $\varepsilon > 0$

$$P(d(S_{\Theta_n, T}) < \varepsilon) \rightarrow 1 \text{ a.s. } n \rightarrow \infty$$

Large deviations approach:

Two conditions need to hold for post-consistency

1) Conditional large deviations for one process on $X \times Y$

2) Prove exponential continuity over the map $\theta \mapsto \mu_\theta$; allows us to extend the LDP for one family of all of θ .

Given a Polish space Z and a lower semicontinuous fcn. $I: Z \rightarrow [0, \infty]$.

A family $(\eta_t)_{t \in T}$ of prob measures satisfies the large deviations principle with rate fcn I if for every closed set $E \in Z$

$$\limsup_{t \rightarrow \infty} \frac{1}{t} \log \eta_t(E) \leq \inf_{z \in E} I(z)$$

and open set $V \subset Z$

$$\liminf_{t \rightarrow \infty} \frac{1}{t} \log \eta_t(V) \geq - \inf_{z \in V} I(z)$$

Step 4: For a fixed θ show

$$\lim_{t \rightarrow \infty} \frac{1}{t} \log Z_t^\theta(\gamma) = \inf_{\lambda \in \mathcal{J}(s; \nu)} \left\{ \int \lambda d\lambda + F(\lambda, \pi) \right\} \\ = -V(\theta).$$

Exponential continuity:

$\{\mu_\theta\}_\theta$ is exponentially continuous w.r.t. \mathcal{Q} if the following holds.

For all $\theta \in \Theta$ it holds for ν -a.e. $\gamma \in \mathcal{Y}$ the following limit exists

$$\lim_{t \rightarrow \infty} \frac{1}{t} \log \int_X \exp(-l_\theta^t(x, \gamma)) d\mu_\theta(x) \\ =: -V(\theta)$$

and if $(\theta_t)_{t \in \mathbb{T}}$ is a family of parameters s.t. $\theta_t \rightarrow \theta$ in Θ then

$$\lim_{t \rightarrow \infty} \log \int_X \exp(-l_{\theta_t}^t(x, \gamma)) d\mu_{\theta_t}(x) = -V(\theta)$$

Prop. $\{\mu_\theta\}_{\theta \in \Theta}$ is an exponentially continuous family w.r.t. l and π is a Borel measure on Θ . For ν -almost every $y \in \mathcal{Y}$

$$\lim_{t \rightarrow \infty} -\frac{1}{t} \log Z_t^\pi(y) = \inf_{\theta \in \text{supp}(\pi)} V(\theta)$$