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What is asymptotically testable and what is not?

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Part I

Introduction and Motivation

Asymptotic symmetric testing

Observe *i.i.d.* $X^n \sim P^n$, model $P \in \mathcal{P}$. For disjoint $B, V \subset \mathcal{P}$,

$$H_0 : P \in B, \quad \text{or} \quad H_1 : P \in V.$$

Tests $\phi_n : \mathcal{X}^n \rightarrow [0, 1]$; asymptotically, require:

$$\text{(type-I)} \quad P^n \phi_n \rightarrow 0 \text{ for } P \in B, \text{ and,}$$

$$\text{(type-II)} \quad P^n (1 - \phi_n) \rightarrow 0 \text{ for } P \in V.$$

Equivalently, we want,

A testing procedure that chooses for B or V based on X^n for every $n \geq 1$, has **property (D)** if it is wrong only a finite number of times with P^∞ -probability one.

Property (D) is sometimes referred to as “discernibility”.

Some examples and unexpected answers (I)

Consider non-parametric regression with $f : X \rightarrow \mathbb{R}$ and test for smoothness,

$$H_0 : f \in C^1(X \rightarrow \mathbb{R}), \quad H_1 : f \in C^2(X \rightarrow \mathbb{R}),$$

Consider a non-parametric density estimation with $p : \mathbb{R} \rightarrow [0, \infty)$ and test for square-integrability,

$$H_0 : \int x^2 p(x) dx < \infty, \quad H_1 : \int x^2 p(x) dx = \infty.$$

Practical problem we cannot use the data to determine with asymptotic certainty, if CLT applies with our data.

Some examples and unexpected answers (II)

Coin-flip $X^n \sim \text{Bernoulli}(p)^n$ with $p \in [0, 1]$.

Consider Cover's **rational mean problem**: test for rationality:

$$H_0 : p \in [0, 1] \cap \mathbb{Q}, \quad H_1 : p \in [0, 1] \setminus \mathbb{Q}.$$

Consider also Dembo and Peres's **irrational alternative**:

$$H_0 : p \in [0, 1] \cap \mathbb{Q}, \quad H_1 : p \in [0, 1] \cap \sqrt{2} + \mathbb{Q},$$

Consider ultimately **fractal hypotheses**, e.g. with Cantor set C ,

$$H_0 : p \in C, \quad H_1 : p \in [0, 1] \setminus C.$$

The Le Cam-Schwartz theorem

Theorem 6.1 (Le Cam-Schwartz, 1960) *Let \mathcal{P} be a model for i.i.d. data X^n with disjoint subsets B, V . The following are equivalent:*

- i. there exist (uniformly) consistent tests for B vs V ,*
- ii. there is a sequence of \mathcal{U}_∞ -uniformly continuous $\psi_n : \mathcal{P} \rightarrow [0, 1]$,*

$$\psi_n(P) \rightarrow 1_V(P), \quad (1)$$

(uniformly) for all $P \in B \cup V$.

Topological context uniform space $(\mathcal{P}, \mathcal{U}_\infty)$.

The Dembo-Peres theorem

Theorem 7.1 (Dembo and Peres, 1995) *Let \mathcal{P} be a model dominated by Lebesgue measure μ for i.i.d. data X^n . Model subsets B, V that are contained in disjoint countable unions of closed sets for Prokhorov's weak topology have tests with property (D). If there exists an $\alpha > 1$ such that $\int (dP/d\mu)^\alpha d\mu < \infty$ for all $P \in \mathcal{P}$, then the converse is also true.*

Topological context L^1 -weakly compact, dominated model \mathcal{P} with Prokhorov's weak topology.

Three forms of testability

Definition 8.1 (ϕ_n) is a *uniform test sequence* for B vs V , if,

$$\sup_{P \in B} P^n \phi_n \rightarrow 0, \quad \sup_{Q \in V} Q^n (1 - \phi_n) \rightarrow 0. \quad (2)$$

Definition 8.2 (ϕ_n) is a *pointwise test sequence* for B vs V , if,

$$\phi_n(X^n) \xrightarrow{P} 0, \quad \phi_n(X^n) \xrightarrow{Q} 1, \quad (3)$$

for *all* $P \in B$ and $Q \in V$.

Definition 8.3 (ϕ_n) is a *Bayesian test sequence* for B vs V , if,

$$\phi_n(X^n) \xrightarrow{P} 0, \quad \phi_n(X^n) \xrightarrow{Q} 1, \quad (4)$$

for Π -almost-all $P \in B$ and $Q \in V$.

Questions

Existence

Existence of uniform tests?

Existence of pointwise tests?

Existence of Bayesian tests?

Construction

How does one model-select? Are there constructive solutions?

Examples

Select the correct directed, acyclical graph in a graphical model;
select the right number of clusters in a clustering model.

Part II

Existence

Uniform testability has exponential power

Proposition 11.1 *Let \mathcal{P} be a model for i.i.d. data with disjoint B and V . The following are equivalent:*

i. *there exists a uniform test sequence (ϕ_n) ,*

$$\sup_{P \in B} P^n \phi_n \rightarrow 0, \quad \sup_{Q \in V} Q^n (1 - \phi_n) \rightarrow 0,$$

ii. *there is an exponentially powerful uniform test sequence (ψ_n) , i.e. there is a $D > 0$ such that,*

$$\sup_{P \in B} P^n \psi_n \leq e^{-nD}, \quad \sup_{Q \in V} Q^n (1 - \psi_n) \leq e^{-nD}.$$

The model as a uniform space (I)

Take \mathcal{X} a separable metrizable space, with Borel σ -algebra \mathcal{B} .

The class \mathcal{F}_n contains all bounded, \mathcal{B}^n -measurable $f : \mathcal{X}^n \rightarrow \mathbb{R}$.

For every $n \geq 1$ and $f \in \mathcal{F}_n$, define the entourage,

$$W_{n,f} = \{(P, Q) \in \mathcal{P} \times \mathcal{P} : |P^n f - Q^n f| < 1\}.$$

Defines uniformity \mathcal{U}_n (with topology \mathcal{I}_n). Take $\mathcal{U}_\infty = \bigcup_{n \geq 1} \mathcal{U}_n$.

$$P \rightarrow Q \text{ in } \mathcal{I}_\infty \quad \Leftrightarrow \quad \int f dP^n \rightarrow \int f dQ^n,$$

for all $n \geq 1$ and all $f \in \mathcal{F}_n$. Note also,

$$\mathcal{U}_C \subset \mathcal{U}_1 \subset \cdots \subset \mathcal{U}_\infty \subset \mathcal{U}_{TV}.$$

The model as a uniform space (II)

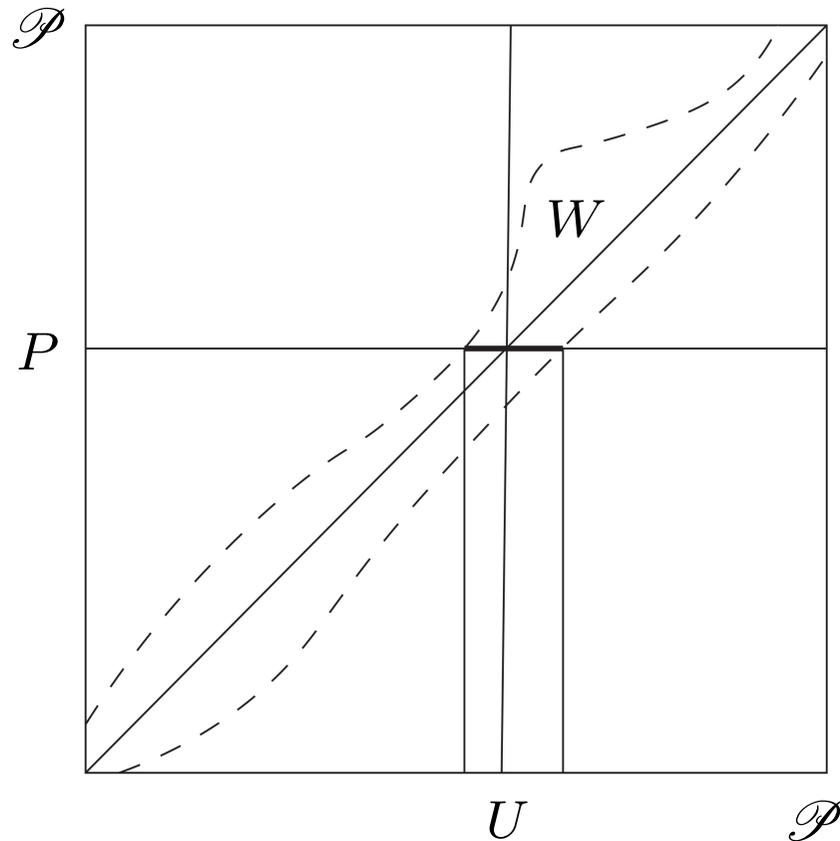


Fig 1. Let $P \in \mathcal{P}$ and entourage W be given. A neighbourhood U corresponds to $U = \{Q \in \mathcal{P} : (Q, P) \in W\}$

Uniform separation (I)

Definition 14.1 Subsets $B, V \subset \mathcal{P}$ are *uniformly separated by \mathcal{U}_∞* , if there exists an entourage $W \in \mathcal{U}_\infty$ such that,

$$(B \times V \cup V \times B) \cap W = \emptyset.$$

In other words, there are $J, m \geq 1$, $\epsilon > 0$ and bounded, measurable functions $f_1, \dots, f_J : \mathcal{X}^m \rightarrow [0, 1]$ such that, for any $P, Q \in B \cup V$, if,

$$\max_{1 \leq j \leq J} |P^m f_j - Q^m f_j| < \epsilon,$$

then *either $P, Q \in B$, or $P, Q \in V$* . (If the model is \mathcal{T}_∞ -compact, $m = 1$ suffices).

Uniform separation (II)

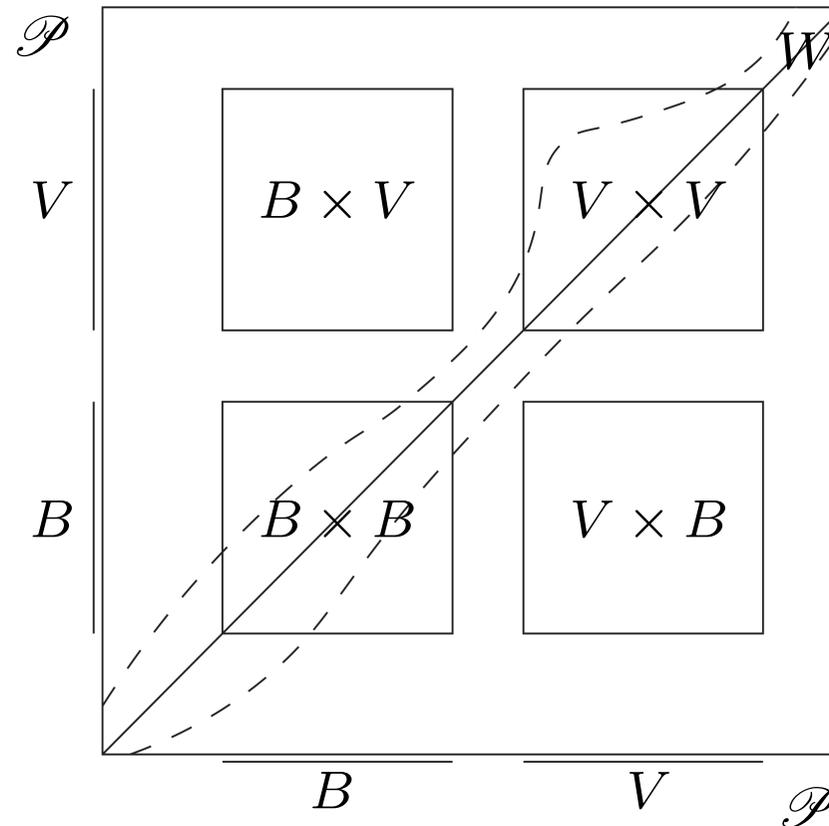


Fig 2. Let $B, V \subset \mathcal{P}$ and entourage W be given. W separates B and V if $B \times V$ and $V \times B$ do not meet W .

Characterisation of uniform testability

Theorem 16.1 *Let \mathcal{P} be a model for i.i.d. data with disjoint B and V . The following are equivalent:*

- (i.) *there exist uniform tests ϕ_n for B versus V ,*
- (ii.) *the subsets B and V are uniformly separated by \mathcal{U}_∞ .*

Corollary 16.2 (Parametrised models) *Suppose $\mathcal{P} = \{p_\theta : \theta \in \Theta\}$, with (Θ, d) compact, metric space and $\theta \rightarrow P_\theta$ identifiable and \mathcal{I}_∞ -continuous, (that is, for every $f \in \mathcal{F}_n$, $\theta \mapsto \int f dP_\theta^n$ is continuous). If $B_0, V_0 \subset \Theta$ with $d(B_0, V_0) > 0$, then the images $B = \{P_\theta : \theta \in B_0\}$, $V = \{P_\theta : \theta \in V_0\}$ are uniformly testable.*

Closures are important

Proposition 17.1 Let \mathcal{P} be a model for i.i.d. data and let B, V be disjoint model subsets with \mathcal{I}_∞ -closures \bar{B} and \bar{V} . If B, V are uniformly separated by \mathcal{U}_∞ , then $\bar{B} \cap \bar{V} = \emptyset$. If \mathcal{P} is relatively \mathcal{I}_∞ -compact, the converse is also true.

Theorem 17.2 (Dunford-Pettis) Assume \mathcal{P} is dominated by a probability measure Q with densities in $\mathcal{P}_Q \subset L^1(Q)$; \mathcal{P}_Q is relatively weakly compact, if and only if, for every $\epsilon > 0$ there is an $M > 0$ such that,

$$\sup_{P \in \mathcal{P}} \int_{\{dP/dQ > M\}} \frac{dP}{dQ} dQ < \epsilon,$$

that is, \mathcal{P}_Q is uniformly Q -integrable.

Pointwise testability: equivalent formulations

Proposition 18.1 *Let \mathcal{P} be a model for i.i.d. data and let B, V be disjoint model subsets. The following are equivalent:*

i. *there are tests (ϕ_n) such that, for all $P \in B$ and $Q \in V$,*

$$P^n \phi_n \rightarrow 0, \quad Q^n (1 - \phi_n) \rightarrow 0,$$

ii. *there are tests (ϕ_n) such that, for all $P \in B$ and $Q \in V$,*

$$\phi_n(X^n) \xrightarrow{P} 0, \quad (1 - \phi_n(X^n)) \xrightarrow{Q} 0,$$

iii. *there are tests (ϕ_n) such that, for all $P \in B$ and $Q \in V$,*

$$\phi_n(X^n) \xrightarrow{P\text{-a.s.}} 0, \quad (1 - \phi_n(X^n)) \xrightarrow{Q\text{-a.s.}} 0.$$

Pointwise testability from consistent estimators

Consistent estimators $\hat{P}_n : \mathcal{X}^n \rightarrow \mathcal{P}$: for all P and nbd U of P ,

$$P^n(\hat{P}_n(X^n) \in U) \rightarrow 1, \text{ as } n \rightarrow \infty.$$

For open $B, V \subset \mathcal{P}$, define $\phi_n(X^n) = \mathbf{1}\{\hat{P}_n \in V\}$. For any $P \in B$, B is a neighbourhood of P so $P^n\phi_n = P^n(\hat{P}_n \in V) \leq P^n(\hat{P}_n \notin B) \rightarrow 0$. For any $Q \in V$, $Q^n(1 - \phi_n) \rightarrow 0$. So (ϕ_n) is a **pointwise test sequence** for B vs V .

Restrict to $\mathcal{P}' = B \cup V$, then B and V are *clopen sets*.

Proposition 19.1 *If $P \in \mathcal{P}$ can be estimated consistently and B is clopen, there exist **pointwise tests** for B vs its complement.*

Necessary conditions: pointwise non-testability

Suppose that there exist pointwise tests (ϕ_n) for B, V . Define,

$$g_n : \mathcal{P} \rightarrow [0, 1] : P \mapsto P^n \phi_n,$$

which are all \mathcal{U}_∞ -uniformly continuous.

Proposition 20.1 *If there is a pointwise test (ϕ_n) for B vs V , then B, V are both G_δ - and F_σ -sets with respect to \mathcal{T}_∞ (in the subspace $B \cup V$).*

Corollary 20.2 *Suppose $\mathcal{P} = B \cup V$ is Polish in the \mathcal{T}_∞ -topology. Pairs B, V that are pointwise testable, are both Polish spaces.*

Corollary 20.3 *If there exists a Baire subspace D of \mathcal{P} in which both $D \cap B$ and $D \cap V$ are dense, then B is not testable versus V .*

Pointwise non-testability: examples (I)

Example 21.1 *Is Cover's rational means problem testable?*

Dunford-Pettis theorem shows that \mathcal{P} is \mathcal{I}_∞ -compact and $[0, 1] \rightarrow \mathcal{P} : p \mapsto P_p$ is a \mathcal{I}_∞ -homeomorphism. Since $[0, 1]$ is a complete metric space, \mathcal{P} is a Baire space for the \mathcal{I}_∞ -topology. Because both $[0, 1] \cap \mathbb{Q}$ and $[0, 1] \setminus \mathbb{Q}$ are dense in $[0, 1]$, the images $\mathcal{P}_0 := \{P_p : p \in [0, 1] \cap \mathbb{Q}\}$ and $\mathcal{P}_1 := \{P_p : p \in [0, 1] \setminus \mathbb{Q}\}$ are \mathcal{I}_∞ -dense in \mathcal{P} : there is no pointwise test for $p \in [0, 1] \cap \mathbb{Q}$ versus $p \in [0, 1] \setminus \mathbb{Q}$.

Pointwise non-testability: examples (II)

Example 22.1 *Is Dembo and Peres's irrational alternative testable?*

Any countable \mathcal{P} is Polish in the discrete topology. Any subset B of \mathcal{P} is a countable union of closed sets ($B = \cup_{b \in B} \{b\}$), so it remains possible that there exists a pointwise test for Dembo and Peres's problem.

Example 22.2 *Is Cantor's fractal alternative testable?*

The interval $[0, 1]$ is Polish and \mathcal{P} is homeomorphic. The Cantor set C is closed and its complement is open. Open sets in metrizable spaces are F_σ -sets. So it remains possible there exists a pointwise test for Cantor's fractal alternative.

Pointwise non-testability: examples (III)

Example 23.1 *Is integrability of a real-valued X , $P|X| < \infty$, testable?*

Model $\mathcal{P} = \{\text{all probability distributions on } \mathbb{R}\}$. \mathcal{P} is *Baire space* for \mathcal{I}_{TV} . Define,

$$B = \{P \in \mathcal{P} : P|X| < \infty\}, \quad V = \{P \in \mathcal{P} : P|X| = \infty\}.$$

B cannot be tested versus V .

Namely Let $P \in B$ and $Q \in V$ be given. For any $0 < \epsilon < 1$, $P' = (1 - \epsilon)P + \epsilon Q$ satisfies $\|P' - P\| = \epsilon\|(P + Q)\| \leq 2\epsilon$, but $P' \in V$. Conclude that V lies \mathcal{I}_{TV} -dense in \mathcal{P} .

Conversely, Q is tight, so for every $\epsilon > 0$, there exists an $M > 0$ such that $|Q(A) - Q(A||X| \leq M)| < \epsilon$ for all measurable $A \subset \mathbb{R}$. Since $Q(\cdot||X| \leq M) \in B$, we also see that B lies \mathcal{I}_{TV} -dense in \mathcal{P} .

Pointwise testability in dominated models

Definition 24.1 *The testing problem has a (uniform) representation on X , if there exists a \mathcal{T}_∞ -(uniformly-)continuous, surjective map $f : B \cup V \rightarrow X$ such that $f(B) \cap f(V) = \emptyset$.*

Definition 24.2 *The model is parametrised by Θ , if there exists a \mathcal{T}_∞ -continuous bijection $P : \Theta \rightarrow \mathcal{P}$ (i.e. for every $m \geq 1$ and measurable $f : \mathcal{X}^m \rightarrow [0, 1]$, the map $\theta \mapsto \int f dP_\theta^m$ is continuous).*

If Θ is compact, any parametrization is a homeomorphism, so the inverse gives rise to representations of testing problems in \mathcal{P} .

Characterisation of pointwise testability

Theorem 25.1 *Let \mathcal{P} be a **dominated** model for i.i.d. data with disjoint B, V . The following are equivalent,*

- i. there exists a **pointwise test** for B vs V ,*
- ii. the problem has a **representation** $f : B \cup V \rightarrow X$ on a **normal space** X and there exist disjoint F_σ -sets $B', V' \subset X$ such that $f(B) \subset B', f(V) \subset V'$,*
- iii. the problem has a **uniform representation** $\psi : B \cup V \rightarrow X$ on a **separable, metrizable space** X with $\psi(B), \psi(V)$ both F_σ - and G_δ -sets.*

Pointwise testability: corollaries (I)

Corollary 26.1 *Suppose that \mathcal{P} is **dominated** and there exist disjoint F_σ -sets B', V' in the **completion** $\hat{\mathcal{P}}$ (for \mathcal{U}_∞) with $B \subset B', V \subset V'$. Then B is **pointwise testable versus** V .*

Corollary 26.2 *Suppose that \mathcal{P} is **dominated** and **complete** (for \mathcal{U}_∞) with disjoint subsets B, V . Then B is **pointwise testable versus** V , if and only if, there exist disjoint F_σ -sets $B', V' \subset \mathcal{P}$ with $B \subset B', V \subset V'$.*

Pointwise testability: corollaries (II)

Corollary 27.1 *Suppose that \mathcal{P} is dominated and TV-totally-bounded. Then disjoint $B, V \subset \mathcal{P}$ are pointwise testable, if and only if, B, V are both F_σ - and G_δ -sets in $B \cup V$ (for \mathcal{I}_{TV}).*

Corollary 27.2 *Suppose that \mathcal{P} is dominated by a probability measure, with a uniformly integrable family of densities. Then disjoint $B, V \subset \mathcal{P}$ are pointwise testable, if and only if, B, V are both F_σ - and G_δ -sets in $B \cup V$ (for \mathcal{I}_C).*

Pointwise testability: examples (I)

Example 28.1 *Is independence of two events A and B testable?*

Let $A, B \in \mathcal{B}$ be msb subsets. Consider,

$$H_0 : P(A \cap B) = P(A)P(B), \quad H_1 : P(A \cap B) \neq P(A)P(B).$$

Define \mathcal{U}_1 -continuous $f_i : \mathcal{P} \rightarrow [0, 1]$, ($i = 1, 2, 3$),

$$f_1(P) = P(A \cap B), \quad f_2(P) = P(A), \quad f_3(P) = P(B),$$

and continuous $g : [0, 1]^3 \rightarrow [1, -1]$, $g(x_1, x_2, x_3) = x_1 - x_2x_3$. Now,

$$h : \mathcal{P} \rightarrow [0, 1] : P \mapsto |g \circ (f_1, f_2, f_3)(P)|,$$

is \mathcal{U}_1 -continuous. Then $B = h^{-1}(\{0\})$ is closed (for \mathcal{I}_∞) and (since the complement V' is open in $[0, 1]$, it is F_σ , so) $V = h^{-1}(V')$ is F_σ (for \mathcal{I}_∞). So independence of events A and B is asymptotically testable.

Pointwise testability: examples (II)

Example 29.1 *Is independence of real-valued X and Y testable?*

Let $A_k \in \sigma_X, B_l \in \sigma_Y$ be generators. Consider,

$H_0 : \forall_{k,l} P(A_k \cap B_l) = P(A_k)P(B_l), H_1 : \exists_{k,l} P(A_k \cap B_l) \neq P(A_k)P(B_l).$

Define \mathcal{U}_1 -continuous $f_{kl,i} : \mathcal{P} \rightarrow [0, 1], (i = 1, 2, 3),$

$$f_{kl,1}(P) = P(A_k \cap B_l), \quad f_{k,2}(P) = P(A_k), \quad f_{l,3}(P) = P(B_l),$$

and continuous $g : [0, 1]^3 \rightarrow [1, -1], g(x_1, x_2, x_3) = x_1 - x_2x_3.$ Now,

$$h : \mathcal{P} \rightarrow [0, 1]^{\mathbb{N}} : P \mapsto (|g \circ (f_{kl,1}, f_{k,2}, f_{l,3})(P)| : k, l \geq 1),$$

is \mathcal{U}_1 -continuous. Then $B = h^{-1}(\{0\})$ is closed (for \mathcal{I}_∞) and (since the complement V' is open in $[0, 1]^{\mathbb{N}}$, it is F_σ , so) $V = h^{-1}(V')$ is F_σ (for \mathcal{I}_∞). So independence of X and Y is asymptotically testable.

Bayesian testability: equivalent formulations

Theorem 30.1 *Let a model $(\mathcal{P}, \mathcal{G}, \Pi)$ with $B, V \in \mathcal{G}$ be given, with $\Pi(B) > 0, \Pi(V) > 0$. The following are equivalent,*

i. *there exist **Bayesian tests** for B vs V ,*

ii. *there are tests ϕ_n such that for **Π -almost-all** $P \in B, Q \in V$,*

$$P^n \phi_n \rightarrow 0, \quad Q^n (1 - \phi_n) \rightarrow 0,$$

iii. *there are tests $\phi_n : \mathcal{X}^n \rightarrow [0, 1]$ such that,*

$$\int_B P^n \phi_n d\Pi(P) + \int_V Q^n (1 - \phi_n) d\Pi(Q) \rightarrow 0,$$

iv. *for **Π -almost-all** $P \in B, Q \in V$,*

$$\Pi(V|X^n) \xrightarrow{P} 0, \quad \Pi(B|X^n) \xrightarrow{Q} 0.$$

Characterisation of Bayesian testability

Definition 31.1 Given model $(\mathcal{P}, \mathcal{G}, \Pi)$. An event $B \in \mathcal{B}^\infty$ is called a Π -zero-one set, if $P^\infty(B) \in \{0, 1\}$, for Π -almost-all $P \in \mathcal{P}$. A model subset $G \in \mathcal{G}$ is called a Π -one set if there is a Π -zero-one set B such that $G = \{P \in \mathcal{P} : P^\infty(B) = 1\}$.

Proposition 31.2 (Martingale convergence) Let $(\mathcal{P}, \mathcal{G}, \Pi)$ be given. Let V be a Π -one set. Then, for Π -almost-all $P \in \mathcal{P}$,

$$\Pi(V|X^n) \xrightarrow{P\text{-a.s.}} 1_V(P). \quad (5)$$

Theorem 31.3 Let $(\mathcal{P}, \mathcal{G})$ be a measurable model with a prior Π that is a Radon measure and hypotheses B, V . There is a Bayesian test sequence for B vs V , if and only if, B, V are \mathcal{G} -measurable.

Part III

Constructive results

Bayesian testing power

Denote the density for the local prior predictive distribution $P_n^{\Pi|B}$ with respect to $\mu_n = P_n^{\Pi|B} + P_n^{\Pi|V}$ by $p_{B,n}$, and similar for $P_n^{\Pi|V}$.

Proposition 33.1 *Let $(\mathcal{P}, \mathcal{G}, \Pi)$ be a model with measurable B, V . There are tests ϕ_n such that,*

$$\begin{aligned} \int_B P^n \phi_n d\Pi(P) + \int_V Q^n (1 - \phi_n) d\Pi(Q) \\ \leq \int \left(\Pi(B) p_{B,n}(x) \right)^\alpha \left(\Pi(V) p_{V,n}(x) \right)^{1-\alpha} d\mu_n(x), \end{aligned} \quad (6)$$

for every $n \geq 1$ and any $0 \leq \alpha \leq 1$.

Proposition 33.2 *For every $n \geq 1$, the test,*

$$\phi_n(X^n) = 1\{X^n : \Pi(V|X^n) \geq \Pi(B|X^n)\},$$

based on posterior odds has optimal Bayesian testing power.

Posterior odds model selection for frequentists

Theorem 34.1 For all $n \geq 1$, let the model be a probability space $(\mathcal{P}, \mathcal{G}, \Pi_n)$. Consider *disjoint, measurable* $B, V \subset \Theta$ with $\Pi_n(B), \Pi_n(V) > 0$ such that,

i. There are *Bayesian tests* for B vs V of *power* $a_n \downarrow 0$,

$$\int_B P^n \phi_n d\Pi_n(P) + \int_V Q^n (1 - \phi_n) d\Pi_n(Q) = o(a_n),$$

ii. for all $P \in B$, $P^n \triangleleft a_n^{-1} P_n^{\Pi_n|B}$; for all $Q \in V$, $Q^n \triangleleft a_n^{-1} P_n^{\Pi_n|V}$.

Then the indicators $\phi_n(X^n) = 1\{X^n : \Pi(V|X^n) \geq \Pi(B|X^n)\}$ for posterior odds form a *pointwise* test sequence for B vs V .

See arXiv:1611.08444 [math.ST]

Remote contiguity

Definition 35.1 Given $(P_n), (Q_n)$ and a $a_n \downarrow 0$, Q_n is a_n -remotely contiguous w.r.t. P_n ($Q_n \triangleleft a_n^{-1} P_n$), if for any msb $\psi_n : \mathcal{X}^n \rightarrow [0, 1]$

$$P_n \psi_n = o(a_n) \quad \Rightarrow \quad Q_n \psi_n = o(1)$$

Lemma 35.2 $Q_n \triangleleft a_n^{-1} P_n$ if any of the following holds:

- (i) For any bnd msb $T_n : \mathcal{X}^n \rightarrow \mathbb{R}$, $a_n^{-1} T_n \xrightarrow{P_n} 0$, implies $T_n \xrightarrow{Q_n} 0$
- (ii) Given $\epsilon > 0$, there is a $\delta > 0$ s.t. $Q_n(dP_n/dQ_n < \delta a_n) < \epsilon$ f.l.e.n.
- (iii) There is a $b > 0$ s.t. $\liminf_{n \rightarrow \infty} b a_n^{-1} P_n(dQ_n/dP_n > b a_n^{-1}) = 1$
- (iv) Given $\epsilon > 0$, there is a $c > 0$ such that $\|Q_n - Q_n \wedge c a_n^{-1} P_n\| < \epsilon$
- (v) Under Q_n , every subsequence of $(a_n(dP_n/dQ_n)^{-1})$ has a further subsequence that converges in \mathcal{T}_C .

Example: KL-neighbourhoods

Example 36.1 Let \mathcal{P} be a model for i.i.d. data X^n . Let P_0, P and $\epsilon > 0$ be such that $-P_0 \log(dP/dP_0) < \epsilon^2$. Then, for large enough n ,

$$\frac{dP^n}{dP_0^n}(X^n) \geq e^{-\frac{n}{2}\epsilon^2}, \quad (7)$$

with P_0^n -probability one. So for any tests ψ_n ,

$$P^n \psi_n \geq e^{-\frac{1}{2}n\epsilon^2} P_0^n \psi_n. \quad (8)$$

So if $P^n \phi_n = o(\exp(-\frac{1}{2}n\epsilon^2))$ then $P_0^n \phi_n = o(1)$: $P_0^n \triangleleft a_n^{-1} P^n$ with $a_n = \exp(-\frac{1}{2}n\epsilon^2)$.

Consistent model selection

Let \mathcal{P} be a model for *i.i.d.* data $X^n \sim P^n$, ($n \geq 1$), and suppose that $(\mathcal{P}, \mathcal{G}, \Pi)$ has finite, measurable partition,

$$P \in \mathcal{P} = \mathcal{P}_1 \cup \dots \cup \mathcal{P}_M.$$

Model-selection Which $1 \leq i \leq M$? (such that $P \in \mathcal{P}_i$)

Theorem 37.1 Assume that for all $1 \leq i < j \leq M$,

\mathcal{P}_i and \mathcal{P}_j are \mathcal{U}_∞ -uniformly separated.

Let $1 \leq i \leq M$ be such that $P \in \mathcal{P}_i$. If Π is a KL-prior, then indicators for posterior odds,

$$\phi_n(X^n) = 1 \left\{ X^n : \Pi(\mathcal{P}_i | X^n) \geq \sum_{j \neq i} \Pi(\mathcal{P}_j | X^n) \right\},$$

are a pointwise test for \mathcal{P}_i vs $\cup_{j \neq i} \mathcal{P}_j$.

Example: select the DAG (I)

Observe an *i.i.d.* X^n of vectors of discrete random variables $X_i = (X_{1,i}, \dots, X_{k,i}) \in \mathbb{Z}^k$. We assume that $X \sim P$ follows a graphical model,

$$P_{\mathcal{A}, \theta}(X_1 \in B_1, \dots, X_k \in B_k) = \prod_{i=1}^k P_{\theta_i}(X_i \in B_i | \mathcal{A}_i)$$

where $\mathcal{A}_i \subset \{1, \dots, k\}$ denotes the *parents* of X_k . Together, the \mathcal{A}_i describe a directed, a-cyclical graph.

Family \mathcal{F} of kernels $p_{\theta}(\cdot | \cdot) : \mathbb{Z} \times \mathbb{Z}^l \rightarrow [0, 1]$, for $\theta \in \Theta$, $1 \leq l \leq k$. Assume that Θ is compact and,

$$\theta \mapsto \sum_{x \in \mathbb{Z}} f(x) P_{\theta}(x | z_1, \dots, z_l)$$

is continuous, for every bounded $f : \mathbb{Z} \rightarrow \mathbb{R}$ and all $z_1, \dots, z_l \in \mathbb{Z}$.

Example: select the DAG (II)

The DAG $\mathcal{A} = (\mathcal{A}_i : 1 \leq i \leq k)$ represents a number of conditional independence statements concerning the components X_1, \dots, X_k : for all $1 \leq i < j \leq k$, given $X_l = z$ for all $l \in \mathcal{A}_i \cup \mathcal{A}_j$, X_i is independent of X_j .

Define the submodels $\mathcal{P}_{\mathcal{A}} = \{P_{\mathcal{A},\theta} : \theta \in \Theta^k\}$, for all \mathcal{A} . Given a conditional independence relation for \mathcal{A} , we require that, for all θ , all $z \in \mathbb{Z}$, all $A, B \subset \mathbb{Z}$, any $\mathcal{A}' \neq \mathcal{A}$,

$$\left| P_{\mathcal{A}',\theta}(X_i \in A, X_j \in B | X_l = z) - P_{\mathcal{A}',\theta}(X_i \in A | X_l = z) P_{\mathcal{A}',\theta}(X_j \in B | X_l = z) \right| > \epsilon,$$

for some $\epsilon > 0$ that depends only on \mathcal{A} and \mathcal{A}' .

With a KL-prior posterior odds for $\mathcal{P}_{\mathcal{A}}$ select the correct DAG \mathcal{A} .

Example: how many clusters? (I)

Observe *i.i.d.* $X^n \sim P^n$, where P dominated with density p .

Clusters Family \mathcal{F} of kernels $\varphi_\theta : \mathbb{R} \rightarrow [0, \infty)$, with parameter $\theta \in \Theta$. Assume Θ compact and,

$$\theta \mapsto \int f(x)\varphi_\theta(x) dx,$$

is continuous, for every bounded, measurable $f : \mathbb{R} \rightarrow \mathbb{R}$. Define $\Theta'_M = \Theta^M / \sim$.

Model Assume that there is an $M > 0$ such that p can be written as,

$$p_{\lambda,\theta}(x) = \sum_{m=1}^M \lambda_m p_{\theta_m}(x),$$

for some $M \geq 1$, with $\lambda \in S_M = \{\lambda \in [0, 1]^M : \sum_m \lambda_m = 1\}$, $\theta \in \Theta'_M$.

Example: how many clusters? (II)

Assume M less than some known M' . Choose prior $\Pi_{\lambda,M}$ for $\lambda \in S_M$ such that, for some $\epsilon > 0$,

$$\Pi_{\lambda,M}(\lambda \in S_M : \epsilon < \min\{\lambda_m\}, \max\{\lambda_m\} < 1 - \epsilon) = 1.$$

For $\theta \in \Theta'_M$ also choose a prior $\Pi_{\theta,M}$ that 'stays away from the edges'. Define,

$$\Pi = \sum_{M=1}^{M'} \mu_M \Pi_{\lambda,M} \times \Pi_{\theta,M}.$$

(for $\sum_M \mu_M = 1$).

If Π is a KL-prior, posterior odds select the correct number of clusters M . If there are no M' and ϵ known, there are sequences $M'_n \rightarrow \infty$ and $\epsilon_n \downarrow 0$ with priors Π_n that finds the correct number of clusters.