

Kernels Based Tests with Non-asymptotic Bootstrap Approaches for Two-sample Problems

Magalie Fromont, Béatrice Laurent, Matthieu Lerasle, Patricia Reynaud-Bouret (COLT-2012)

Zoltán Szabó

Machine Learning Journal Club, Gatsby Unit

November 2, 2015

Task: two-sample problem

- Given: 2 independent samples

$$Z^{(1)} = \{Z_1^{(1)}, \dots, Z_{(N_1)}^{(1)}\} \sim s_1,$$

$$Z^{(2)} = \{Z_1^{(2)}, \dots, Z_{(N_2)}^{(2)}\} \sim s_2.$$

- We want to test
 - $H_0 : s_1 = s_2$, against
 - $H_1 : s_1 \neq s_2$.

Task: two-sample problem

- Given: 2 independent samples

$$Z^{(1)} = \{Z_1^{(1)}, \dots, Z_{(N_1)}^{(1)}\} \sim s_1,$$

$$Z^{(2)} = \{Z_1^{(2)}, \dots, Z_{(N_2)}^{(2)}\} \sim s_2.$$

- We want to test
 - $H_0 : s_1 = s_2$, against
 - $H_1 : s_1 \neq s_2$.
- [\mathcal{Z} : measurable space, e.g. 'nice' $\mathcal{Z} \subseteq \mathbb{R}^d$.]

- Error of
 - **first** kind: H_0 is true, but we reject it.
 - **second** kind: H_0 is false, but we accept it.

- Error of
 - **first** kind: H_0 is true, but we reject it.
 - **second** kind: H_0 is false, but we accept it.
- Test is called of **level** $\alpha \in (0, 1)$ if

$$\mathbb{P}(\text{first kind of error}) \leq \alpha.$$

- Difficulty: hard to find the *non-asymptotic* quantile for α -levelness! \rightarrow small-sample domain.

Two-sample problem: density model – classical setup

- Examples = different $Z^{(i)}$ generating mechanisms.
- Here: $N_1 = n_1, N_2 = n_2$: fixed,
 - $Z^{(1)} = \{Z_1^{(1)}, \dots, Z_{(n_1)}^{(1)}\} \sim s_1$
 - $Z^{(2)} = \{Z_1^{(2)}, \dots, Z_{(n_2)}^{(2)}\} \sim s_2$

Two-sample problem: density model – classical setup

- Examples = different $Z^{(i)}$ generating mechanisms.
- Here: $N_1 = n_1, N_2 = n_2$: fixed,
 - $Z^{(1)} = \{Z_1^{(1)}, \dots, Z_{(n_1)}^{(1)}\} \sim s_1[\ll \nu]$,
 - $Z^{(2)} = \{Z_1^{(2)}, \dots, Z_{(n_2)}^{(2)}\} \sim s_2[\ll \nu]$,
 - $[\nu$:
 - non-atomic: $z \in \mathcal{Z}$ s.t. $\nu(\{z\}) > 0$ can not happen.
 - σ -finite measure: $\mathcal{Z} = \dot{\cup}_{i \in I} P_i, I$: countable, $\nu(P_i) < \infty$.]
 - $[s_1, s_2 \in L^2(\mathcal{Z}, \nu)$.]

Two-sample problem: heteroscedastic regression model

- $N_1 = n_1, N_2 = n_2$: fixed,
- $Z^{(1)}, Z^{(2)}$ i.i.d.

$$\begin{aligned} Z_i^{(1)} &= (X_i^{(1)}, Y_i^{(1)}) & Y_i^{(1)} &= s_1(X_i^{(1)}) + \sigma(X_i^{(1)})\xi_i^{(1)}, \\ Z_i^{(2)} &= (X_i^{(2)}, Y_i^{(2)}) & Y_i^{(2)} &= s_2(X_i^{(2)}) + \sigma(X_i^{(2)})\xi_i^{(2)}, \end{aligned}$$

$$(X_i^{(1)}, \xi_i^{(1)}) \stackrel{\text{distr}}{=} (X_i^{(2)}, \xi_i^{(2)}), \mathbb{E}[\xi_i^{(1)} | X_i^{(1)}] = 0, \mathbb{E}\left[\left(\xi_i^{(1)}\right)^2 | X_i^{(1)}\right] = 1,$$

Two-sample problem: heteroscedastic regression model

- $N_1 = n_1, N_2 = n_2$: fixed,
- $Z^{(1)}, Z^{(2)}$ i.i.d.

$$\begin{aligned} Z_i^{(1)} &= (X_i^{(1)}, Y_i^{(1)}) & Y_i^{(1)} &= s_1(X_i^{(1)}) + \sigma(X_i^{(1)})\xi_i^{(1)}, \\ Z_i^{(2)} &= (X_i^{(2)}, Y_i^{(2)}) & Y_i^{(2)} &= s_2(X_i^{(2)}) + \sigma(X_i^{(2)})\xi_i^{(2)}, \end{aligned}$$

$$(X_i^{(1)}, \xi_i^{(1)}) \stackrel{\text{distr}}{=} (X_i^{(2)}, \xi_i^{(2)}), \mathbb{E}[\xi_i^{(1)} | X_i^{(1)}] = 0, \mathbb{E}\left[\left(\xi_i^{(1)}\right)^2 | X_i^{(1)}\right] = 1,$$

- $[s_1, s_2, \sigma \in L^2(X, P_X), \mathcal{Z} = \mathcal{X} \times \mathcal{Y}, \mathcal{Y} \subseteq \mathbb{R}.]$

Two-sample problem: Poisson process model

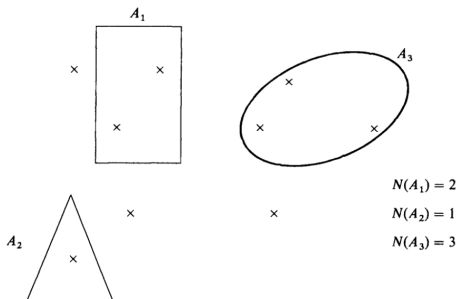
- $X \sim \text{Poisson}[\lambda] : \mathbb{P}(X = n) = \frac{\lambda^n}{n!} e^{-\lambda}, n = 0, 1, \dots; \lambda \geq 0.$

Two-sample problem: Poisson process model

- $X \sim \text{Poisson}[\lambda] : \mathbb{P}(X = n) = \frac{\lambda^n}{n!} e^{-\lambda}, n = 0, 1, \dots; \lambda \geq 0.$
- # of raisins in a cake / stars on the sky:



Poisson process model



An N process producing random points in $(\mathcal{Z}, \mathcal{A}, m)$ is called PP if

- 1 $N(A) \sim \text{Poisson}[m(A)]$, for all $A \in \mathcal{A}$
- 2 $\{A_i\}_{i \in I}$ disjoint, I : countable $\Rightarrow \{N(A_i)\}_{i \in I}$ are independent.

- If \mathcal{Z} is finite ($m(\mathcal{Z}) < \infty$):
 - Z_1, Z_2, \dots i.i.d., $\mathbb{P}(Z_i \in A) = \frac{m(A)}{m(\mathcal{Z})}$, $\forall A \in \mathcal{A}$,
 - $p \sim \text{Poisson}[m(\mathcal{Z})]$, independent of X_i -s.
 - $N(A) = \sum_{i=1}^p \chi\{Z_i \in A\}$ is good!

Poisson process: it exists, construction

- If \mathcal{Z} is finite ($m(\mathcal{Z}) < \infty$):
 - Z_1, Z_2, \dots i.i.d., $\mathbb{P}(Z_i \in A) = \frac{m(A)}{m(\mathcal{Z})}$, $\forall A \in \mathcal{A}$,
 - $p \sim \text{Poisson}[m(\mathcal{Z})]$, independent of X_i -s.
 - $N(A) = \sum_{i=1}^p \chi\{Z_i \in A\}$ is good!
- If \mathcal{Z} is σ -finite:

$$\mathcal{Z} = \dot{\cup}_{i \in I} P_i, \quad P_i \leftrightarrow N_i, \quad N = \sum_{i \in I} N_i.$$

Two-sample problem: Poisson process model

- $Z^{(1)} = \{Z_1^{(1)}, \dots, Z_{(N_1)}^{(1)}\}$, $m := s_1$ intensity
- $Z^{(2)} = \{Z_1^{(2)}, \dots, Z_{(N_2)}^{(2)}\}$, $m := s_2$
- N_1, N_2 : Poisson r.v.-s.

Two-sample problem: Poisson process model

- $Z^{(1)} = \{Z_1^{(1)}, \dots, Z_{(N_1)}^{(1)}\}$, $m := s_1$ intensity $[\ll \mu]$,
- $Z^{(2)} = \{Z_1^{(2)}, \dots, Z_{(N_2)}^{(2)}\}$, $m := s_2$ $[\ll \mu]$,
- N_1, N_2 : Poisson r.v.-s.
- $[\mu$: non-atomic, σ -finite.]
- $[\mu = n\nu$; ν : non-atomic, σ -finite; ???]

After the examples, back to two-sampling (density model)

- Pooled samples: $Z = Z^{(1)} \cup Z^{(2)}$, $n = n_1 + n_2$. Test statistics:

$$T_K = \sum_{i \neq j \in \{1, \dots, n\}} K(Z_i, Z_j) (\epsilon_i^0 \epsilon_j^0 + c_{n_1, n_2}),$$
$$c_{n_1, n_2} = \frac{1}{n_1 n_2 (n_1 - 1 + n_2 - 1)},$$
$$a_{n_1, n_2} = \sqrt{\frac{1}{n_1 (n_1 - 1)} - c_{n_1, n_2}}, \quad b_{n_1, n_2} = -a_{n_2, n_1},$$
$$\epsilon_i^0 = \begin{cases} a_{n_1, n_2} & Z_i \in Z^{(1)} \\ b_{n_1, n_2} & Z_i \in Z^{(2)}. \end{cases}$$

- Example: $K =$ reproducing kernel.

T_K makes sense

Let $(K \diamond p)(z) = \langle K(\cdot, z), p \rangle$, $\langle f, g \rangle = \int_{\mathcal{Z}} f(z)g(z)d\nu(z)$. **Unbiasedness:**

$$\mathbb{E}_{s_1, s_2}[T_K] \stackrel{?}{=} \langle K \diamond (s_1 - s_2), s_1 - s_2 \rangle$$

Let $(K \diamond p)(z) = \langle K(\cdot, z), p \rangle$, $\langle f, g \rangle = \int_{\mathcal{Z}} f(z)g(z)d\nu(z)$. **Unbiasedness:**

$$\begin{aligned}\mathbb{E}_{s_1, s_2}[T_K] &\stackrel{?}{=} \langle K \diamond (s_1 - s_2), s_1 - s_2 \rangle \\ &= \int_{\mathcal{Z}^2} K(z, z')(s_1 - s_2)(z)(s_1 - s_2)(z')d\nu(z)d\nu(z')\end{aligned}$$

Let $(K \diamond p)(z) = \langle K(\cdot, z), p \rangle$, $\langle f, g \rangle = \int_{\mathcal{Z}} f(z)g(z)d\nu(z)$. **Unbiasedness:**

$$\begin{aligned}\mathbb{E}_{s_1, s_2}[T_K] &\stackrel{?}{=} \langle K \diamond (s_1 - s_2), s_1 - s_2 \rangle \\ &= \int_{\mathcal{Z}^2} K(z, z')(s_1 - s_2)(z)(s_1 - s_2)(z')d\nu(z)d\nu(z') \\ &= \sum_{\ell=1}^2 \int_{\mathcal{Z}^2} K(z, z')s_\ell(z)s_\ell(z')d\nu(z)d\nu(z') - 2 \int_{\mathcal{Z}^2} K(z, z')s_1(z)s_2(z')d\nu(z)d\nu(z')\end{aligned}$$

Let $(K \diamond p)(z) = \langle K(\cdot, z), p \rangle$, $\langle f, g \rangle = \int_{\mathcal{Z}} f(z)g(z)d\nu(z)$. **Unbiasedness:**

$$\begin{aligned} \mathbb{E}_{s_1, s_2}[T_K] &\stackrel{?}{=} \langle K \diamond (s_1 - s_2), s_1 - s_2 \rangle \\ &= \int_{\mathcal{Z}^2} K(z, z')(s_1 - s_2)(z)(s_1 - s_2)(z')d\nu(z)d\nu(z') \\ &= \sum_{\ell=1}^2 \int_{\mathcal{Z}^2} K(z, z')s_\ell(z)s_\ell(z')d\nu(z)d\nu(z') - 2 \int_{\mathcal{Z}^2} K(z, z')s_1(z)s_2(z')d\nu(z)d\nu(z') \\ &= \mathbb{E}_{s_1, s_2} \left[\underbrace{\sum_{\ell=1}^2 \sum_{i \neq j \in \{1, \dots, n_\ell\}} K(z_i^{(\ell)}, z_j^{(\ell)}) c_{n_\ell} - 2 \sum_{i=1}^{n_1} \sum_{j=1}^{n_2} K(z_i^{(1)}, z_j^{(2)}) \frac{1}{n_1 n_2}}_{=T_K} \right], \end{aligned}$$

where $c_{n_\ell} = \frac{1}{n_\ell(n_\ell-1)}$ ($\ell = 1, 2$).

$\langle K \diamond (s_1 - s_2), s_1 - s_2 \rangle$ is finite

By CBS ($\|\cdot\| = \|\cdot\|_{L^2(\mathcal{Z}, \nu)}$)

$$|\langle K \diamond (s_1 - s_2), s_1 - s_2 \rangle| \leq \underbrace{\|K \diamond (s_1 - s_2)\|}_{=:(*)_1} \underbrace{\|s_1 - s_2\|}_{=:(*)_2},$$

where

- $(*)_1$: assumed to be finite,
- $(*)_2$: $< \infty$ since $s_1, s_2 \in L^2(\mathcal{Z}, \nu)$.

If K is a reproducing kernel:

$$\begin{aligned}\mathbb{E}_{s_1, s_2}[T_K] &= \int_{\mathcal{Z}^2} K(z, z')(s_1 - s_2)(z)(s_1 - s_2)(z')d\nu(z)d\nu(z') \\ &= \left\| \int_{\mathcal{Z}} K(\cdot, z)(s_1 - s_2)(z)d\nu(z) \right\|_{H(K)}^2 \\ &= \|\mu_{s_1} - \mu_{s_2}\|_{H(K)}^2, \quad \mu_s = \mu_{s, K}.\end{aligned}$$

$\Rightarrow \mathbb{E}_{s_1, s_2}[T_K] = 0 \Leftrightarrow s_1 = s_2$, if K is characteristic.

K -examples: K is a projection kernel

Let $\{\phi_\lambda\}_{\lambda \in \Lambda}$ be an ONS.

$$K(z, z') = \sum_{\lambda \in \Lambda} \phi_\lambda(z) \phi_\lambda(z'),$$

$$\int_{\mathcal{Z}} K(z, z') f(z') d\nu(z') = \text{proj}_S(f)(z),$$

$$S = \text{span}(\{\phi_\lambda\}_{\lambda \in \Lambda}),$$

$$\langle K \diamond (s_1 - s_2), s_1 - s_2 \rangle = \|\text{proj}_S(s_1 - s_2)\|_{L^2(\mathcal{Z}, \nu)}^2.$$

Example: $\{\phi_\lambda\}_{\lambda \in \Lambda}$ Fourier/Haar basis.

K -examples: K is a convolution kernel

$\mathcal{Z} = \mathbb{R}^d$, $\nu = \text{Lebesgue measure}$, $k \in L^2(\mathbb{R}^d)$, $k(-x) = k(x)$, $h_i > 0$ ($\forall i$).

$$K(z, z') = \frac{1}{\prod_{i=1}^d h_i} k\left(\frac{z_1 - z'_1}{h_1}, \dots, \frac{z_d - z'_d}{h_d}\right),$$

$$\langle K \diamond (s_1 - s_2), s_1 - s_2 \rangle = \langle k_h * (s_1 - s_2), s_1 - s_2 \rangle_{L^2(\mathbb{R}^d)},$$

$$(f * g)(x) = \int_{\mathbb{R}^d} f(y)g(x - y)dy,$$

$$k_h(u_1, \dots, u_d) = \frac{1}{\prod_{i=1}^d h_i} k\left(\frac{u_1}{h_1}, \dots, \frac{u_d}{h_d}\right).$$

- Recall: $T_K = \sum_{i \neq j \in \{1, \dots, n\}} K(Z_i, Z_j) (\epsilon_i^0 \epsilon_j^0 + c_{n_1, n_2})$,

$$\epsilon_i^0 = \begin{cases} a_{n_1, n_2} & i \in \{1, \dots, n_1\} \\ b_{n_1, n_2} & i \in \{n_1 + 1, \dots, n_1 + n_2\}. \end{cases}$$

- $R = (R_1, \dots, R_n)$: rnd permutation, independent of Z .

$$\epsilon_i = \begin{cases} a_{n_1, n_2} & i \in \{R_1, \dots, R_{n_1}\} \\ b_{n_1, n_2} & i \in \{R_{n_1+1}, \dots, R_{n_1+n_2}\}. \end{cases},$$

$$T_K^\epsilon = \sum_{i \neq j \in \{1, \dots, n\}} K(Z_i, Z_j) (\epsilon_i \epsilon_j + c_{n_1, n_2}).$$

- Recall: $T_K = \sum_{i \neq j \in \{1, \dots, n\}} K(Z_i, Z_j) (\epsilon_i^0 \epsilon_j^0 + c_{n_1, n_2})$,

$$\epsilon_i^0 = \begin{cases} a_{n_1, n_2} & i \in \{1, \dots, n_1\} \\ b_{n_1, n_2} & i \in \{n_1 + 1, \dots, n_1 + n_2\}. \end{cases}$$

- $R = (R_1, \dots, R_n)$: rnd permutation, independent of Z .

$$\epsilon_i = \begin{cases} a_{n_1, n_2} & i \in \{R_1, \dots, R_{n_1}\} \\ b_{n_1, n_2} & i \in \{R_{n_1+1}, \dots, R_{n_1+n_2}\}. \end{cases},$$

$$T_K^\epsilon = \sum_{i \neq j \in \{1, \dots, n\}} K(Z_i, Z_j) (\epsilon_i \epsilon_j + c_{n_1, n_2}).$$

- $T_K|Z \stackrel{\text{distr}}{=} T_K^\epsilon|Z$.

- $q_{K,1-\alpha}^{(Z)}$: $(1 - \alpha)$ -quantile of $T_K^\epsilon|Z \stackrel{distr}{=} T_K|Z$.
- Reject H_0 if $T_K > q_{K,1-\alpha}^{(Z)}$.

- $q_{K,1-\alpha}^{(Z)}$: $(1 - \alpha)$ -quantile of $T_K^\epsilon | Z \stackrel{distr}{=} T_K | Z$.
- Reject H_0 if $T_K > q_{K,1-\alpha}^{(Z)}$.
- Given $\alpha \in (0, 1)$, under H_0

$$\mathbb{P}_{s_1, s_2} \left(T_K > q_{K,1-\alpha}^{(Z)} \mid Z \right) \leq \alpha.$$

- $q_{K,1-\alpha}^{(Z)}$: $(1 - \alpha)$ -quantile of $T_K^\epsilon | Z \stackrel{distr}{=} T_K | Z$.
- Reject H_0 if $T_K > q_{K,1-\alpha}^{(Z)}$.
- Given $\alpha \in (0, 1)$, under H_0

$$\mathbb{P}_{s_1, s_2} \left(T_K > q_{K,1-\alpha}^{(Z)} \mid Z \right) \leq \alpha.$$

- Taking expectation over Z , $\sup [\mathbb{P}_{(H_0)}(A) = \sup_{(s_1, s_2), s_1=s_2} \mathbb{P}_{s_1, s_2}(A)]$:

$$\mathbb{P}_{(H_0)}(\Phi_{K,\alpha} = 1) \leq \alpha,$$

$$\Phi_{K,\alpha} = \chi \left\{ T_K > q_{K,1-\alpha}^{(Z)} \right\}.$$

- Given:
 - $\{K_m\}_{m=1}^M$ kernels.
 - $\{w_m\}_{m=1}^M$ positive weights s.t. $\sum_m e^{-w_m} \leq 1$.

- Given:
 - $\{K_m\}_{m=1}^M$ kernels.
 - $\{w_m\}_{m=1}^M$ positive weights s.t. $\sum_m e^{-w_m} \leq 1$.
- Notation: T_{K_m} , $T_{K_m}^\epsilon$; $q_{m,1-u}^{(Z)} = (1-u)$ -quantile of $T_{K_m}^\epsilon | Z$.

Aggregated test

- Given:
 - $\{K_m\}_{m=1}^M$ kernels.
 - $\{w_m\}_{m=1}^M$ positive weights s.t. $\sum_m e^{-w_m} \leq 1$.
- Notation: $T_{K_m}, T_{K_m}^\epsilon; q_{m,1-u}^{(Z)} = (1-u)$ -quantile of $T_{K_m}^\epsilon | Z$.
- For $\alpha \in (0, 1)$

$$u_\alpha^{(Z)} := \sup \left\{ u > 0 : \mathbb{P} \left(\underbrace{\sup_m (T_{K_m}^\epsilon - q_{m,1-ue^{-w_m}}^{(Z)})}_{\geq 1 \text{ test rejects}} > 0 \mid Z \right) \leq \alpha \right\}.$$

Aggregated test

- Given:
 - $\{K_m\}_{m=1}^M$ kernels.
 - $\{w_m\}_{m=1}^M$ positive weights s.t. $\sum_m e^{-w_m} \leq 1$.
- Notation: $T_{K_m}, T_{K_m}^\epsilon$; $q_{m,1-u}^{(Z)} = (1-u)$ -quantile of $T_{K_m}^\epsilon | Z$.
- For $\alpha \in (0, 1)$

$$u_\alpha^{(Z)} := \sup \left\{ u > 0 : \underbrace{\mathbb{P} \left(\sup_m (T_{K_m}^\epsilon - q_{m,1-ue^{-w_m}}^{(Z)}) > 0 \mid Z \right)}_{\geq 1 \text{ test rejects}} \leq \alpha \right\}.$$

- The test: $\exists m : T_{K_m}^\epsilon - q_{m,1-u_\alpha^{(Z)}}^{(Z)} e^{-w_m} > 0$; it is of level α .

- Focus: two-sample problem.
- Unbiasedness and exact quantile for the test statistics.
- Single kernel and aggregated tests of level α .

