

Testing for Change-Point

$$H_0: \theta_0 = \theta_k$$

$$H_k: \theta_0 \neq \theta_k \quad \begin{cases} \theta_0 = \theta, & r_1 = \dots = r \\ \theta_0 \neq \theta, & t_1, \dots, t_r \text{ for some } h. \end{cases}$$

$$L_n(\theta_1, \theta, r) = \sum_{i=1}^n (y_i - \theta_1 y_{r_1})^2 + \sum_{i=r_1+1}^r (y_i - \theta_2 y_{r_2})^2$$

$$\min_{\theta_1, \theta_2} L_n(\theta_1, \theta_2, r)$$

$$= L_n(\hat{\theta}_1(r), \hat{\theta}_2(r), r) \quad \text{Under } H_0.$$

Under H_0 ,

$$\min_{\theta} L_n(\theta) = \sum_{i=1}^n (y_i - \bar{y}_n)^2$$

$$\min_{\theta} L_n(\theta) = L_n(\bar{y}_n)$$

$$\hat{\theta} = \sum_{i=1}^n y_i y_{r_1} / \sum_{i=1}^n y_{r_i}^2$$

The Likelihood ratio test for a given r ,

$$L_n(\hat{\theta}) - L_n(\hat{\theta}_1(r), \hat{\theta}_2(r), r)$$

$$= \sum_{i=1}^n (y_i - \bar{y}_n)^2 - \sum_{i=1}^r (y_i - \hat{\theta}_1(r) y_{r_1})^2 - \sum_{i=r_1+1}^r (y_i - \hat{\theta}_2(r) y_{r_2})^2$$

$$= \sum_{i=1}^n [(y_i - \bar{y}_n)^2 - \sum_{i=1}^r (y_i - \hat{\theta}_1(r) y_{r_1})^2 - \sum_{i=r_1+1}^r (y_i - \hat{\theta}_2(r) y_{r_2})^2]$$

$$= - \left(\sum_{i=1}^n y_i \varepsilon_i \right)^2 / \sum_{i=1}^n y_i^2$$

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$$+ \left(\sum_{i=1}^n y_i \varepsilon_i \right)^2 / \sum_{i=1}^n y_i^2 + \left(\sum_{i=1}^n y_i \varepsilon_i \right)^2 / \sum_{i=1}^n y_i^2$$

$$\frac{1}{n} \sum_{i=1}^n y_i^2 = E y_i^2 + o_p(1), \quad \frac{1}{n(n-1)} \sum_{i \neq j=1}^n y_i^2 = E y_i^2 + o_p(1)$$

$$S_y^2 = E y_i^2$$

$$= + \frac{1}{S_y^2} \left[\frac{1}{n} \left(\sum_{i=1}^n y_i \varepsilon_i \right)^2 + \frac{1}{n-1} \left(\sum_{i=1}^n y_i^2 \varepsilon_i \right)^2 - \frac{1}{n} \left(\sum_{i=1}^n y_i^2 \varepsilon_i \right)^2 \right] + o_p(1)$$

$$= \frac{1}{S_y^2} \frac{\left[n \left(\sum_{i=1}^n y_i \varepsilon_i \right)^2 - n \left(\sum_{i=1}^n y_i \varepsilon_i \right)^2 \right]^2}{n(n-1)}$$

$$= \frac{1}{S_y^2} \frac{\left[\frac{1}{n} \left(\sum_{i=1}^n y_i \varepsilon_i \right)^2 - \frac{1}{n} \left(\sum_{i=1}^n y_i \varepsilon_i \right)^2 \right]^2}{\frac{1}{n} \left(1 - \frac{1}{n} \right)}$$

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$$\frac{1}{\sqrt{n}} \sum_{i=1}^{[n\tau]} y_i \varepsilon_i \xrightarrow{d} B(\tau) \cdot \sigma_y \quad \text{i.i.d.}$$

$$\max_{[n\tau_1] \leq k \leq n-[n\tau_2]} \left[\mathcal{L}_n(\hat{\Phi}) - \mathcal{L}_n(\hat{\Phi}_1(n), \hat{\Phi}_2(n), k) \right]$$

$$\frac{\Delta}{\tau(1-\tau)} \max_{\tau_1 \leq \tau \leq 1-\tau_1} \frac{[B(\tau) - \tau B(1)]^2}{\tau(1-\tau)}$$

where $\tau_1 \in (0, \frac{1}{2})$.

Note that

$$\max_{\phi \leq k \leq n} \left[\mathcal{L}_n(\hat{\Phi}) - \mathcal{L}_n(\hat{\Phi}_1(n), \hat{\Phi}_2(n), k) \right] \xrightarrow{P} \infty$$

Lemma. Let $\{G_t: t=1, 2, \dots\}$ be i.i.d. with $E G_t = 0$ and $\text{Var}(G_t) = 1$. If $E|G_t|^{2+\epsilon} < \infty$, for some $\epsilon > 0$, then, for each $y \in (0, 1)$, any x

$$P\left(\frac{1}{a_n} \left[\max_{\log n \leq k \leq pn} \left| \frac{1}{\sqrt{k}} \sum_{t=1}^k G_t \right|^2 - b_n \right] \leq x\right) \rightarrow e^{-e^{-\frac{x}{1/2}}}$$

where $Q_n = \sqrt{b_n / (2 \log \log n)}$,

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$$b_n = \frac{[2 \log \log n + \log \log \log n / (2 - \log \Gamma(1/2))]^2}{2 \log \log n}$$

$\Gamma(x)$ — Gamma function.

Derive

$$W_n \stackrel{\text{max}}{=} \max_{m < k < n=m}$$

$$W_n(k) = L_n(\hat{F}) - L_n(\hat{F}_1(k), \hat{F}_2(k), k)$$

$$P\left(\max_{0 < k < n-1} \frac{W_n(k) - b_n}{Q_n} \leq x\right) \rightarrow e^{-2e^{-\frac{x}{2}}}$$

$$\text{as } n \rightarrow \infty.$$

Derive.

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$$S_1(n) = \frac{1}{\sqrt{n}} \sum_{t=1}^n y_t \varepsilon_t, \quad S_2(n) = \frac{1}{\sqrt{n-k}} \sum_{t=k+1}^n y_t \varepsilon_t.$$

To make it simple, Assume that $\sigma_y^2 = 1$.

Let $\mu \in (0, 0.5)$.

$$\left| \max_{\log n \leq k \leq \mu n} W_n(n) - \max_{\log n \leq k \leq \mu n} S_1(n) \right|$$

$$\leq \max_{\log n \leq k \leq \mu n} \left| W_n(n) - S_1(n) \right|$$

$$= \max_{\log n \leq k \leq \mu n} \left| \frac{n}{n-k} \left(\frac{1}{\sqrt{n}} \sum_{t=k+1}^n y_t \varepsilon_t \right)^2 - \left(\frac{1}{\sqrt{n}} \sum_{t=1}^n y_t \varepsilon_t \right)^2 \right|$$

$$\xrightarrow{2} \max_{0 \leq \tau \leq \mu} \left| \frac{[B(1) - B(\tau)]^2}{1-\tau} - B(1)^2 \right|$$

$$\xrightarrow{1} 0 \quad \text{as } \mu \rightarrow 0.$$

Thus, For $\forall \varepsilon > 0$,

$$\lim_{\mu \rightarrow 0} \sup_{n \rightarrow \infty} P \left(\left| \max_{\log n \leq k \leq \mu n} W_n(n) - \max_{\log n \leq k \leq \mu n} S_1(n) \right| \geq \varepsilon \right) = 0.$$

Similarly

$$\lim_{\mu \rightarrow 0} \lim_{h \rightarrow 0} \sup P \left(\left| \max_{k \leq n-k} W_n(k) - \max_{k \leq n-k} S_2^2(k) \right| > \varepsilon \right) = 0.$$

For $\forall \mu$,

$$\max_{\mu_n \leq k \leq n-\mu_n} W_n(k) \xrightarrow{d} \max_{\mu_0 \leq \tau \leq 1-\mu_0} \frac{[B(\tau) - \tau B(1)]^2}{\tau(1-\tau)}$$

$$P \left(\max_{\log n \leq k \leq n - \log n} \frac{W_n(k) - b_n}{b_n} \leq X \right)$$

$$= P \left(\max \left\{ \begin{array}{l} \max_{\log n \leq k \leq n - \log n} \frac{W_n(k) - b_n}{a_n} \\ \max_{\log n \leq k \leq n - \log n} \frac{W_n(k) - b_n}{a_n} \\ \max_{\mu_n \leq k \leq n - \mu_n} \frac{W_n(k) - b_n}{a_n} \end{array} \right\} \leq X \right)$$

$$= P \left(\max_{\substack{1 \leq k \leq n \\ \log k \leq n^\mu}} \frac{W(n-k) - b_n}{a_n} \leq x, \max_{\substack{1 \leq k \leq n \\ \log k \leq n^\mu}} \frac{W(k) - b_n}{a_n} \leq x \right) + o(\mu)$$

$$= P \left(\max_{\log n \leq k \leq n^\mu} \frac{S_1(k) - b_n}{a_n} \leq x, \max_{\log n \leq k \leq n^\mu} \frac{S_2(k) - b_n}{a_n} \leq x \right) + o(\mu)$$

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Strong Law of Large Numbers Principle:

$$\frac{1}{\sqrt{k}} \sum_{i=1}^k Y_i \xi_i = \frac{1}{\sqrt{k}} \sum_{i=1}^k G_{1+i} + o(k^{-\delta}) \text{ a.s.}$$

for some $\delta > 0$.

$$\frac{1}{\sqrt{k}} \sum_{i=-k}^{-1} Y_i \xi_i = \frac{1}{\sqrt{k}} \sum_{i=1}^k G_{2+i} + o(k^{-\delta}) \text{ a.s.}$$

where $\{G_{1+i}\}$ and $\{G_{2+i}\}$ are independent and

$G_{1+i} \sim N(0, \text{Var}(Y_1 \xi_1))$,

$G_{2+i} \sim \text{i.i.d. } N(0, \text{Var}(Y_1 \xi_1))$.

$$\left| \max_{\log n \leq k \leq np} S_1^2(k) - \max_{\log n \leq k \leq np} \left[\frac{1}{\sqrt{k}} \sum_{t=1}^k G_{1,t} \right] \right|$$

$$\leq \max_{\log n \leq k \leq np} \left| \sqrt{\log k} \left[S_1(k) - \frac{1}{\sqrt{k}} \sum_{t=1}^k G_{1,t} \right] \right|$$

$$\leq \frac{|S_1(k)| + \left| \frac{1}{\sqrt{k}} \sum_{t=1}^k G_{1,t} \right|}{\sqrt{\log k}} = o_p(1)$$

Similarly

$$\left| \max_{\log n \leq n-k \leq np} S_2^2(k) - \max_{\log n \leq n-k \leq np} \left[\frac{1}{\sqrt{n-k}} \sum_{t=1}^{n-k} G_{2,t} \right] \right| = o_p(1)$$

Thus,

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$$= P \left(\max_{\log n \leq k \leq np} \frac{\left| \frac{1}{\sqrt{k}} \sum_{t=1}^k G_{1,t} \right|^2 - b_n}{a_n} \geq \alpha \right)$$

$$\leq P \left(\max_{\log n \leq n-k \leq np} \frac{\left| \frac{1}{\sqrt{n-k}} \sum_{t=1}^{n-k} G_{2,t} \right|^2 - b_n}{a_n} \geq \alpha \right) + o_p(1)$$

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$$= P(\max_{\log n \leq k \leq n^p} \frac{|\frac{1}{k} \sum_{i=1}^k G_{ik}|^2 - h_n}{a_n} \leq x)$$

$$\times P(\max_{\log n \leq n-k \leq n^p} \frac{|\frac{1}{k+n} \sum_{i=1}^{k+n} G_{i, n+i}|^2 - h_n}{a_n} \leq x)$$

$$\Rightarrow e^{-2e^{-\frac{x}{2}}} \rightarrow e^{-2e^{-\frac{x}{2}}}$$

$$\text{Let } g_n = \log \log \log (\max_{1 \leq i \leq n} f_i, n)$$

$$\max_{0 \leq k \leq \log n} [S_1(k)]^2 = \max_{0 \leq k \leq \log n} \log \log k \left[\frac{k}{\int_k^{\log \log k} \frac{1}{t_1} dt_1} \right]^2$$

$$= O(\log \log \log n) = O(g_n).$$

$$a_n = 2 \log \log n + \log \log n / 2 - \log n^{1/2}$$

$$\max_{\log n \leq k \leq n^p} [S_1(k)]^2 / a_n = o(1).$$

Similarly

$$\max_{0 \leq n-1 \leq \log n} [S_2(n)]^2 / a_n = o(1).$$

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