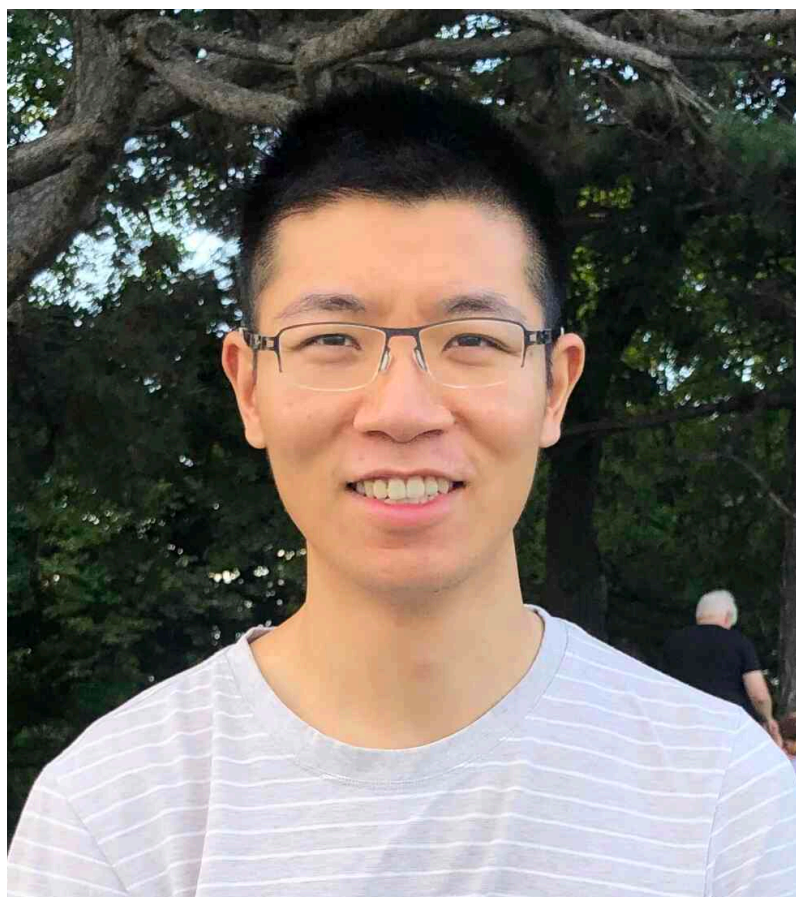


# Is Adaptive Robust Confidence Interval Possible?

Chao Gao  
University of Chicago

June 2024



**Yuetian Luo**

# Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\theta, 1)$$

# Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\theta, 1)$$

$$\mathbb{P} \left( \theta \in \left[ \bar{X} \pm \frac{1.96}{\sqrt{n}} \right] \right) \geq 0.95$$



# Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

# Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

**optimal  
estimation**

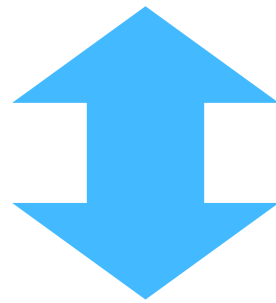
$$\mathbb{P} \left( \left| \hat{\theta}_{\text{median}} - \theta \right| \leq \sqrt{2\pi} \left( \frac{1.36}{\sqrt{n}} + \epsilon \right) \right) \geq 0.95$$

# Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

**optimal  
estimation**

$$\mathbb{P} \left( \left| \hat{\theta}_{\text{median}} - \theta \right| \leq \sqrt{2\pi} \left( \frac{1.36}{\sqrt{n}} + \epsilon \right) \right) \geq 0.95$$



**optimal  
CI**

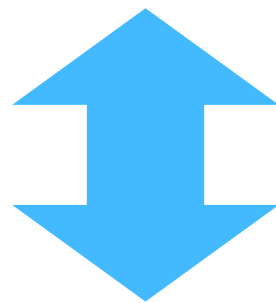
$$\mathbb{P} \left( \theta \in \left[ \hat{\theta}_{\text{median}} \pm \sqrt{2\pi} \left( \frac{1.36}{\sqrt{n}} + \epsilon \right) \right] \right) \geq 0.95$$

# Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

**optimal  
estimation**

$$\mathbb{P} \left( \left| \hat{\theta}_{\text{median}} - \theta \right| \leq \sqrt{2\pi} \left( \frac{1.36}{\sqrt{n}} + \epsilon \right) \right) \geq 0.95$$



**optimal  
CI**

$$\mathbb{P} \left( \theta \in \left[ \hat{\theta}_{\text{median}} \pm \sqrt{2\pi} \left( \frac{1.36}{\sqrt{n}} + \epsilon \right) \right] \right) \geq 0.95$$

known

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

random interval  $\hat{C} = \hat{C}(X_1, \dots, X_n)$

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

random interval  $\hat{C} = \hat{C}(X_1, \dots, X_n)$

**coverage**

$$\inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} \left( \theta \in \hat{C} \right) \geq 0.95$$

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

random interval  $\hat{C} = \hat{C}(X_1, \dots, X_n)$

**coverage**

$$\inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} \left( \theta \in \hat{C} \right) \geq 0.95$$

**length**

$$\inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} \left( \text{length} \left( \hat{C} \right) \leq r(n, \epsilon) \right) \geq 0.99$$



# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

random interval  $\hat{C} = \hat{C}(X_1, \dots, X_n)$

**coverage**  $\inf_{\epsilon \in [0, 1/2]} \inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} (\theta \in \hat{C}) \geq 0.95$

**length**  $\inf_{\epsilon \in [0, 1/2]} \inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} (\text{length}(\hat{C}) \leq r(n, \epsilon)) \geq 0.99$

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

**Theorem.** For any adaptive CI that satisfies coverage, its length must satisfy

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

**Theorem.** For any adaptive CI that satisfies coverage, its length must satisfy

$$r(n, \epsilon) \gtrsim \frac{1}{\sqrt{\log n}} + \frac{1}{\sqrt{\log(1/\epsilon)}}$$

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

**Theorem.** For any adaptive CI that satisfies coverage, its length must satisfy

$$r(n, \epsilon) \gtrsim \frac{1}{\sqrt{\log n}} + \frac{1}{\sqrt{\log(1/\epsilon)}}$$

compared with the non-adaptive rate  $\frac{1}{\sqrt{n}} + \epsilon$

# Adaptive Robust CI

**Theorem.** With  $F_n(t) = \frac{1}{n} \sum_{i \in [n]} \mathbb{I}\{X_i \leq t\}$ , the CI

# Adaptive Robust CI

**Theorem.** With  $F_n(t) = \frac{1}{n} \sum_{i \in [n]} \mathbb{I}\{X_i \leq t\}$ , the CI

$$\hat{C} = \left[ \begin{array}{l} \max_{t \in [1, \log n]} \left( F_n^{-1} (2 (1 - \Phi(t))) + t - \frac{1}{t} \right), \\ \min_{t \in [1, \log n]} \left( F_n^{-1} (1 - 2 (1 - \Phi(t))) - t + \frac{1}{t} \right) \end{array} \right]$$

# Adaptive Robust CI

**Theorem.** With  $F_n(t) = \frac{1}{n} \sum_{i \in [n]} \mathbb{I}\{X_i \leq t\}$ , the CI

$$\hat{C} = \left[ \max_{t \in [1, \log n]} \left( F_n^{-1}(2(1 - \Phi(t))) + t - \frac{1}{t} \right), \right. \\ \left. \min_{t \in [1, \log n]} \left( F_n^{-1}(1 - 2(1 - \Phi(t))) - t + \frac{1}{t} \right) \right]$$

satisfies coverage, and its length satisfies

$$r(n, \epsilon) \lesssim \frac{1}{\sqrt{\log n}} + \frac{1}{\sqrt{\log(1/\epsilon)}} \quad \text{w.h.p.}$$

# A Testing Problem



# A Testing Problem

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\theta, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\theta, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q$$

$$\inf_{\epsilon \in \{0, 1/2\}} \inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} \left( \theta \in \hat{C} \right) \geq 1 - \alpha$$

$$\inf_{\epsilon \in \{0, 1/2\}} \inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} \left( \text{length} \left( \hat{C} \right) \leq r(n, \epsilon) \right) \geq 1 - \beta$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\theta, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q$$

$$\phi = \mathbb{I}\{\theta + r \in \hat{C}(X)\}$$


$$\inf_{\epsilon \in \{0, 1/2\}} \inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} (\theta \in \hat{C}) \geq 1 - \alpha$$

$$\inf_{\epsilon \in \{0, 1/2\}} \inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} (\text{length}(\hat{C}) \leq r(n, \epsilon)) \geq 1 - \beta$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\theta, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q$$

$$\hat{C} = \{\theta + r : \phi_{\theta+r}(X) = 1\}$$

$$\phi = \mathbb{I}\{\theta + r \in \hat{C}(X)\}$$

$$\inf_{\epsilon \in \{0, 1/2\}} \inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} (\theta \in \hat{C}) \geq 1 - \alpha$$

$$\inf_{\epsilon \in \{0, 1/2\}} \inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} (\text{length}(\hat{C}) \leq r(n, \epsilon)) \geq 1 - \beta$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

$$\mathbb{E}_{H_0} \mathbb{I}\{X \geq t\} = \exp\left(-\frac{1}{2}t^2\right)$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

$$\mathbb{E}_{H_0} \mathbb{I}\{X \geq t\} = \exp\left(-\frac{1}{2}t^2\right)$$

$$\mathbb{E}_{H_1} \mathbb{I}\{X \geq t\} \geq \frac{1}{2} \exp\left(-\frac{1}{2}(t-r)^2\right)$$



# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

$$\mathbb{E}_{H_0} \mathbb{I}\{X \geq t\} = \exp\left(-\frac{1}{2}t^2\right) \geq \frac{1}{n}$$

$$\mathbb{E}_{H_1} \mathbb{I}\{X \geq t\} \geq \frac{1}{2} \exp\left(-\frac{1}{2}(t-r)^2\right)$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

$$\mathbb{E}_{H_0} \mathbb{I}\{X \geq t\} = \exp\left(-\frac{1}{2}t^2\right) \underset{\text{red}}{\geq} \frac{1}{n}$$

$$\mathbb{E}_{H_1} \mathbb{I}\{X \geq t\} \geq \frac{1}{2} \exp\left(-\frac{1}{2}(t-r)^2\right) \underset{\text{red}}{\gg} \exp\left(-\frac{1}{2}t^2\right)$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

$$\mathbb{E}_{H_0} \mathbb{I}\{X \geq t\} = \exp\left(-\frac{1}{2}t^2\right) \underset{\text{red}}{\geq} \frac{1}{n}$$

$$\mathbb{E}_{H_1} \mathbb{I}\{X \geq t\} \geq \frac{1}{2} \exp\left(-\frac{1}{2}(t-r)^2\right) \underset{\text{red}}{\gg} \exp\left(-\frac{1}{2}t^2\right)$$

**Set**  $t \asymp \sqrt{\log n}$ .

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

$$\mathbb{E}_{H_0} \mathbb{I}\{X \geq t\} = \exp\left(-\frac{1}{2}t^2\right) \underset{\text{red}}{\geq} \frac{1}{n}$$

$$\mathbb{E}_{H_1} \mathbb{I}\{X \geq t\} \geq \frac{1}{2} \exp\left(-\frac{1}{2}(t-r)^2\right) \underset{\text{red}}{\gg} \exp\left(-\frac{1}{2}t^2\right)$$

**Set**  $t \asymp \sqrt{\log n}$ . **When**  $r \gtrsim \frac{1}{\sqrt{\log n}}$ , can separate the two hypotheses.

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

$$q(x) \stackrel{?}{=} 2\phi(x) - \phi(x - r)$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

$$q(x) \propto (2\phi(x) - \phi(x - r)) \mathbb{I}\{|x| \leq t\}$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

$$q(x) \propto (2\phi(x) - \phi(x - r)) \mathbb{I}\{|x| \leq t\}$$

$$\chi^2 \left( \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q \parallel N(\theta, 1) \right) \ll \frac{1}{n}$$



# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

$$q(x) \propto (2\phi(x) - \phi(x - r)) \mathbb{I}\{|x| \leq t\}$$

$$\chi^2 \left( \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q \parallel N(\theta, 1) \right) \ll \frac{1}{n} \quad \exp \left( -\frac{1}{2}x^2 \right) \geq \frac{1}{2} \exp \left( -\frac{1}{2}(x - r)^2 \right) \text{ for all } |x| \leq t$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

$$q(x) \propto (2\phi(x) - \phi(x - r)) \mathbb{I}\{|x| \leq t\}$$

$$\chi^2 \left( \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q \middle\| N(\theta, 1) \right) \ll \frac{1}{n} \quad \exp \left( -\frac{1}{2}x^2 \right) \geq \frac{1}{2} \exp \left( -\frac{1}{2}(x - r)^2 \right) \text{ for all } |x| \leq t$$

Set  $t \asymp \sqrt{\log n}$ .

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(r, 1) + \frac{1}{2}Q$$

$$q(x) \propto (2\phi(x) - \phi(x - r)) \mathbb{I}\{|x| \leq t\}$$

$$\chi^2 \left( \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q \parallel N(\theta, 1) \right) \ll \frac{1}{n} \quad \exp \left( -\frac{1}{2}x^2 \right) \geq \frac{1}{2} \exp \left( -\frac{1}{2}(x - r)^2 \right) \text{ for all } |x| \leq t$$

Set  $t \asymp \sqrt{\log n}$ . When  $r \lesssim \frac{1}{\sqrt{\log n}}$ , cannot separate the two hypotheses.

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\theta, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\theta, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q$$

$$\phi_{\theta+r}(X) = \mathbb{I} \left\{ \frac{1}{n} \sum_{i=1}^n \mathbb{I}\{X_i > \theta + t\} > 1 - \Phi(t) + \frac{1}{\sqrt{n}} \right\}$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\theta, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q$$

$$\phi_{\theta+r}(X) = \mathbb{I} \left\{ \frac{1}{n} \sum_{i=1}^n \mathbb{I}\{X_i > \theta + t\} > 1 - \Phi(t) + \frac{1}{\sqrt{n}} \right\}$$

$$\{\theta + r : \phi_{\theta+r}(X) = 1\}$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\theta, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q$$

$$\phi_{\theta+r}(X) = \mathbb{I} \left\{ \frac{1}{n} \sum_{i=1}^n \mathbb{I}\{X_i > \theta + t\} > 1 - \Phi(t) + \frac{1}{\sqrt{n}} \right\}$$

$$\{\theta + r : \phi_{\theta+r}(X) = 1\} = \left\{ \theta + r : \theta < F_n^{-1} \left( \Phi(t) - \frac{1}{\sqrt{n}} \right) - t \right\}$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\theta, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q$$

$$\phi_{\theta+r}(X) = \mathbb{I} \left\{ \frac{1}{n} \sum_{i=1}^n \mathbb{I}\{X_i > \theta + t\} > 1 - \Phi(t) + \frac{1}{\sqrt{n}} \right\}$$

$$\begin{aligned} \{\theta + r : \phi_{\theta+r}(X) = 1\} &= \left\{ \theta + r : \theta < F_n^{-1} \left( \Phi(t) - \frac{1}{\sqrt{n}} \right) - t \right\} \\ &= \left\{ \theta : \theta < F_n^{-1} \left( \Phi(t) - \frac{1}{\sqrt{n}} \right) - t + r \right\} \end{aligned}$$



# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\theta, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q$$

$$\{\theta + r : \phi_{\theta+r}(X) = 1\} = \left\{ \theta : \theta < F_n^{-1} \left( \Phi(t) - \frac{1}{\sqrt{n}} \right) - t + r \right\}$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q$$

$$\{\theta + r : \phi_{\theta+r}(X) = 1\} = \left\{ \theta : \theta < F_n^{-1} \left( \Phi(t) - \frac{1}{\sqrt{n}} - \epsilon \right) - t + r \right\}$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q$$

$$\begin{aligned} \{\theta + r : \phi_{\theta+r}(X) = 1\} &= \left\{ \theta : \theta < F_n^{-1} \left( \Phi(t) - \frac{1}{\sqrt{n}} - \epsilon \right) - t + r \right\} \\ &= \left\{ \theta : \theta < F_n^{-1} (1 - 2(1 - \Phi(t))) - t + \frac{1}{t} \right\} \end{aligned}$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}N(\theta + r, 1) + \frac{1}{2}Q$$

$$\begin{aligned} \{\theta + r : \phi_{\theta+r}(X) = 1\} &= \left\{ \theta : \theta < F_n^{-1} \left( \Phi(t) - \frac{1}{\sqrt{n}} - \epsilon \right) - t + r \right\} \\ &= \left\{ \theta : \theta < F_n^{-1} (1 - 2(1 - \Phi(t))) - t + \frac{1}{t} \right\} \end{aligned}$$

$$\hat{C} = \left[ \begin{array}{l} \max_{t \in [1, \log n]} \left( F_n^{-1} (2(1 - \Phi(t))) + t - \frac{1}{t} \right), \\ \min_{t \in [1, \log n]} \left( F_n^{-1} (1 - 2(1 - \Phi(t))) - t + \frac{1}{t} \right) \end{array} \right]$$

What about other distributions?

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)\text{Laplace}(\theta, 1) + \epsilon Q$$

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)\text{Laplace}(\theta, 1) + \epsilon Q$$
$$\frac{1}{2} \exp(-|x - \theta|)$$

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)\text{Laplace}(\theta, 1) + \epsilon Q$$
$$\frac{1}{2} \exp(-|x - \theta|)$$

**Theorem.** For Laplace distribution, any adaptive CI that satisfies coverage must have length

$$r(n, \epsilon) \gtrsim 1$$



# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} L(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}L(r, 1) + \frac{1}{2}Q$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} L(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}L(r, 1) + \frac{1}{2}Q$$

$$q(x) = \exp(-|x|) - \frac{1}{2} \exp(-|x - r|) \geq 0 \text{ for all } x \in \mathbb{R}$$

# A Testing Problem

$$H_0 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} L(0, 1)$$

$$H_1 : X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \frac{1}{2}L(r, 1) + \frac{1}{2}Q$$

$$q(x) = \exp(-|x|) - \frac{1}{2} \exp(-|x - r|) \geq 0 \text{ for all } x \in \mathbb{R}$$

When  $r \leq \log 2$ , cannot separate the two hypotheses.

# Adaptive Robust CI

	<b>density</b>	<b>optimal estimation error</b>	<b>optimal CI length</b>
<b>t-distribution</b>	$(1 + x^2)^{-\frac{\nu+1}{2}}$		
<b>generalized Gaussian</b>	$\exp(- x ^\beta)$		
<b>mollifier</b>	$\exp\left(-\frac{1}{(1-x^2)^\beta}\right) \mathbb{I}\{ x  \leq 1\}$		
<b>Bates</b>	$U_1 + \dots + U_k \quad k \geq 2$		

# Adaptive Robust CI

	<b>density</b>	<b>optimal estimation error</b>	<b>optimal CI length</b>
<b>t-distribution</b>	$(1 + x^2)^{-\frac{\nu+1}{2}}$	$\frac{1}{\sqrt{n}} + \epsilon$	
<b>generalized Gaussian</b>	$\exp(- x ^\beta)$	$\frac{1}{\sqrt{n}} + \epsilon$	
<b>mollifier</b>	$\exp\left(-\frac{1}{(1-x^2)^\beta}\right) \mathbb{I}\{ x  \leq 1\}$	$\frac{1}{\sqrt{n}} + \epsilon$	
<b>Bates</b>	$U_1 + \dots + U_k \quad k \geq 2$	$\frac{1}{\sqrt{n}} + \epsilon$	

# Adaptive Robust CI

	<b>density</b>	<b>optimal estimation error</b>	<b>optimal CI length</b>
<b>t-distribution</b>	$(1 + x^2)^{-\frac{\nu+1}{2}}$	$\frac{1}{\sqrt{n}} + \epsilon$	1
<b>generalized Gaussian</b>	$\exp(- x ^\beta)$	$\frac{1}{\sqrt{n}} + \epsilon$	$\left(\frac{1}{\log n} + \frac{1}{\log(1/\epsilon)}\right)^{\frac{(\beta-1)_+}{\beta}}$
<b>mollifier</b>	$\exp\left(-\frac{1}{(1-x^2)^\beta}\right) \mathbb{I}\{ x  \leq 1\}$	$\frac{1}{\sqrt{n}} + \epsilon$	$\left(\frac{1}{\log n} + \frac{1}{\log(1/\epsilon)}\right)^{\frac{\beta+1}{\beta}}$
<b>Bates</b>	$U_1 + \dots + U_k \quad k \geq 2$	$\frac{1}{\sqrt{n}} + \epsilon$	$\left(\frac{1}{n} + \epsilon\right)^{1/k}$

# High-Dimensional Confidence Set

# Adaptive Robust CS

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, I_p) + \epsilon Q$$



# Adaptive Robust CS

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, I_p) + \epsilon Q$$

random set  $\hat{C} = \hat{C}(X_1, \dots, X_n)$

# Adaptive Robust CS

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, I_p) + \epsilon Q$$

random set  $\hat{C} = \hat{C}(X_1, \dots, X_n)$

**coverage**  $\inf_{\epsilon \in [0, 1/2]} \inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} (\theta \in \hat{C}) \geq 0.95$

# Adaptive Robust CS

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, I_p) + \epsilon Q$$

random set  $\hat{C} = \hat{C}(X_1, \dots, X_n)$

**coverage**  $\inf_{\epsilon \in [0, 1/2]} \inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} \left( \theta \in \hat{C} \right) \geq 0.95$

**length**  $\inf_{\epsilon \in [0, 1/2]} \inf_{\theta, Q} \mathbb{P}_{\epsilon, \theta, Q} \left( \text{radius} \left( \hat{C} \right) \leq r(n, p, \epsilon) \right) \geq 0.99$

# Adaptive Robust CS

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, I_p) + \epsilon Q$$

**Theorem.** For any adaptive CS that satisfies coverage, its length must satisfy

$$r(n, p, \epsilon) \gtrsim \frac{1}{\sqrt{\log(n/p)}} + \frac{1}{\sqrt{\log(1/\epsilon)}}$$

Moreover, there exists an algorithm that achieves the lower bound.

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

$$\hat{C} = \left[ \begin{array}{l} \max_{t \in [1, \log n]} \left( F_n^{-1} (2(1 - \Phi(t))) + t - \frac{1}{t} \right), \\ \min_{t \in [1, \log n]} \left( F_n^{-1} (1 - 2(1 - \Phi(t))) - t + \frac{1}{t} \right) \end{array} \right]$$

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

$$\hat{R} = \min_{t \in [1, \log n]} \left( F_n^{-1} (1 - 2(1 - \Phi(t))) - t + \frac{1}{t} \right) \\ - \max_{t \in [1, \log n]} \left( F_n^{-1} (2(1 - \Phi(t))) + t - \frac{1}{t} \right)$$

# Adaptive Robust CI

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, 1) + \epsilon Q$$

$$\hat{R} = \min_{t \in [1, \log n]} \left( F_n^{-1} (1 - 2(1 - \Phi(t))) - t + \frac{1}{t} \right) \\ - \max_{t \in [1, \log n]} \left( F_n^{-1} (2(1 - \Phi(t))) + t - \frac{1}{t} \right)$$

$$\hat{C} = \left[ \hat{\theta}_{\text{median}} - \hat{R}, \hat{\theta}_{\text{median}} + \hat{R} \right]$$



# Adaptive Robust CS

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, I_p) + \epsilon Q$$

# Adaptive Robust CS

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, I_p) + \epsilon Q$$

$$\hat{C} = \left\{ \theta : \|\theta - \hat{\theta}\| \leq \hat{R} \right\}$$

# Adaptive Robust CS

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, I_p) + \epsilon Q$$

$$\hat{C} = \left\{ \theta : \|\theta - \hat{\theta}\| \leq \hat{R} \right\}$$

Tukey's median  
LRV or DKKLMS

# Adaptive Robust CS

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, I_p) + \epsilon Q$$

$$\hat{C} = \left\{ \theta : \|\theta - \hat{\theta}\| \leq \hat{R} \right\}$$

$$F_n(u, t) = \frac{1}{n} \sum_{i=1}^n \mathbb{I}\{u^T X_i \leq t\}$$

Tukey's median  
LRV or DKKLMS

# Adaptive Robust CS

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} (1 - \epsilon)N(\theta, I_p) + \epsilon Q$$

$$\hat{C} = \left\{ \theta : \|\theta - \hat{\theta}\| \leq \hat{R} \right\}$$

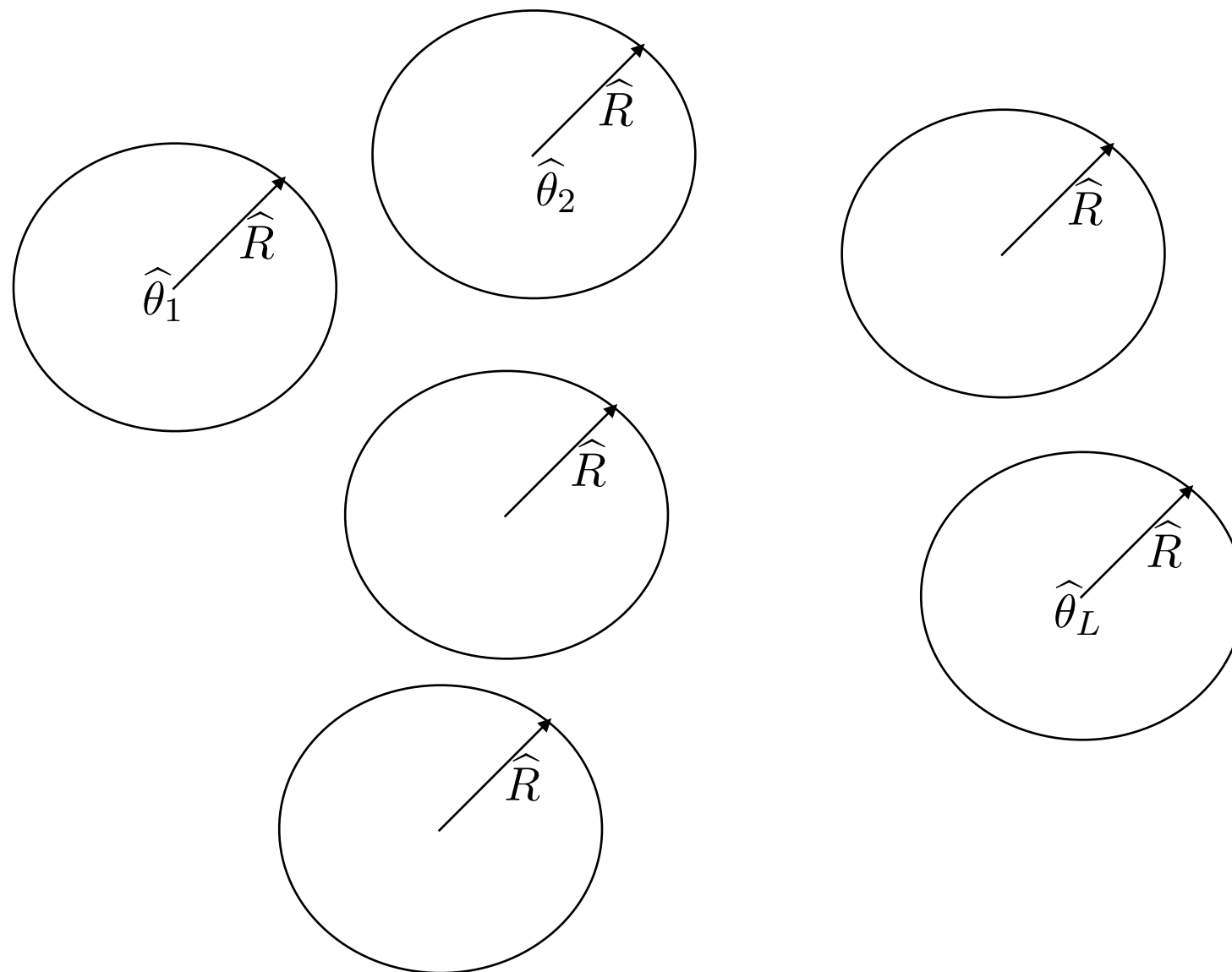
$$F_n(u, t) = \frac{1}{n} \sum_{i=1}^n \mathbb{I}\{u^T X_i \leq t\}$$

Tukey's median  
LRV or DKKLMS

$$\hat{R} = \max_{\|u\|=1} \left[ \min_{t \in [1, \log(n/p)]} \left( F_n^{-1}(u, 1 - 2(1 - \Phi(t))) - t + \frac{1}{t} \right) - \max_{t \in [1, \log(n/p)]} \left( F_n^{-1}(u, 2(1 - \Phi(t))) + t - \frac{1}{t} \right) \right]$$

# Robust CS and List-Decodable Estimation

# Adaptive Robust CS



Thank You