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**Proportion estimators, Lebesgue-Stieltjes  
integral equations, and concentration of  
measures**

- Background on proportion estimation
- Brief review on existing methods
- The new method
- Summary

## Large-scale multiple hypotheses testing

- Identify among hundreds of drugs those that may have a specific side effect
- Identify among thousands of genes those that might be associated with a disease
- Identify among millions of SNPs those that might be associated with the yield of a crop

## Four basic questions

Signals are entities that provide scientific information or that a scientist is searching for. Given a data set, one asks 4 basic questions:

- ① Are there any signals? (Testing)
- ② If there are signals, how many signals? (Testing)
- ③ If there are signals, where are they? (Testing)
- ④ If there are signals, what are their magnitudes? (Estimation)

## The proportion of false nulls

- Convention: “a signal” is identified with “a false null”, and “a noise” identified with “a true null”
- Among  $m$  null hypotheses, the proportion of false nulls, i.e., the signal proportion, is defined as

$$\pi_{1,m} = \frac{\text{Number of false nulls}}{m}$$

- Example: in gene-disease association study,  $\pi_{1,m}$  is the ratio of the number of disease-contributing genes to the total number of genes under investigation

- Dense regime:  $\liminf_{m \rightarrow \infty} \pi_{1,m} > 0$ , i.e., there is always a positive proportion of signals
- Moderately sparse regime:  $\pi_{1,m} = O(m^{-s})$  for  $0 < s < 0.5$
- Critically sparse regime:  $\pi_{1,m} = O(m^{-0.5})$
- Very sparse regime:  $\pi_{1,m} = O(m^{-s})$  for  $0.5 < s < 1$

## Formulation of estimation problem

- Data:  $\{z_i\}_{i=1}^m$  rv's each with mean or median  $\mu_i$
- Hypotheses:  $H_{i0} : \mu_i = \mu_0$  vs  $H_{i1} : \mu_i \neq \mu_0$
- Indices of signals:  $I_{1,m} = \{1 \leq i \leq m : \mu_i \neq \mu_0\}$
- Signal proportion:  $\pi_{1,m} = m_1 m^{-1}$ , where  $m_1 = |I_{1,m}|$
- Target: to consistently estimate  $\pi_{1,m}$

Four approaches to proportion estimation:

- ① Hard thresholding:  $H_{i0}$  is false when  $|z_i| \geq t$  for some  $t > 0$
- ② Deconvolving two-component mixture model:  $\{z_i\}_{i=1}^m$  identically distributed as

$$f(x) = (1 - \pi_{1,m})f_0(x) + \pi_{1,m}f_1(x)$$

- ③ Bounding an excess of uniform empirical process
- ④ Fourier transform, implemented for Gaussian family or mixtures with a Gaussian component



## A brief review on existing approaches

Disadvantages of the four approaches:

- Only Approaches 3 and 4 are consistent when  $\pi_{1,m}$  tends to zero
- Approach 4 is not applicable when CDFs of rv's do not consist at least one component from a location-shift family
- Consistency of Approach 3 requires CDFs to be absolutely continuous
- Consistency of Approaches 1 and 2 require restrictive modelling assumptions (e.g.,  $z_i$ 's corresponding to false nulls are identically distributed)

## The strategy: set-up

- Let  $\mathbf{z} = (z_1, \dots, z_m)$  and  $\boldsymbol{\mu} = (\mu_1, \dots, \mu_m)$
- Denote by  $F_{\mu_i}$  the CDF of  $z_i$  such that  $F_{\mu_i} \in \mathcal{F} = \{F_{\mu} : \mu \in U\}$ , where  $U$  has a non-empty interior
- Each  $F_{\mu}$  is uniquely determined by  $\mu$
- Recall:  $H_{i0} : \mu_i = \mu_0$  vs  $H_{i1} : \mu_i \neq \mu_0$

## The strategy: constructing “Phase function”

- For each  $\mu \in U$ , approximate  $1_{\{\mu \neq \mu_0\}}$  by  $\psi(t, \mu; \mu_0)$ ,  $t \in \mathbb{R}$  such that

$$\lim_{t \rightarrow \infty} \psi(t, \mu_0; \mu_0) = 1 \text{ and } \lim_{t \rightarrow \infty} \psi(t, \mu; \mu_0) = 0 \text{ for } \mu \neq \mu_0 \quad (1)$$

- Define “phase function”

$$\varphi_m(t, \boldsymbol{\mu}) = \frac{1}{m} \sum_{i=1}^m [1 - \psi(t, \mu_i; \mu_0)] \quad (2)$$

- The phase function:  $\lim_{t \rightarrow \infty} \varphi_m(t, \boldsymbol{\mu}) = \pi_{1,m}$  for any fixed  $m$  and  $\boldsymbol{\mu}$ , i.e., the “Oracle”

$$\Lambda_m(\boldsymbol{\mu}) = \lim_{t \rightarrow \infty} \varphi_m(t, \boldsymbol{\mu}) \quad (3)$$

## The strategy: constructing “Empirical phase function”

- Construct  $K : \mathbb{R}^2 \rightarrow \mathbb{R}$  that is independent of  $\mu \neq \mu_0$  and solves the Lebesgue-Stieltjes integral equation

$$\psi(t, \mu; \mu_0) = \int K(t, x; \mu_0) dF_\mu(x) \quad (4)$$

- Define “empirical phase function”

$$\hat{\varphi}_m(t, \mathbf{z}) = \frac{1}{m} \sum_{i=1}^m [1 - K(t, z_i; \mu_0)] \quad (5)$$

- The empirical phase function: for any fixed  $m, t$  and  $\boldsymbol{\mu}$

$$\mathbb{E}[\hat{\varphi}_m(t, \mathbf{z})] = \varphi_m(t, \boldsymbol{\mu}) \quad (6)$$

## The strategy: intuition for consistency

- Recall “phase function”  $\varphi_m(t, \boldsymbol{\mu})$  and  $\lim_{t \rightarrow \infty} \varphi_m(t, \boldsymbol{\mu}) = \pi_{1,m}$
- Recall “empirical phase function”  $\hat{\varphi}_m(t, \mathbf{z})$  and

$$\mathbb{E}[\hat{\varphi}_m(t, \mathbf{z})] = \varphi_m(t, \boldsymbol{\mu})$$

- Intuition:  $\hat{\varphi}_m(t, \mathbf{z})$  will consistently estimate  $\pi_{1,m}$ , i.e.,

$$\Pr\left(\left|\frac{\hat{\varphi}_m(t_m, \mathbf{z})}{\pi_{1,m}} - 1\right| \rightarrow 0\right) \rightarrow 1 \quad \text{as } m \rightarrow \infty \quad (7)$$

if

$$e_m(t_m) = \left|\hat{\varphi}_m(t_m, \mathbf{z}) - \varphi_m(t_m, \boldsymbol{\mu})\right| \rightarrow 0 \text{ at appropriate speed}$$

## The strategy: main difficulties in implementation

- Phase functions may not exist, i.e., no solution in  $\psi$  to

$$\lim_{t \rightarrow \infty} \psi(t, \mu_0; \mu_0) = 1 \text{ and } \lim_{t \rightarrow \infty} \psi(t, \mu; \mu_0) = 0 \text{ for } \mu \neq \mu_0$$

or in  $K$  (independent of  $\mu \neq \mu_0$ ) to

$$\psi(t, \mu; \mu_0) = \int K(t, x; \mu_0) dF_\mu(x)$$

- The oscillation

$$e_m(t_m) = |\hat{\varphi}_m(t_m, \mathbf{z}) - \varphi_m(t_m, \boldsymbol{\mu})|$$

may not converge to 0 when  $K$  is not a Lipschitz or bounded transform of  $\{z_i\}_{i=1}^m$

When  $\{z_i\}_{i=1}^m$  are mutually independent, uniformly consistent estimators  $\hat{\varphi}_m(t, \mathbf{z})$  are constructed respectively for the estimation problem in the settings where

- Random variables have Riemann-Lebesgue type characteristic functions (RL Type CFs)
- Random variables have CDFs that are members of natural exponential families with support  $\mathbb{N}$
- Random variables have CDFs that are members of natural exponential families with separable moments

## Construction I: families with Riemann-Lebesgue type CFs

- Recall  $\mathcal{F} = \{F_\mu : \mu \in U\}$  and let  $\hat{F}_\mu(t) = \int e^{itx} dF_\mu(x)$  be the CF of  $F_\mu$  where  $\iota = \sqrt{-1}$
- Let  $r_\mu$  be the modulus of  $\hat{F}_\mu$ , so that  $\hat{F}_\mu = r_\mu e^{ih_\mu}$

### Definition (Riemann-Lebesgue type CF)

If  $\{t \in \mathbb{R} : \hat{F}_{\mu_0}(t) = 0\} = \emptyset$ ,  $\sup_{t \in \mathbb{R}} \frac{r_\mu(t)}{r_{\mu_0}(t)} < \infty$  for each  $\mu \in U \setminus \{\mu_0\}$ , and

$$\lim_{t \rightarrow \infty} \frac{1}{t} \int_{[-t, t]} \frac{\hat{F}_\mu(y)}{\hat{F}_{\mu_0}(y)} dy = 0, \quad (8)$$

then  $\hat{\mathcal{F}} = \{\hat{F}_\mu : \mu \in U\}$  is said to be of “Riemann-Lebesgue type (RL type)” at  $\mu_0$  on  $U$ .



## Construction I: families with Riemann-Lebesgue type CFs

- Let  $\omega : [-1, 1] \rightarrow \mathbb{R}_+$  be bounded and Lebesgue-integrates to 1

Theorem (RVs with RL Type CFs)

Let  $\hat{\mathcal{F}}$  be of RL type and define  $K : \mathbb{R}^2 \rightarrow \mathbb{R}$  as

$$K(t, x; \mu_0) = \int_{[-1,1]} \frac{\omega(s) \cos(tsx - h_{\mu_0}(ts))}{r_{\mu_0}(ts)} ds. \quad (9)$$

Then

$$\psi(t, \mu; \mu_0) = \int_{[-1,1]} \omega(s) \frac{r_{\mu}(ts)}{r_{\mu_0}(ts)} \cos(h_{\mu}(ts) - h_{\mu_0}(ts)) ds. \quad (10)$$

## Construction I: application to location-shift families

- $\mathcal{F} = \{F_\mu : \mu \in U\}$  is a location-shift family if and only if  $z + \mu'$  has CDF  $F_{\mu+\mu'}$  whenever  $z$  has CDF  $F_\mu$  for  $\mu, \mu + \mu' \in U$

Corollary (Construction I for Location-shift families)

If  $\mathcal{F}$  is a location-shift family for which  $\{t \in \mathbb{R} : \hat{F}_{\mu_0}(t) = 0\} = \emptyset$ , then

$$\begin{aligned}\psi(t, \mu; \mu_0) &= \int K(t, y + (\mu - \mu_0); \mu_0) dF_{\mu_0}(y) \\ &= \int_{[-1,1]} \omega(s) \cos(ts(\mu - \mu_0)) ds.\end{aligned}$$

## Construction I: some examples

- Gaussian family Normal  $(\mu, \sigma^2)$  with mean  $\mu$  and std dev  $\sigma > 0$ , for which

$$K(t, x; \mu_0) = \int_{[-1,1]} \exp(2^{-1} t^2 s^2 \sigma^2) \omega(s) \exp(\iota t s (x - \mu_0)) ds$$

- Laplace family Laplace  $(\mu, 2\sigma^2)$  with mean  $\mu$  and std dev  $\sqrt{2}\sigma > 0$ , for which

$$K(t, x; \mu_0) = \int_{[-1,1]} (1 + \sigma^2 t^2 s^2) \omega(s) \exp(\iota t s (x - \mu_0)) ds$$

- Cauchy family Cauchy  $(\mu, \sigma)$  with median  $\mu$  and scale parameter  $\sigma > 0$ , for which

$$K(t, x; \mu_0) = \int_{[-1,1]} \exp(\sigma |t s|) \omega(s) \exp(\iota t s (x - \mu_0)) ds$$

## Definition (Uniform consistency class)

Given a family  $\mathcal{F}$ , the sequence of sets  $\mathcal{Q}_m(\boldsymbol{\mu}, t; \mathcal{F}) \subseteq \mathbb{R}^m \times \mathbb{R}$  for each  $m \in \mathbb{N}_+$  is called a “uniform consistency class” for the estimator  $\hat{\varphi}_m(t, \mathbf{z})$  if

$$\Pr\left(\sup_{\boldsymbol{\mu} \in \mathcal{Q}_m(\boldsymbol{\mu}, t; \mathcal{F})} \left| \pi_{1,m}^{-1} \sup_{t \in \mathcal{Q}_m(\boldsymbol{\mu}, t; \mathcal{F})} \hat{\varphi}_m(t, \mathbf{z}) - 1 \right| \rightarrow 0\right) \rightarrow 1.$$

- Define

$$\mathcal{B}_m(\rho) = \{\boldsymbol{\mu} \in \mathbb{R}^m : m^{-1} \sum_{i=1}^m |\mu_i - \mu_0| \leq \rho\} \text{ for some } \rho > 0$$

$$\text{and } u_m = \min\{|\mu_j - \mu_0| : \mu_j \neq \mu_0\}$$

## Construction I: applied to location-shift families

### Theorem (Uniform consistency for Construction I)

If  $\mathcal{F}$  is a “good” location-shift family, then a uniform consistency class is

$$\mathcal{Q}_m(\boldsymbol{\mu}, t; \mathcal{F}) = \left\{ \begin{array}{l} R_m(\rho) = O\left(m^{\vartheta'}\right), \tau_m \leq \gamma_m, u_m \geq \frac{\log \log m}{\gamma'' \tau_m}, \\ t \in [0, \tau_m], \lim_{m \rightarrow \infty} \pi_{1,m}^{-1} \Upsilon(q, \tau_m, \gamma_m, r_{\mu_0}) = 0 \end{array} \right\},$$

where  $R_m(\rho) \sim \rho$ ,  $\gamma_m = \gamma' \log m$

$$\Upsilon(q, \tau_m, \gamma_m, r_{\mu_0}) = \frac{2 \|\omega\|_{\infty} \sqrt{2q\gamma_m}}{\sqrt{m}} \sup_{t \in [0, \tau_m]} \int_{[0,1]} \frac{ds}{r_{\mu_0}(ts)}.$$

Moreover, for all sufficiently large  $m$ , with probability at least  $1 - o(1)$ ,

$$\sup_{\boldsymbol{\mu} \in \mathcal{B}_m(\rho)} \sup_{t \in [0, \tau_m]} |\hat{\varphi}_m(t, \mathbf{z}) - \varphi_m(t, \boldsymbol{\mu})| \leq \Upsilon(q, \tau_m, \gamma_m, r_{\mu_0}).$$

## Construction I: interpretation of uniform consistency class

For location-shift families, recall  $r_{\mu_0}$  as the modulus of CF  $\hat{F}_{\mu_0}$  for CDF  $F_{\mu_0}$  and

$$\Upsilon(q, \tau_m, \gamma_m, r_{\mu_0}) = \frac{2 \|\omega\|_{\infty} \sqrt{2q\gamma_m}}{\sqrt{m}} \sup_{t \in [0, \tau_m]} \int_{[0,1]} \frac{ds}{r_{\mu_0}(ts)}$$

- The speed of convergence and uniform consistency class for the estimator  $\hat{\varphi}_m(t, \mathbf{z})$  is mainly determined by  $r_{\mu_0}$  (not by  $r_{\mu}$  with  $\mu \neq \mu_0$ ) and

$$u_m = \min \{ |\mu_j - \mu_0| : \mu_j \neq \mu_0 \}$$

- Uniform consistency can be achieved when  $\{z_i\}_{i=1}^m$  do not have bounded first-order absolute moment (e.g., Cauchy family)
- It is unlikely that  $\hat{\varphi}_m(t, \mathbf{z})$  can be consistent for  $\pi_{1,m} = O(m^{-0.5})$

## A brief review on Natural Exponential Family (NEF)

- Let  $\beta$  be a positive non-Dirac Radon measure on  $\mathbb{R}$
- Let  $L(\theta) = \int e^{x\theta} \beta(dx)$  for  $\theta \in \mathbb{R}$  and  $\Theta$  be the maximal open set containing  $\theta$  such that  $L(\theta) < \infty$
- Assume  $\Theta$  is not empty and let  $\kappa(\theta) = \log L(\theta)$ . Then

$$\mathcal{F} = \{G_\theta : G_\theta(dx) = \exp(\theta x - \kappa(\theta)) \beta(dx), \theta \in \Theta\}$$

forms an NEF with respect to the basis  $\beta$

- Fact:  $\Theta$  is convex if it has a non-empty interior;  $L$  is analytic on the strip  $A_\Theta = \{z \in \mathbb{C} : \Re(z) \in \Theta\}$
- Fact:  $\mathcal{F}$  can be equivalently characterized as  $\mathcal{F} = \{F_\mu : \mu \in U\}$ , where  $\mu : \Theta \rightarrow U$  is the mean function with  $U = \mu(\Theta)$

## Constructions II and III: notational set-up

- $\mu_0$  is identified with  $\theta_0$ , and  $\mu = \mu(\theta)$  for  $\theta \in \Theta$
- Reuse  $\psi$  but take it as a function of  $t$  and  $\theta$
- Reuse  $K$  but take it as a function of  $t$  and  $\theta$ , such that

$$\psi(t, \theta; \theta_0) = \int K(t, x; \theta_0) dG_\theta(x) \text{ for } G_\theta \in \mathcal{F}.$$

- Let  $\boldsymbol{\theta} = (\theta_1, \dots, \theta_m)$ , the phase function

$$\varphi_m(t, \boldsymbol{\theta}) = \frac{1}{m} \sum_{i=1}^m [1 - \psi(t, \theta_i; \theta_0)]$$

and the empirical phase function

$$\hat{\varphi}_m(t, \mathbf{z}) = \frac{1}{m} \sum_{i=1}^m [1 - K(t, z_i; \theta_0)]$$



### Outside families with Riemann-Lebesgue type CFs

- Fourier transform used to construct  $K$  is no longer applicable
- $K$  does not always exist as a solution to

$$\psi(t, \theta; \theta_0) = \int K(t, x; \theta_0) dG_\theta(x) \text{ for } G_\theta \in \mathcal{F}$$

- Even if  $K$  does exist, the transform induced by  $K$  on  $\{z_i\}_{i=1}^m$  is no longer Lipschitz or bounded, which causes great difficulty in deriving concentration inequalities for the oscillation

$$e_m(t_m) = |\hat{\varphi}_m(t_m, \mathbf{z}) - \varphi_m(t_m, \boldsymbol{\mu})|$$

- Special functions and their asymptotics are needed since NEFs are often related to these functions

Assume the basis  $\beta$  for  $\mathcal{F}$  is discrete with support  $\mathbb{N}$ , i.e., there exists a positive sequence  $\{c_k\}_{k \geq 0}$  such that

$$\beta = \sum_{k=0}^{\infty} c_k \delta_k$$

- The power series  $H(z) = \sum_{k=0}^{\infty} c_k z^k$  with  $z \in \mathbb{C}$  must have a positive radius of convergence  $R_H$
- $h$  is the generating function (GF) of  $\beta$
- If  $\beta$  is a probability measure, then  $(-\infty, 0] \subseteq \Theta$  and  $R_H \geq 1$ , and vice versa

Theorem (Construction for discrete NEF with support  $\mathbb{N}$ )

Let  $\mathcal{F}$  be an NEF generated by  $\beta$  with support  $\mathbb{N}$ . For  $x \in \mathbb{N}$  and  $t \in \mathbb{R}$  set

$$K(t, x; \theta_0) = H(e^{\theta_0}) \int_{[-1,1]} \frac{(ts)^x \cos\left(\frac{\pi x}{2} - tse^{\theta_0}\right)}{H^{(x)}(0)} \omega(s) ds. \quad (11)$$

Then

$$\begin{aligned} \psi(t, \theta; \theta_0) &= \int K(t, x; \theta_0) dG_\theta(x) \\ &= \frac{H(e^{\theta_0})}{H(e^\theta)} \int_{[-1,1]} \cos\left(st(e^\theta - e^{\theta_0})\right) \omega(s) ds. \end{aligned} \quad (12)$$

- Poisson family Poisson ( $\mu$ ) with mean  $\mu > 0$ . The basis

$$\beta = \sum_{k=0}^{\infty} \frac{1}{k!} \delta_k \text{ with } L(\theta) = \exp(e^\theta) \text{ for } \theta \in \mathbb{R},$$

$$\mu(\theta) = e^\theta, H(z) = e^z \text{ with } R_H = \infty \text{ and } H^{(k)}(0) = 1 \text{ for all } k \in \mathbb{N}$$

- Negative Binomial family NegBinomial ( $\theta, n$ ) with  $\theta < 0$  and  $n \in \mathbb{N}_+$ . The basis is

$$\beta = \sum_{k=0}^{\infty} \frac{c_k^*}{k!} \delta_k \text{ with } c_k^* = \frac{(k+n-1)!}{(n-1)!} \text{ for } k \in \mathbb{N},$$

$$L(\theta) = (1 - e^\theta)^{-n} \text{ with } \theta < 0, \mu(\theta) = ne^\theta (1 - e^\theta)^{-1},$$

$$H(z) = (1 - z)^{-n} \text{ with } R_H = 1, \text{ and } H^{(k)}(0) = c_k^* \text{ for all } k \in \mathbb{N}$$

## Construction II: uniform consistency classes

Set  $\eta = e^\theta$  for  $\theta \in \Theta$  and  $\boldsymbol{\eta} = (e^{\theta_1}, \dots, e^{\theta_m})$ .

Theorem (Uniform consistency for Construction II)

Let  $\mathcal{F}$  be an NEF generated by  $\beta$  with support  $\mathbb{N}$  and  $\rho$  a finite, positive constant. If  $\frac{1}{c_k k!} \leq \frac{C}{k!}$  for all  $k \in \mathbb{N}$ , then a uniform consistency class is

$$\mathcal{Q}_{\text{II},1}(\boldsymbol{\theta}, t, \pi_{1,m}; \gamma) = \left\{ \begin{array}{l} \|\boldsymbol{\theta}\|_\infty \leq \rho, \pi_{1,m} \geq m^{(\gamma-1)/2}, \\ t = 2^{-1} \|\boldsymbol{\eta}\|_\infty^{-1/2} \gamma \log m, \\ \lim_{m \rightarrow \infty} t \min_{i \in I_{1,m}} |\eta_0 - \eta_i| = \infty \end{array} \right\}$$

for any fixed  $\gamma \in (0, 1]$ .

## Construction II: uniform consistency classes

- The GF of Poisson family has a constant derivative sequence at 0:  $H^{(k)}(0) = c_k k! = 1$  for all  $k \in \mathbb{N}$

### Theorem (Uniform consistency for Construction II)

*For Poisson family, a uniform consistency class is*

$$\mathcal{Q}_{\text{II},2}(\boldsymbol{\theta}, t, \pi_{1,m}; \gamma) = \left\{ \begin{array}{l} \|\boldsymbol{\theta}\|_{\infty} \leq \rho, \pi_{1,m} \geq m^{(\gamma'-1)/2}, \\ t = \sqrt{\|\boldsymbol{\eta}\|_{\infty}^{-1/2} \gamma \log m}, \\ \lim_{m \rightarrow \infty} t \min_{i \in I_{1,m}} |\eta_0 - \eta_i| = \infty \end{array} \right\}$$

*for any fixed  $\gamma \in (0, 1)$  and  $\gamma' > \gamma$ .*

## Construction III: continuous NEFs with separable moments

- Assume  $0 \in \Theta$ , so that  $\beta$  is a probability measure with finite moments of all orders
- Let

$$\tilde{c}_n(\theta) = \frac{1}{L(\theta)} \int x^n e^{\theta x} \beta(dx) = \int x^n dG_\theta(x) \quad \text{for } n \in \mathbb{N} \quad (13)$$

be the moment sequence for  $G_\theta \in \mathcal{F}$

- Fact: (13) is the Mellin transform of the measure  $G_\theta$
- Caution: Mellin transform is not Fourier transform even though they are connected

### Definition (NEF with separable moments)

If there exist two functions  $\zeta, \xi : \Theta \rightarrow \mathbb{R}$  and a sequence  $\{\tilde{a}_n\}_{n \geq 0}$  that satisfy the following:

- $\xi(\theta) \neq \xi(\theta_0)$  whenever  $\theta \neq \theta_0$ ,  $\zeta(\theta) \neq 0$  for all  $\theta \in U$ , and  $\zeta$  does not depend on any  $n \in \mathbb{N}$ ,
- $\tilde{c}_n(\theta) = \xi^n(\theta) \zeta(\theta) \tilde{a}_n$  for each  $n \in \mathbb{N}$  and  $\theta \in \Theta$ ,
- $\Psi(t, \theta) = \sum_{n=0}^{\infty} \frac{t^n \xi^n(\theta)}{\tilde{a}_n n!}$  is absolutely convergent pointwise in  $(t, \theta) \in \mathbb{R} \times \Theta$ ,

then the moment sequence  $\{\tilde{c}_n(\theta)\}_{n \geq 0}$  is called “separable” (at  $\theta_0$ ).



## Theorem (Construction for NEFs with Separable Moments)

Assume that the NEF  $\mathcal{F}$  has a separable moment sequence  $\{\tilde{c}_n(\theta)\}_{n \geq 0}$  at  $\theta_0$ . For  $t, x \in \mathbb{R}$  set

$$K(t, x; \mu_0) = \frac{1}{\zeta(\theta_0)} \int_{[-1,1]} \sum_{n=0}^{\infty} \frac{(-tsx)^n \cos\left(\frac{\pi}{2}n + ts\xi(\theta_0)\right)}{\tilde{a}_n n!} \omega(s) ds. \quad (14)$$

Then

$$\begin{aligned} \psi(t, \mu; \mu_0) &= \int K(t, x; \theta_0) dG_\theta(x) \\ &= \frac{\zeta(\theta)}{\zeta(\theta_0)} \int_{[-1,1]} \cos(ts(\xi(\theta_0) - \xi(\theta))) \omega(s) ds. \end{aligned} \quad (15)$$

## Construction III: one example

Let  $\Gamma$  be Euler's Gamma function. Gamma family Gamma  $(\theta, \sigma)$ :

- Its basis  $\frac{d\beta}{dv}(x) = \frac{x^{\sigma-1}e^{-x}}{\Gamma(\sigma)} \mathbf{1}_{(0,\infty)}(x) dx$  with  $\sigma > 0$
- $L(\theta) = (1 - \theta)^{-\sigma}$  and  $\mu(\theta) = \frac{\sigma}{1-\theta}$
- CDF and PDF:

$$\frac{dG_{\theta}}{dv}(x) = f_{\theta}(x) = (1 - \theta)^{\sigma} \frac{e^{\theta x} x^{\sigma-1} e^{-x}}{\Gamma(\sigma)} \mathbf{1}_{(0,\infty)}(x) \text{ for } \theta < 1$$

- The Gamma family subsumes Exponential family and central Chi-square family

## Construction III: one example

- Fact:

$$\tilde{c}_n(\theta) = (1-\theta)^\sigma \int_0^\infty \frac{e^{-y} y^{n+\sigma-1}}{\Gamma(\sigma)} \frac{dy}{(1-\theta)^{n+\sigma-1}} = \frac{\Gamma(n+\sigma)}{\Gamma(\sigma)} \frac{1}{(1-\theta)^n},$$

which implies  $\xi(\theta) = (1-\theta)^{-1}$ ,  $\tilde{a}_n = \frac{\Gamma(n+\sigma)}{\Gamma(\sigma)}$  and  $\zeta \equiv 1$

- Construction III:

$$K(t, x; \theta_0) = \int_{[-1,1]} \omega(s) \sum_{n=0}^{\infty} \frac{(-tsx)^n \Gamma(\sigma) \cos\left(\frac{\pi}{2}n + \frac{ts}{1-\theta_0}\right)}{n! \Gamma(\sigma+n)} ds,$$

for which

$$\psi(t, \theta; \theta_0) = \int_{[-1,1]} \cos\left(ts\left((1-\theta_0)^{-1} - (1-\theta)^{-1}\right)\right) \omega(s) ds$$

## Construction III: comparison with Construction I

- Recall Construction I applied to location-shift family:

$$\begin{aligned}\psi(t, \mu; \mu_0) &= \int K(t, x; \mu_0) dF_\mu(x) \\ &= \int K(t, y + (\mu - \mu_0); \mu_0) dF_{\mu_0}(y),\end{aligned}$$

where Fourier transform:  $K(t, x; \mu_0) \mapsto K(t, y + (\mu - \mu_0); \mu_0)$

- Construction III for Gamma family:

$$\begin{aligned}\psi(t, \theta; \theta_0) &= \int K(t, x; \theta_0) dG_\theta(x) \\ &= \int K\left(t, \frac{y}{1-\theta}; \theta_0\right) \beta(dy),\end{aligned}$$

where Mellin transform:  $K(t, x; \theta_0) \mapsto K(t, y(1-\theta)^{-1}; \theta_0)$

## Construction III: uniform consistency class

- Set  $u_{3,m} = \min_{1 \leq i \leq m} \{1 - \theta_i\}$  and  $\xi(\theta) = (1 - \theta)^{-1}$

### Theorem (Uniform Consistency of Construction III)

*Consider Gamma family with a fixed  $\sigma > 0$ . Let  $\rho > 0$  be a finite constant. If  $\sigma > 3/4$ , then a uniform consistency class is*

$$\mathcal{Q}_{\text{III}}(\boldsymbol{\theta}, t, \pi_{1,m}; \gamma) = \left\{ \begin{array}{l} \|\boldsymbol{\theta}\| \leq \rho, t = 4^{-1} \gamma u_{3,m} \log m, \\ \lim_{m \rightarrow \infty} u_{3,m} \log m = \infty, \\ \pi_{1,m} \geq m^{(\gamma-1)/2}, \\ \lim_{m \rightarrow \infty} t \min_{i \in I_{1,m}} |\xi(\theta_0) - \xi(\theta_i)| = \infty \end{array} \right\}$$

*for any fixed  $\gamma \in (0, 1]$ .*

## Theorem (Uniform Consistency of Construction III)

Consider Gamma family with a fixed  $\sigma > 0$ . Let  $\rho > 0$  be a finite constant. If  $\sigma \leq 3/4$ , then

$$\mathcal{Q}_{\text{III}}(\boldsymbol{\theta}, t, \pi_{1,m}; \gamma) = \left\{ \begin{array}{l} \|\boldsymbol{\theta}\| \leq \rho, t = 4^{-1} \gamma u_{3,m} \log m, \\ \pi_{1,m} \geq m^{(\gamma'-1)/2}, \\ \lim_{m \rightarrow \infty} t \min_{i \in I_{1,m}} |\xi(\boldsymbol{\theta}_0) - \xi(\boldsymbol{\theta}_i)| = \infty \end{array} \right\}$$

for any fixed  $\gamma \in (0, 1)$  and  $\gamma' > \gamma$ .

- Solutions to Lebesgue-Stieltjes integral equation can serve as universal construction for proportion estimators
- Under independence, the induced estimators are often uniformly consistent
- The uniform consistency class and speed of convergence for each such estimator are largely determined by
  - the modulus of the CF corresponding to the null parameter for families with Riemann-Lebesgue type CFs (including many location-shift families)
  - the radius of convergence of the generating function of discrete natural exponential families
  - the decaying speed of the moment sequence of continuous natural exponential families

## A few remaining tasks

- Construct estimators via solutions to Lebesgue-Stieltjes integral equations for  $\pi_{1,m}$  for composite null hypotheses such as
  - $H_{i0} : \mu_i \in (a, b)$  vs  $H_{i1} : \mu_i \notin (a, b)$
  - $H_{i0} : \mu_i \leq a$  vs  $H_{i1} : \mu_i > a$
- Uniform consistency of such estimators under dependence outside Gaussian families and their associated mixtures
- Classify natural exponential families with separable moment functions



- Classify distributions whose CFs satisfy

$$\{t \in \mathbb{R} : \hat{F}_\mu(t) = 0\} = \emptyset \quad \text{for } \mu \in U' \quad (16)$$

- Completely understand the relationships between families with Riemann-Lebesgue type CFs, those whose CFs satisfy (16), and infinitely divisible distributions
- Concentration inequalities for sums of non-Lipschitz, unbounded transforms of independent randomize variables

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- Xiongzhi Chen, *Estimators of the proportion of false null hypotheses: I “universal construction via Lebesgue-Stieltjes integral equations and uniform consistency under independence”*, arxiv preprint
- References in the above preprint