

Merging p-values under arbitrary dependence

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Econometrics and Optimal Transport
University of Washington, Seattle, USA, June 12–14, 2023

Agenda

- 1 Merging p-values
- 2 Admissibility
- 3 Rejection regions
- 4 Constructing admissible p-merging functions
- 5 Simulation results and summary

Simple question

Suppose that we are testing the **same hypothesis** using $K \geq 2$ tests and obtain p-values p_1, \dots, p_K . How can we combine them into a **single p-value**?

A question of a long history

- ▶ **Tippett'31, Pearson'33, Fisher'48**: assume independence
- ▶ The **Bonferroni correction**: minimum \times correction (K)
- ▶ We are interested in the case of **no dependence assumption**
- ▶ Can be used to test multiple hypotheses

Meta-analysis

A typical example from meta-analysis

TABLE 1

Data on 10 Studies of Sex Differences in Conformity Using the Fictitious Norm Group Paradigm

Study	Sample size		Effect size d	Student's t	Significance level p	$-2 \log p$	$\Phi^{-1}(p)$	$\log[p/(1-p)]$
	Control n^C	Experimental n^E						
1	118	136	0.35	2.78	0.0029	11.682	-2.758	-5.838
2	40	40	0.37	1.65	0.0510	5.952	-1.635	-2.923
3	61	64	-0.06	-0.33	0.6310	0.921	0.335	0.537
4	77	114	-0.30	-2.03	0.9783	0.044	2.020	3.809
5	32	32	0.70	2.80	0.0034	11.367	-2.706	-5.680
6	45	45	0.40	1.90	0.0305	6.978	-1.873	-3.458
7	30	30	0.48	1.86	0.0341	6.760	-1.824	-3.345
8	10	10	0.85	1.90	0.0367	6.608	-1.790	-3.266
9	70	71	-0.33	-1.96	0.9740	0.053	1.942	3.622
10	60	59	0.07	0.38	0.3517	2.090	-0.381	-0.612

The **sex differences** dataset, from p.35 of **Hedges/Olkin'85**

The value of no assumption

Why no independence assumption?

- ▶ A set of p-values is only **one vector**: no hope to test/verify any dependence model among them
- ▶ **Non-identifiability**: are we rejecting independence or the scientific hypothesis?
- ▶ **Efron'10**, **Large-scale Inference**, p50-p51:
“independence among the p-values ... usually an unrealistic assumption. ... even PRD [positive regression dependence] is unlikely to hold in practice.”

Merging functions

Let \mathcal{H} be a collection of atomless probability measures ...

Definition (p-variables and merging functions)

(i) A **p-variable** is a random variable P that satisfies

$$\sup_{\mathbb{P} \in \mathcal{H}} \mathbb{P}(P \leq \varepsilon) \leq \varepsilon, \quad \varepsilon \in (0, 1).$$

(ii) A **p-merging function** is an increasing Borel function $F : [0, \infty)^K \rightarrow [0, \infty)$ such that $F(P_1, \dots, P_K)$ is a p-variable for all p-variables P_1, \dots, P_K .

- ▶ Controlled type I error under **arbitrary dependence**

Merging functions

- ▶ \mathcal{U} : the set of all uniform $[0, 1]$ random variables under \mathbb{P}

For an increasing Borel $F : [0, \infty)^K \rightarrow [0, \infty)$, equivalent are:

- ▶ F is a p-merging function w.r.t. **some** collection \mathcal{H}
- ▶ F is a p-merging function w.r.t. **all** collections \mathcal{H}
- ▶ fixing \mathbb{P} , $F(\mathbf{P})$ is a p-variable for all $\mathbf{P} \in \mathcal{U}^K$
- ▶ fixing \mathbb{P} , for all $\varepsilon \in (0, 1)$, $\bar{\mathbb{P}}(F \leq \varepsilon) \leq \varepsilon$, where

$$\bar{\mathbb{P}}(F \leq \varepsilon) = \sup \{ \mathbb{P}(F(\mathbf{P}) \leq \varepsilon) : \mathbf{P} \in \mathcal{U}^K \}$$

It suffices to consider $\mathcal{H} = \{\mathbb{P}\}$ for a generic \mathbb{P} and \mathcal{U}^K

- ▶ **Multi-marginal OT problem**: $\sup \{ \mathbb{E}[1_{\{F(\mathbf{P}) \leq \varepsilon\}}] : \mathbf{P} \in \mathcal{U}^K \}$

Existing methods

Without **any assumptions** on the p-values p_1, \dots, p_K

- ▶ $p_{(1)}, \dots, p_{(K)}$ are the ascending order statistics
- ▶ The **Bonferroni** method

$$F(p_1, \dots, p_K) = Kp_{(1)}$$

- ▶ **Order-family** (O-family)

Rüger'78

$$G_{k,K} = (p_1, \dots, p_K) = \frac{K}{k} p_{(k)}$$

- ▶ **Simes–Hommel**

Hommel'83

$$H(p_1, \dots, p_K) = \ell_K \bigwedge_{k=1}^K \frac{K}{k} p_{(k)}; \quad \ell_K = \sum_{k=1}^K \frac{1}{k}$$

Precise merging functions

Definition (precise merging functions)

A p-merging function F is **precise** if, for all $\varepsilon \in (0, 1)$,

$$\bar{\mathbb{P}}(F \leq \varepsilon) = \varepsilon.$$

The Bonferroni method $F(p_1, \dots, p_K) = Kp_{(1)}$

$$\begin{aligned} \mathbb{P} \left(\bigwedge_{k=1}^K p_k \leq \varepsilon/K \right) &= \mathbb{P} \left(\bigcup_{k=1}^K \{p_k \leq \varepsilon/K\} \right) \\ &\leq \sum_{k=1}^K \mathbb{P}(Kp_k \leq \varepsilon) = \sum_{k=1}^K \frac{\varepsilon}{K} = \varepsilon. \end{aligned}$$

- ▶ Equality if $\{Kp_k \leq \varepsilon\}$, $k \in [K]$ are **mutually exclusive**

Merging p-values via averaging

- ▶ Generalized mean

Kolmogorov'30

$$M_{\phi,K}(p_1, \dots, p_K) = \phi^{-1} \left(\frac{\phi(p_1) + \dots + \phi(p_K)}{K} \right),$$

where $\phi : [0, 1] \rightarrow [-\infty, \infty]$ is continuous & strictly monotone

- ▶ **M-family**: for $r \in \mathbb{R} \setminus \{0\}$,

$$M_{r,K}(p_1, \dots, p_K) = \left(\frac{p_1^r + \dots + p_K^r}{K} \right)^{1/r}.$$

- ▶ $\phi(x) = \tan((x - \frac{1}{2})\pi)$: **Cauchy combination**

Liu/Xie'20

Merging p-values via averaging

Special cases:

- ▶ **Arithmetic:** $M_{1,K}(p_1, \dots, p_K) = \frac{1}{K} \sum_{k=1}^K p_k$
- ▶ **Harmonic:** $M_{-1,K}(p_1, \dots, p_K) = \left(\frac{1}{K} \sum_{k=1}^K \frac{1}{p_k} \right)^{-1}$
- ▶ **Quadratic:** $M_{2,K}(p_1, \dots, p_K) = \sqrt{\frac{1}{K} \sum_{k=1}^K p_k^2}$

Limiting cases:

- ▶ **Geometric:** $M_{0,K}(p_1, \dots, p_K) = \left(\prod_{k=1}^K p_k \right)^{1/K}$
- ▶ **Maximum:** $M_{\infty,K}(p_1, \dots, p_K) = \max(p_1, \dots, p_K)$
- ▶ **Minimum:** $M_{-\infty,K}(p_1, \dots, p_K) = \min(p_1, \dots, p_K)$

The cases $r \in \{-1, 0, 1\}$ are known as **Platonic means**.

Merging p-values via averaging

The arithmetic average $M_{1,K}(p_1, \dots, p_K) = \frac{1}{K} \sum_{k=1}^K p_k$ is **not a p-merging function** Rüschendorf'82, Meng'93

$$\bar{\mathbb{P}}(M_{1,K} \leq \varepsilon) = \min(2\varepsilon, 1).$$

- ▶ $\Rightarrow 2M_{1,K}$ is a precise p-merging function

Task. Find $b_{r,K} > 0$ such that (the M-family)

$$F_{r,K} = b_{r,K} M_{r,K} \text{ is precise}$$

- ▶ $M_{r,K}$ increases in $r \implies b_{r,K}$ should decrease in r .

Translation to a risk aggregation problem

For $\alpha \in (0, 1]$ and a random variable X , define

$$Q_\alpha(X) = \inf\{x \in \mathbb{R} : \mathbb{P}(X \leq x) \geq \alpha\}$$

and for a function $F : [0, 1]^K \rightarrow [0, \infty)$, define

$$\underline{Q}_\alpha(F) = \inf \left\{ Q_\alpha(F(\mathbf{P})) : \mathbf{P} \in \mathcal{U}^K \right\}.$$

Translation to a risk aggregation problem

Lemma 1

For $a > 0$, $r \in [-\infty, \infty]$, and $F = aM_{r,K}$, equivalent are:

- (i) F is a p -merging function, i.e., $\overline{\mathbb{P}}(F \leq \varepsilon) \leq \varepsilon$ for all $\varepsilon \in (0, 1)$;
- (ii) $\underline{Q}_\varepsilon(F) \geq \varepsilon$ for all $\varepsilon \in (0, 1)$;
- (iii) $\overline{\mathbb{P}}(F \leq \varepsilon) \leq \varepsilon$ for some $\varepsilon \in (0, 1)$;
- (iv) $\underline{Q}_\varepsilon(F) \geq \varepsilon$ for some $\varepsilon \in (0, 1)$.

The same conclusion holds if all \leq and \geq are replaced by $=$.

- ▶ In statistical practice one only needs to have $\overline{\mathbb{P}}(F \leq \varepsilon) \leq \varepsilon$ for a specific ε , e.g. 0.05, 0.01, ...

Translation to a risk aggregation problem

It boils down to calculate $\underline{Q}_\varepsilon(M_{r,K})$, or equivalently:

(i) for $r > 0$, aggregation of **Beta risks**

$$(\underline{Q}_\varepsilon(M_{r,K}))^r = \inf_{U_1, \dots, U_K \in \mathcal{U}} \left\{ Q_\varepsilon \left(\frac{1}{K} (U_1^r + \dots + U_K^r) \right) \right\}$$

(ii) for $r = 0$, aggregation of **exponential risks**

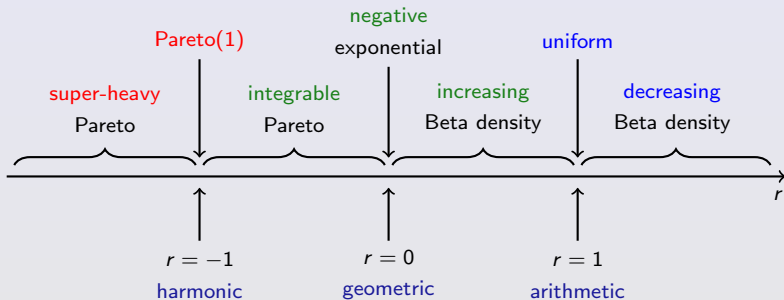
$$\log(\underline{Q}_\varepsilon(M_{r,K})) = \inf_{U_1, \dots, U_K \in \mathcal{U}} \left\{ Q_\varepsilon \left(\frac{1}{K} (\log U_1 + \dots + \log U_K) \right) \right\}$$

(iii) for $r < 0$, aggregation of **Pareto risks**

$$(\underline{Q}_\varepsilon(M_{r,K}))^r = \sup_{U_1, \dots, U_K \in \mathcal{U}} \left\{ Q_{1-\varepsilon} \left(\frac{1}{K} (U_1^r + \dots + U_K^r) \right) \right\}$$

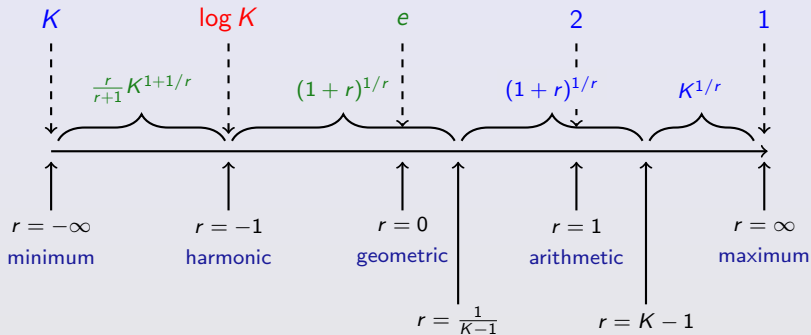
Translation to a risk aggregation problem

Breakdown of U^r (or $\log U$) for $r \in \mathbb{R}$



Main results summary

Constant multiplier in front of $M_{r,K}$



blue: precise; green: asymptotically precise; red: limit

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Admissible p-merging functions

For p-merging functions F and G :

- ▶ F is **symmetric** if $F(\mathbf{p})$ is invariant under permutation of \mathbf{p}
- ▶ F is **homogeneous** if $F(\lambda\mathbf{p}) = \lambda F(\mathbf{p})$ for all $\lambda \in (0, 1]$ and \mathbf{p} with $F(\mathbf{p}) \leq 1$
- ▶ F **dominates** G if $F \leq G$
- ▶ F is **admissible** if it is not dominated by any other one

Properties

- ▶ Admissible \implies precise, lower semicontinuous, grounded
- ▶ Any p-merging function is dominated by an admissible one

Simes function

The Simes function

$$S_K(p_1, \dots, p_K) = \bigwedge_{k=1}^K \frac{K}{k} p_{(k)}$$

Theorem 1

The Simes function S_K is the minimum over all symmetric p-merging functions.

- ▶ S_K is not valid p-merging (only valid under some assumptions)
- ▶ $H_K = \ell_K S_K$ is precise
- ▶ S_K is a lower bound for any symmetric improvement

Simes function

Proof sketch.

- ▶ Take any symmetric p-merging function F and $\mathbf{p} = (p_1, \dots, p_K)$
- ▶ Let $\alpha := S_K(\mathbf{p})/K \implies p_{(k)} \geq k\alpha$ for each k
- ▶ Symmetry and monotonicity of $F \implies$

$$F(\mathbf{p}) = F(p_{(1)}, \dots, p_{(K)}) \geq F(\alpha, 2\alpha, \dots, K\alpha) =: \beta$$

- ▶ Let Π be the set of all permutations of $(\alpha, 2\alpha, \dots, K\alpha)$, and $\mu = U(\Pi)$
- ▶ Take $(P_1, \dots, P_K) \sim K\alpha\mu + (1 - K\alpha)\delta_{(1, \dots, 1)}$
- ▶ For each k , $P_k \sim \sum_{k=1}^K \alpha\delta_{k\alpha} + (1 - K\alpha)\delta_1 \implies P_k$ is a p-variable
- ▶ F is a p-merging function \implies

$$\beta \geq \mathbb{P}(F(P_1, \dots, P_K) \leq \beta) \geq \mathbb{P}((P_1, \dots, P_K) \in \Pi) = K\alpha$$

- ▶ $F(\mathbf{p}) \geq K\alpha = S_K(\mathbf{p}) \implies S_K$ dominates all symmetric p-merging functions
- ▶ $S_K = \bigwedge_{k=1}^K G_{k,K}$

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Rejection regions of admissible p-merging functions

- ▶ The **rejection region** of a p-merging function F at level $\varepsilon \in (0, 1)$:

$$R_\varepsilon(F) := \left\{ \mathbf{p} \in [0, \infty)^K : F(\mathbf{p}) \leq \varepsilon \right\}$$

- ▶ A collection $\{R_\varepsilon \subseteq [0, \infty)^K : \varepsilon \in (0, 1)\}$ of increasing Borel lower sets induces a function $F : [0, \infty)^K \rightarrow [0, 1]$ via

$$F(\mathbf{p}) = \inf\{\varepsilon \in (0, 1) : \mathbf{p} \in R_\varepsilon\} \quad \text{with } \inf \emptyset = 1$$

- ▶ F is p-merging $\iff \mathbb{P}(\mathbf{P} \in R_\varepsilon) \leq \varepsilon$ for all $\varepsilon \in (0, 1)$, $\mathbf{P} \in \mathcal{U}^K$
- ▶ F is homogeneous $\implies R_\varepsilon(F) = \varepsilon A$ for some $A \subseteq [0, \infty)^K$.

Rejection regions of admissible p-merging functions

Admissibility \iff rejection region cannot be enlarged

- ▶ Precision of p-merging \iff classic OT

$$\text{Compute } \sup_{\mathbf{P} \in \mathcal{U}^K} \mathbb{E}[\mathbf{1}_A(\mathbf{P})]$$

- ▶ Admissibility (or optimality) \iff “reverse OT”

Given $\sup_{\mathbf{P} \in \mathcal{U}^K} \mathbb{E}[\mathbf{1}_A(\mathbf{P})] \leq \varepsilon$, find the largest $A \subseteq [0, 1]^K$

- ▶ Such A needs to be nested
- ▶ Techniques in OT can be very helpful

Rejection regions of admissible p-merging functions

Using e-values

Vovk/W.'21

- ▶ A **calibrator** is a decreasing function $f : [0, \infty) \rightarrow [0, \infty]$ satisfying $f = 0$ on $(1, \infty)$ and $\int_0^1 f(x)dx \leq 1$
- ▶ A calibrator f is **admissible** if it is upper semicontinuous, $f(0) = \infty$, and $\int_0^1 f(x)dx = 1$

Representation theorems

Let Δ_K be the standard K -simplex and write $\mathbf{p} = (p_1, \dots, p_K)$.

Theorem 2

For any *admissible homogenous* p -merging function F , there exist $(\lambda_1, \dots, \lambda_K) \in \Delta_K$ and *admissible calibrators* f_1, \dots, f_K such that

$$R_\varepsilon(F) = \varepsilon \left\{ \mathbf{p} \in [0, \infty)^K : \sum_{k=1}^K \lambda_k f_k(p_k) \geq 1 \right\} \text{ for } \varepsilon \in (0, 1). \quad (1)$$

Conversely, for any $(\lambda_1, \dots, \lambda_K) \in \Delta_K$ and calibrators f_1, \dots, f_K , (1) induces a *homogenous* p -merging function.

Representation theorems

Proof sketch.

- ▶ For decreasing functions g_1, \dots, g_K on $[0, \infty)$, denote by

$$\left(\bigoplus_{k=1}^K g_k \right) (x_1, \dots, x_K) := \sum_{k=1}^K g_k(x_k)$$

- ▶ Classic duality ($R_\varepsilon(F)$ is closed and F is precise)

$$\min \left\{ \sum_{k=1}^K \int_0^1 g_k(x) dx : \bigoplus_{k=1}^K g_k \geq \mathbb{1}_{R_\varepsilon(F)} \right\} = \max_{\mathbf{P} \in \mathcal{U}} \mathbb{P}(\mathbf{P} \in R_\varepsilon(F)) = \varepsilon$$

- ▶ Take $(g_1^\varepsilon, \dots, g_K^\varepsilon)$ such that $\bigoplus_{k=1}^K g_k^\varepsilon \geq \mathbb{1}_{R_\varepsilon(F)}$ and $\sum_{k=1}^K \int_0^1 g_k^\varepsilon(x) dx = \varepsilon$
- ▶ Choose g_k^ε to be non-negative and left-continuous
- ▶ Monotonicity

$$\max_{\mathbf{P} \in \mathcal{U}^K} \mathbb{P}(\mathbf{P} \in R_\varepsilon(F)) = \varepsilon \implies \max_{\mathbf{P} \in \mathcal{U}^K} \mathbb{P}(\varepsilon \mathbf{P} \in R_\varepsilon(F)) = 1$$

Representation theorems

Proof sketch (continued).

- ▶ Using duality again

$$\min \left\{ \sum_{k=1}^K \frac{1}{\varepsilon} \int_0^\varepsilon g_k(x) dx : \bigoplus_{k=1}^K g_k \geq \mathbb{1}_{R_\varepsilon(F)} \right\} = 1$$

$$\implies \sum_{k=1}^K \int_0^\varepsilon g_k^\varepsilon(x) dx \geq \varepsilon \implies g_k^\varepsilon(x) = 0 \text{ for } x > \varepsilon$$

- ▶ Define the set $A_\varepsilon := \{\mathbf{p} \in [0, \infty)^K : \sum_{k=1}^K g_k^\varepsilon(p_k) \geq 1\}$
- ▶ $\bigoplus_{k=1}^K g_k^\varepsilon \geq \mathbb{1}_{R_\varepsilon(F)} \implies R_\varepsilon(F) \subseteq A_\varepsilon$
- ▶ By Markov's inequality,

$$\sup_{\mathbf{P} \in \mathcal{U}^K} \mathbb{P} \left(\bigoplus_{k=1}^K g_k^\varepsilon(\mathbf{P}) \geq 1 \right) \leq \sup_{P \in \mathcal{U}} \sum_{k=1}^K \mathbb{E}[g_k^\varepsilon(P)] = \varepsilon$$

- ▶ Define a function F' with rejection region $R_\delta(F') = \delta \varepsilon^{-1} A_\varepsilon$ for $\delta \in (0, 1)$
- ▶ F' is a valid homogeneous p-merging function and $F' \leq F$

Representation theorems

Proof sketch (continued).

- ▶ Admissibility of $F \implies F = F'$, thus

$$R_\varepsilon(F) = A_\varepsilon = \varepsilon \left\{ \mathbf{p} \in [0, \infty)^K : \sum_{k=1}^K g_k^\varepsilon(\varepsilon p_k) \geq 1 \right\} \quad \text{for each } \varepsilon \in (0, 1)$$

- ▶ $\varepsilon^{-1}R_\varepsilon(F) = \varepsilon^{-1}A_\varepsilon$ does not depend on $\varepsilon \in (0, 1)$
- ▶ For a fixed $\varepsilon \in (0, 1)$ and each k , let $\lambda_k := \varepsilon^{-1} \int_0^\varepsilon g_k^\varepsilon(x) dx$ and $f_k : (0, \infty) \rightarrow \mathbb{R}$, $x \mapsto g_k^\varepsilon(\varepsilon x) / \lambda_k$ (if $\lambda_k = 0$, then let $f_k := 1$), and further set $f_k(0) = \infty$
- ▶ For each k with $\lambda_k \neq 0$,

$$\int_0^1 f_k(x) dx = \frac{\int_0^1 \varepsilon g_k^\varepsilon(\varepsilon x) dx}{\int_0^1 g_k^\varepsilon(x) dx} = \frac{\int_0^\varepsilon g_k^\varepsilon(x) dx}{\int_0^1 g_k^\varepsilon(x) dx} = 1$$

$\implies f_k$ is an admissible calibrator

- ▶ Converse statement: Markov's inequality

Representation theorems

Theorem 3

For any F that is *admissible within the family of homogenous symmetric p-merging functions*, there exists an *admissible calibrator* f such that

$$R_\varepsilon(F) = \varepsilon \left\{ \mathbf{p} \in [0, \infty)^K : \frac{1}{K} \sum_{k=1}^K f(p_k) \geq 1 \right\} \text{ for } \varepsilon \in (0, 1). \quad (2)$$

Conversely, for any calibrator f , (2) induces a homogenous symmetric p-merging function.

- ▶ We say f **induces** F if (2) holds
- ▶ Converse: **not true**; calibrator: **not unique**

Examples

Example 1

The p-merging function $F := G_{k,K}$, $k \in \{1, \dots, K\}$, is induced by the calibrator $f : x \mapsto (K/k)\mathbb{1}_{\{x \in [0, k/K]\}}$.

Example 2

In case $K = 2$, the p-merging function $F : \mathbf{p} \mapsto 2M_{1,2}(\mathbf{p})\mathbb{1}_{\{\min(\mathbf{p}) > 0\}}$ is induced by the admissible calibrator $f : x \mapsto (2 - 2x)_+$ on $(0, \infty)$ and $f(0) = \infty$.

- ▶ F is the zero-adjusted version of the arithmetic merging function
- ▶ F is not admissible (dominated by Bonferroni)

Connection to joint mixability

A necessary and sufficient condition for a calibrator f to induce a **precise p-merging function** via (1) is

$$\mathbb{P} \left(\frac{1}{K} \sum_{k=1}^K f(P_k) = 1 \right) = 1 \quad \text{for some } P_1, \dots, P_K \in \mathcal{U}. \quad (3)$$

- ▶ \implies **Joint mixability** (f specifies the quantile) Wang/W.'11'16
- ▶ Difficult to check for a given f in general
- ▶ For a convex f , (3) holds if and only if $f \leq K$ on $(0, 1]$
- ▶ Weaker than admissibility

Connection to e-tests

Connecting a p-test to an e-test: For a fixed $\varepsilon \in (0, 1)$, and an admissible p-merging function F :

$$\mathbf{p} \in R_\varepsilon(F) \iff \sum_{k=1}^K \lambda_k f_k \left(\frac{p_k}{\varepsilon} \right) \geq 1 \iff \sum_{k=1}^K \lambda_k f'_k(p_k) \geq \frac{1}{\varepsilon},$$

where $f'_k(x) := f_k(x/\varepsilon)/\varepsilon$. Four steps:

- (i) Calibrate all p-values to e-values via **admissible calibrators**
 f'_1, \dots, f'_n
- (ii) Merge the e-values via a **weighted arithmetic average**
- (iii) Calibrate the merged e-value to a p-value via $e \mapsto 1/e$
- (iv) Use the **resulting p-value** and the threshold ε for the test

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Sufficient conditions for admissibility

Theorem 4

Suppose that an admissible calibrator f is *strictly convex* or strictly concave on $(0, 1]$, $f(0+) \in (K/(K-1), K]$, and $f(1) = 0$. The p -merging function induced by f is admissible.

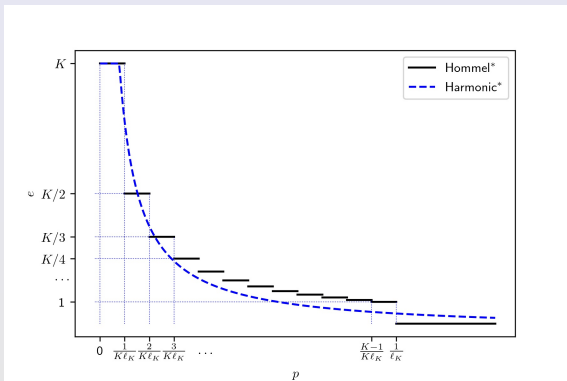
- ▶ Proof based on joint mixability
- ▶ Open question: can *strict convexity* be reduced to convexity?
- ▶ Conditions of this type are not necessary
- ▶ Admissibility holds true also for

$$g : x \mapsto f\left(\frac{x - \eta}{1 - K\eta}\right) \mathbb{1}_{\{x \in (\eta, 1 - (K-1)\eta)\}} + K \mathbb{1}_{\{x \in [0, \eta]\}}$$

Hommel's function

Define the **Hommel*** calibrator f by

$$f : x \mapsto \frac{K \mathbb{1}_{\{\ell_K x \leq 1\}}}{\lceil K \ell_K x \rceil}.$$



Hommel's function and the O-family

Theorem 5

The p -merging function $H_K \wedge 1$ is *dominated* (strictly if $K \geq 4$) by the p -merging function H_K^* induced by the Hommel* calibrator f ,

$$R_\varepsilon(H_K^*) = \varepsilon \left\{ \mathbf{p} \in [0, \infty)^K : \frac{1}{K} \sum_{k=1}^K f(p_k) \geq 1 \right\}, \quad \varepsilon \in (0, 1).$$

Moreover, H_K^* is always *admissible among symmetric* p -merging functions, and it is *admissible if K is not a prime number*.

- ▶ Primality appears in the proof due to factoring the set $[K]$
- ▶ H_K^* is not admissible for $K = 2, 3$ (we guess also 5)

Hommel's function and the O-family

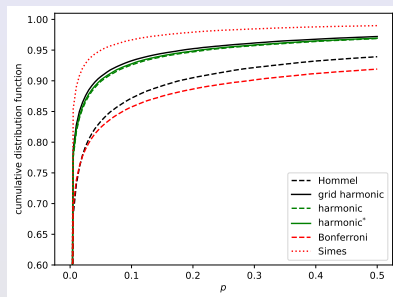
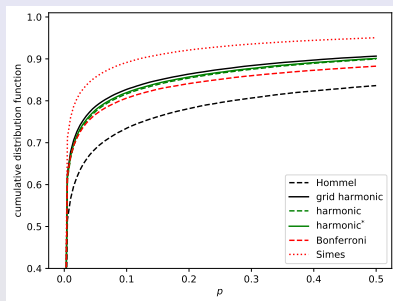
- ▶ The **Bonferroni** method is admissible
- ▶ Members $G_{k,K}$ of the **O-family** are **admissible** after truncation at 1 except for $k = K$
- ▶ Members $F_{r,K}$ of the **M-family** are **not admissible** except for $r = -\infty$
- ▶ $F_{r,K}$ can be strictly improved to $F_{r,K}^*$
- ▶ $F_{r,K}^*$ are admissible unless $r = 1 \Leftarrow$ non-strict convexity
- ▶ $F_{-1,K}^*$ is similar to H_K^*

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- 5 Simulation results and summary

Simulation results

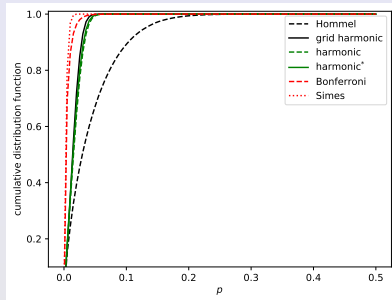
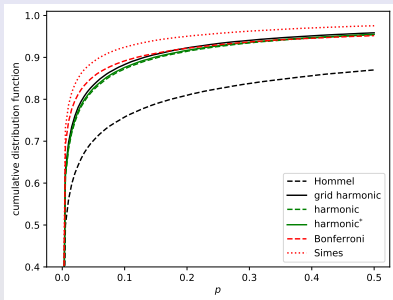
- ▶ Correlated z-tests
- ▶ $K = 10^6$ observations from $N(\mu, 1)$
- ▶ Pairwise correlation: 0.9
- ▶ Last observation: -0.9 correlation with all others
- ▶ $\mu = 0$ for null and $\mu = -5$ for alternative
- ▶ K_1 observations are drawn from the alternative; the rest from the null
- ▶ P-values are $\Phi(x)$
- ▶ $F_{-\infty, K}$ (Bonferroni); H_K (Hommel); $F_{-1, K}$ (harmonic); $F_{-1, K}^*$ (harmonic*'); H_K^* (grid harmonic'); S_K (Simes),

Simulation results



$K_1 = 10^3$ (left panel) and $K_1 = 10^4$ (right panel)

Simulation results



$K_1 = 10^5$ with correlation 0.5 (left panel) and 0 (right panel) in place of 0.9

Simulation results

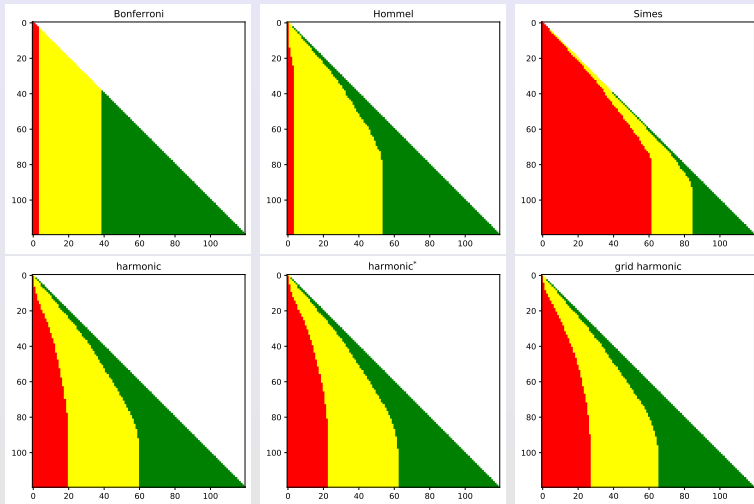
- ▶ GWGS discovery matrix Genovese/Wasserman'04; Goeman/Solari'11
- ▶ $DM_{i,j}$: a p-value for testing “there are less than j true discoveries among the i rejected hypotheses”
- ▶ $\mathcal{N} \subseteq [K]$: nulls
- ▶ Jointly validity: for each $\alpha \in (0, 1)$,

$$\mathbb{P}(\exists(i, j) \in D_\alpha : \#(R_i \setminus \mathcal{N}) < j) \leq \alpha$$

where R_i is the set of i hypotheses with smallest p-values and

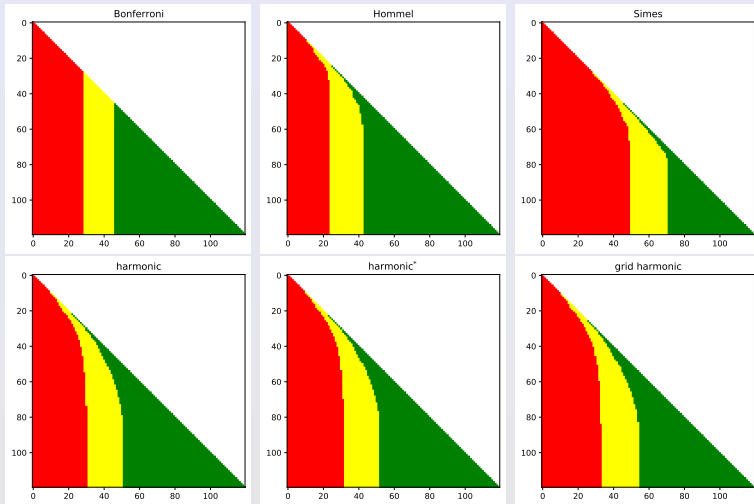
$$D_\alpha = \{(i, j) : DM_{i,j} \leq \alpha\}$$

Simulation results



GWGS discovery matrices with correlation 0.9 and significance levels 1% and 5%

Simulation results



GWGS discovery matrices with correlation 0.5 and significance levels 1% and 5%

Summary

Unsolved mathematical questions

- ▶ Homogeneity assumption in the representation results
- ▶ Strict convexity of calibrator in the sufficient condition for admissibility and $F_{-1,K}^*$
- ▶ Whether H_K^* is inadmissible for all prime K

More applications of multi-marginal OT and reverse OT?

Thank you

Thank you for your kind attention



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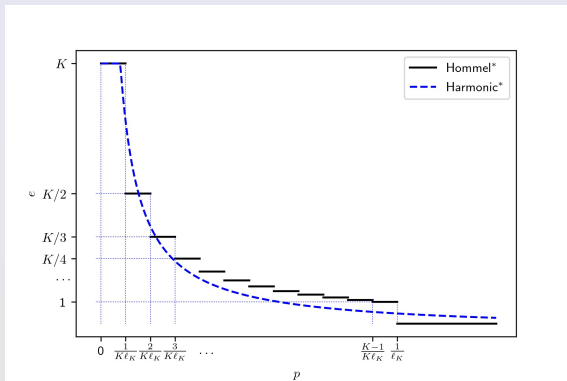
Bin Wang
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- ▶ Vovk/W., [Combining p-values via averaging](#) Biometrika, 2020
- ▶ Vovk/Wang/W., [Admissible ways of merging p-values under arbitrary dependence](#) Annals of Statistics, 2022

Hommel's function and the O-family

Define the **Hommel*** calibrator f by

$$f : x \mapsto \frac{K \mathbb{1}_{\{\ell_K x \leq 1\}}}{\lceil K \ell_K x \rceil}.$$



Hommel's function and the O-family

Theorem 6

The p -merging function $H_K \wedge 1$ is dominated (strictly if $K \geq 4$) by the p -merging function H_K^* induced by the Hommel* calibrator f ,

$$R_\varepsilon(H_K^*) = \varepsilon \left\{ \mathbf{p} \in [0, \infty)^K : \frac{1}{K} \sum_{k=1}^K f(p_k) \geq 1 \right\}, \quad \varepsilon \in (0, 1).$$

Moreover, H_K^* is always admissible among symmetric p -merging functions, and it is **admissible if K is not a prime number**.

Hommel's function and the O-family

Proof sketch.

- ▶ Recall: $H_K(\mathbf{p}) = \ell_K \bigwedge_{k=1}^K \frac{K}{k} p_{(k)}$ where $\ell_K = \sum_{i=1}^K \frac{1}{i}$.
- ▶ Induced by the calibrator $f \implies H_K^*$ is a p-merging function.
- ▶ Verify $H_K \geq H_K^*$: $H_K(\mathbf{p}) \leq \varepsilon \implies$ there exists m such that $K \ell_K p_{(m)} \leq \varepsilon \implies \#\{k : K \ell_K p_k / m \leq \varepsilon\} \geq m \implies$

$$\sum_{k=1}^K \frac{\mathbb{1}_{\{\ell_K p_k \leq \varepsilon\}}}{\lceil K \ell_K p_k / \varepsilon \rceil} \geq \sum_{k=1}^K \frac{1}{m} \mathbb{1}_{\{K \ell_K p_k / \varepsilon \leq m\}} = \frac{1}{m} \#\{k : K \ell_K p_k / m \leq \varepsilon\} \geq 1$$

$$\implies R_\varepsilon(H_K) \subseteq R_\varepsilon(H_K^*) \implies H_K \geq H_K^*.$$

- ▶ Check $H_K = H_K^*$ if and only if $K \leq 3$.

Hommel's function and the O-family

Proof sketch (continued).

- ▶ Suppose H_K^* is not admissible among symmetric p -merging functions.
- ▶ There exists a calibrator g satisfying

$$\left\{ \mathbf{p} \in [0, \infty)^K : \frac{1}{K} \sum_{k=1}^K f(p_k) \geq 1 \right\} \subsetneq \left\{ \mathbf{p} \in [0, \infty)^K : \frac{1}{K} \sum_{k=1}^K g(p_k) \geq 1 \right\}.$$

- ▶ Denote by $\tau := 1/(K\ell_K)$. For $x \in (0, K\tau]$, set $p_1 = \dots = p_m = x$ and $p_{m+1} = \dots = p_K > 1$, where $m := \lceil \tau x \rceil$.
- ▶ $f(x) = K/m \implies \sum_{k=1}^K f(p_k) = K \implies K \leq \sum_{k=1}^K g(p_k) = mg(x) \implies g(x) \geq K/m = f(x)$.
- ▶ $\int_0^{K\tau} g(x) dx \geq \int_0^{K\tau} f(x) dx = 1 \implies g = f$ almost everywhere on $[0, 1]$.
- ▶ f is left-continuous $\implies g \leq f$, a contradiction.
- ▶ The admissibility statement for non-prime K is much more complicated.

Hommel's function and the O-family

- ▶ $S_K \leq H_K^* \leq H_K \implies 1/\ell_K \leq H_K^*/H_K \leq 1$
- ▶ H_K^* may not be admissible for a prime K

Example 3

In case $K = 2$, $H_2^* = H_2 : (p_1, p_2) \mapsto (3p_{(1)}) \wedge (\frac{3}{2}p_{(2)})$ is strictly dominated by $F : (p_1, p_2) \mapsto (3p_1) \wedge (\frac{3}{2}p_2)$, which is a (non-symmetric) p-merging function because

$$\mathbb{P}(F(P_1, P_2) \leq \alpha) \leq \mathbb{P}\left(P_1 \leq \frac{1}{3}\alpha\right) + \mathbb{P}\left(P_2 \leq \frac{2}{3}\alpha\right) \leq \alpha.$$

Hommel's function and the O-family

Theorem 7

The p -merging function $\mathbf{p} \mapsto G_{k,K}(\mathbf{p}) \wedge \mathbb{1}_{\{\min(\mathbf{p}) > 0\}}$ is admissible for $k = 1, \dots, K - 1$, and it is admissible among symmetric p -merging functions for $k = K$.

The M-family

- ▶ $F_{r,K} = (b_{r,K} M_{r,K}) \wedge 1$
- ▶ For $r \neq \{-1, 0\}$ and $r < 1/(K-1)$, denote by c_r the unique number $c \in (0, 1/K)$ solving the equation

$$\frac{(K-1)(1-(K-1)c)^r + c^r}{K} = \frac{(1-(K-1)c)^{r+1} - c^{r+1}}{(r+1)(1-Kc)}$$

- ▶ c_{-1} and c_0 are limits of c_r
- ▶ Set $c_r := 0$ for $r \geq 1/(K-1)$
- ▶ Write $d_r := 1 - (K-1)c_r$

The M-family

Proposition 1

For $K \geq 3$ and $r \in (-\infty, \frac{1}{K-1})$,

$$b_{r,K} = 1/M_{r,K}(c_r, d_r, \dots, d_r).$$

- ▶ If $r < s$ and $rs > 0$, then

$$K^{1/s-1/r} b_{r,K} \leq b_{s,K} \leq b_{r,K}$$

The M-family

For $r < 0$:

- ▶ Rejection region

$$\begin{aligned}
 R_\varepsilon(F_{r,K}) &= \varepsilon \left\{ \mathbf{p} \in [0, \infty)^K : \frac{\sum_{k=1}^K p_k^r}{c_r^r + (K-1)d_r^r} \geq 1 \right\} \\
 &= \varepsilon \left\{ \mathbf{p} \in [0, \infty)^K : \sum_{k=1}^K \frac{p_k^r - d_r^r}{c_r^r - d_r^r} \geq 1 \right\}.
 \end{aligned}$$

- ▶ Define a calibrator

$$f_r : x \mapsto K \left(\frac{x^r - d_r^r}{c_r^r - d_r^r} \wedge 1 \right)_+.$$

- ▶ f_r is strictly convex on $[c_r, d_r]$.

The M-family

Let $F_{r,K}^*$ be the p-merging function induced by f_r , i.e.,

$$R_\varepsilon(F_{r,K}^*) = \varepsilon \left\{ \mathbf{p} \in [0, \infty)^K : \sum_{k=1}^K \left(\frac{p_k^r - d_r^r}{c_r^r - d_r^r} \right)_+ \geq 1 \right\}, \quad \varepsilon \in (0, 1).$$

- ▶ $R_\varepsilon(F_{r,K}) \subset R_\varepsilon(F_{r,K}^*)$
- ▶ $F_{r,K}^*$ is admissible

The M-family

Theorem 8

For $K \geq 3$ and $r \in (-\infty, K - 1)$, $F_{r,K}$ is strictly dominated by the p -merging function $F_{r,K}^*$ defined via, for $\mathbf{p} \in (0, \infty)^K$ and $\varepsilon \in (0, 1)$,

$$F_{r,K}^*(\mathbf{p}) \leq \varepsilon \iff F_{r,K}(\mathbf{p} \wedge (\varepsilon d_r \mathbf{1})) \leq \varepsilon \text{ or } \min(\mathbf{p}) = 0.$$

Moreover, $F_{r,K}^*$ is admissible unless $r = 1$.

The M-family

Proposition 2

For $K \geq 3$ and $\mathbf{p} \in [0, \infty)^K$, we have, if $r \in (-\infty, 1/(K-1))$,

$$F_{r,K}^*(\mathbf{p}) = \left(\bigwedge_{m=1}^K \frac{M_{r,m}(p_{(1)}, \dots, p_{(m)})}{M_{r,m}(c_r, d_r, \dots, d_r)} \right) \wedge 1,$$

and, if $r \in [1/(K-1), K-1)$, with the convention $\cdot/0 = \infty$,

$$F_{r,K}^*(\mathbf{p}) = \left(\bigwedge_{m=1}^K \frac{M_{r,m}(p_{(1)}, \dots, p_{(m)})}{\left(1 - \frac{rK}{(r+1)m}\right)_+} \right) \wedge \mathbb{1}_{\{p_{(1)} > 0\}}.$$

The M-family

Proposition 3

For $r < s$, $K \geq 2$ and $a, b > 0$, the following statements hold.

- (i) $aM_{r,K} \leq bM_{s,K}$ if and only if $a \leq b$.
- (ii) $bM_{s,K} \leq aM_{r,K}$ if and only if $rs > 0$ and $aK^{-1/r} \geq bK^{-1/s}$.

Proposition 4

Suppose $r \neq s$. If $K = 2$, $F_{r,K} \geq F_{s,K}$ if and only if $1 \leq r < s$ or $s < r \leq 1$. If $K \geq 3$, $F_{r,K} \geq F_{s,K}$ if and only if $K - 1 \leq r < s$.

Magnitude of improvement

Proposition 5

For $K \geq 3$, we have

$$\inf_{\mathbf{p} > \mathbf{0}} \frac{F_{1,K}^*(\mathbf{p})}{F_{1,K}(\mathbf{p})} = \inf_{\mathbf{p} > \mathbf{0}} \frac{F_{0,K}^*(\mathbf{p})}{F_{0,K}(\mathbf{p})} = 0, \quad \inf_{\mathbf{p} > \mathbf{0}} \frac{F_{-1,K}^*(\mathbf{p})}{F_{-1,K}(\mathbf{p})} = 1 - (K-1)c_{-1},$$

and

$$\min_{\mathbf{p} > \mathbf{0}} \frac{H_K^*(\mathbf{p})}{H_K(\mathbf{p})} = \min \left\{ t > 0 : \sum_{k=1}^K \frac{\mathbb{1}_{\{t \geq k/K\}}}{\lceil k/t \rceil} \geq 1 \right\} =: \gamma_K.$$

Moreover, $c_{-1} \sim 1/(K \log K)$ and $\gamma_K \sim 1/\log K$ as $K \rightarrow \infty$.

Magnitude of improvement

- ▶ $F_{-1,K}^*$ improves $F_{-1,K}$ only by a factor $1 - 1/\log K \sim 1$
- ▶ H_K^* can improve H_K by a significant factor of $1/\log K$
- ▶ $H_K^*(\mathbf{p})/H_K(\mathbf{p}) = \gamma_K$ is attained by $\mathbf{p} = (\alpha, 2\alpha, \dots, K\alpha)$ for $\alpha \in (0, 1/K\ell_K]$.
- ▶ Since $H_K = \ell_K S_K$ and

$$\gamma_K \sim 1/\log K \sim 1/\ell_K,$$

H_K^* performs similarly to the Simes function S_K for some values of \mathbf{p} above