

QuTE: decentralized multiple testing on sensor networks with FDR control

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Outline

- Motivation & Background
 - FDR control: Distributed VS Centralized
- QuTE: Query-Test-Exchange
 - Methodology
 - Performance guarantee: proof of two theorems
 - Multi-round QuTE
- Simulation
 - Erdos-Renyi random graph
 - Planar grid
- Discussion



Background

FDR(False Discovery Rate)

$$FDR = \mathbb{E}\left[\frac{V}{\max\{R,1\}}\right]$$

V: # of fasle discoveries

R: # total discoveries

- BH procedure
 - Order the *p*-values: $P_{(1)} \le P_{(2)} \le \cdots \le P_{(n)}$
 - Reject the smallest \hat{k} *p*-values: $\hat{k} = \max\{k: P_{(k)} \leq \frac{k}{n}\alpha\}$
 - Control at level $FDR \leq \alpha$



Motivation

- Centralized control
 - A central entity with access to all N p-values
 - Perform BH procedure
- Decentralized control
 - Wireless senor networks
 - Low power --> limited communication range
 - Each sensor has access to only a subset of p-values
 - Overall FDR control ?



Query-Test-Exchange (QuTE)

- Problem setup
 - Network represented by a graph: $G=(\mathcal{V},\mathcal{E})$
 - Agent a is responsible for testing: $\mathcal{H}_a = \{H_{a,i}\}_{i=1}^{n_a}$
 - True null: \mathcal{H}_a^0
 - Total # of hypotheses: $N = \sum_{a \in \mathcal{V}} n_a$
 - Overall true null: $\mathcal{H}^0 = \bigcup_{a \in \mathcal{V}} \mathcal{H}^0_a$
- Communication model
 - Nodes only communicate with their neighbors
 - Perfect p-value transmission: no noise & quantization
 - Static topology, p-value, etc.



QuTE---Methodology

- Initialization
 - Computes local *p*-vector: $\mathbf{P}_a = (P_{a,1}, P_{a,2}, \cdots, P_{a,n_a})$
- Step 1: Query
 - Each agent queries its neighbors for their p-vectors
 - Denote S_a as the set of p-values agent a gets after query
- Step 2: Test
 - Agent a runs BH procedure on the $|\mathcal{S}_a|$ p-values
 - Target level $\alpha_a := lpha rac{|\mathcal{S}_a|}{N}$
 - Reject all *p*-values smaller than $\alpha_a \frac{\widehat{k}^a}{|\mathcal{S}_a|} = (\alpha \frac{|\mathcal{S}_a|}{N}) \frac{\widehat{k}^a}{|\mathcal{S}_a|} = \alpha \frac{\widehat{k}^a}{N}$
- Step 3: Exchange
 - Exchange rejection decisions among neighbors



QuTE---Methodology

- Query
- Test
 - Each agent/node tests all the local p-values & those available at its neighbors (assume that no hypothesis is tested at two or more nodes, i.e., \mathcal{H}_a 's are non-overlapping)
 - There might be <u>computation redundancy</u> (explain later)

Exchange

- An agent rejects any local hypothesis rejected by its own test, or by the tests of any of its neighbors
- Hypothesis rejected by any node will be rejected in the overall FDR control



Theorem 1 Suppose that the p-values are independent, or positively dependent. Then for any graph topology, the QuTE algorithm achieves FDR control at level $\alpha \frac{|\mathcal{H}_0|}{N}$.

- Performing local BH results in overall FDR control
- Each agent only gather information from neighbors
- Two extremes
 - >> empty graph: Bonferroni test ($\alpha \frac{|\mathcal{S}_a|}{N} = \frac{\alpha}{N}$)
 - >> complete graph: centralized BH



Theorem 1 Suppose that the p-values are independent, or positively dependent. Then for any graph topology, the QuTE algorithm achieves FDR control at level $\alpha \frac{|\mathcal{H}_0|}{N}$.

Proof: denote \widehat{k}^{qute} as the overall # of rejections

$$\operatorname{define} \widehat{k}^{(a)} = \max_{s \in \operatorname{Neighbors}(a) \cup \{a\}} \widehat{k}^s$$

>> maximum # of rejections among a and its neighbors

[Claim] $P_{a,j}$ is rejected iff. $P_{a,j} \leq \alpha \frac{\widehat{k}^{(a)}}{N}$ (since is rejected if it is locally rejected, or rejected by any of a's neighbors)



Theorem 1 Suppose that the p-values are independent, or positively dependent. Then for any graph topology, the QuTE algorithm achieves FDR control at level $\alpha^{\frac{|\mathcal{H}_0|}{N}}$.

Proof: [Claim] $P_{a,j}$ is rejected iff. $P_{a,j} \leq \alpha \frac{k^{(a)}}{N}$.

$$FDR = \mathbb{E}\left[\frac{V}{R}\right] = \mathbb{E}\left[\frac{\sum_{a \in \mathcal{V}} \sum_{j \in \mathcal{H}_a^0} \mathbb{1}\left\{P_{a,j} \le \alpha \frac{\widehat{k}^{(a)}}{N}\right\}}{\widehat{k}^{qute}}\right]$$

$$\stackrel{(1)}{=} \sum_{a \in \mathcal{V}} \sum_{j \in \mathcal{H}_a^0} \mathbb{E} \left[\frac{\mathbb{1} \left\{ P_{a,j} \leq \alpha \frac{\widehat{k}^{(a)}}{N} \right\}}{\widehat{k}^{qute}} \right]$$

$$\stackrel{(2)}{\leq} \sum_{a \in \mathcal{V}} \sum_{j \in \mathcal{H}_a^0} \mathbb{E} \left[\frac{\mathbb{1} \left\{ P_{a,j} \leq \alpha \frac{\widehat{k}^{(a)}}{N} \right\}}{\widehat{k}^{(a)}} \right] \qquad (2) : \widehat{k}^{(a)} \leq \widehat{k}^{qute}, \, \forall a \in \mathcal{V}$$

$$(2): \widehat{k}^{(a)} \le \widehat{k}^{qute}, \, \forall a \in \mathcal{V}$$



Proof:
$$FDR = \mathbb{E}\left[\frac{V}{R}\right] = \mathbb{E}\left[\frac{\sum_{a \in \mathcal{V}} \sum_{j \in \mathcal{H}_a^0} \mathbb{1}\left\{P_{a,j} \leq \alpha \frac{\widehat{k}^{(a)}}{N}\right\}}{\widehat{k}^{qute}}\right]$$

$$\stackrel{(1)}{=} \sum_{a \in \mathcal{V}} \sum_{j \in \mathcal{H}_a^0} \mathbb{E}\left[\frac{\mathbb{1}\left\{P_{a,j} \leq \alpha \frac{\widehat{k}^{(a)}}{N}\right\}}{\widehat{k}^{qute}}\right]$$

$$\stackrel{(2)}{\leq} \sum_{a \in \mathcal{V}} \sum_{j \in \mathcal{H}_a^0} \mathbb{E} \left[\frac{\mathbb{1} \left\{ P_{a,j} \leq \alpha \frac{\widehat{k}^{(a)}}{N} \right\}}{\widehat{k}^{(a)}} \right]$$

$$\stackrel{(3)}{=} \sum_{a \in \mathcal{V}} \sum_{j \in \mathcal{H}_a^0} \frac{\alpha}{N} \mathbb{E} \left[\frac{\mathbb{1} \left\{ P_{a,j} \leq \alpha \frac{\widehat{k}^{(a)}}{N} \right\}}{\alpha \frac{\widehat{k}^{(a)}}{N}} \right]$$

$$\stackrel{(4)}{\leq} \sum_{a \in \mathcal{V}} \sum_{j \in \mathcal{H}_a^0} \frac{\alpha}{N}.1$$

$$= \sum_{a \in \mathcal{V}} \alpha \frac{|\mathcal{H}_a^0|}{N}$$

$$\stackrel{(5)}{=} \quad \alpha \frac{|\mathcal{H}^0|}{N}$$

[Corollary]

Under independence positive dependence, this item is no greater than 1. [ref 1,2]

$$(5): \sum_{a \in \mathcal{V}} |\mathcal{H}_a^0| = |\mathcal{H}^0|$$



QuTE---Multiple Rounds

- Run the QuTE algorithm c (>1) times
 - Each node/agent can query p-vectors from all the other nodes within distance c.
 - Equivalent to enlarging the communication range
- Performance guarantee

Theorem 2 Suppose that the p-values are independent, or positively dependent. Then the multi-step QuTE algorithm with c>1 rounds of communication guarantees that $FDR \leq \alpha \frac{|\mathcal{H}^0|}{N}$.

>> Proof is essentially the same as Theorem 1



Simulation

Generation of p-values

$$X \sim \mu + \mathcal{N}(0,1)$$

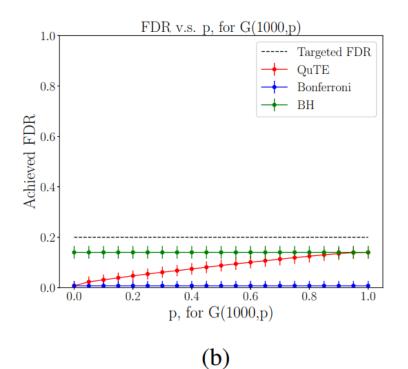
$$P = 1 - \Phi(X)$$

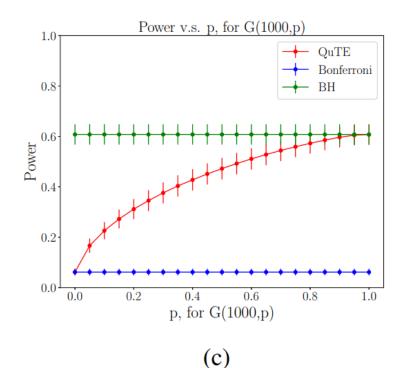
- $\mu=0$ for nulls and $\mu>0 (\mu=2)$ for alternatives
- Target FDR: $\alpha = 0.2$
- Null-proportion: $\pi_0 = \frac{|\mathcal{H}^0|}{N} = 0.7$
- Erdos-Renyi random graph G(N, p)
 - Any pair of vertices are connected w.p. p, in total N vertices
- Planar grid
 - Rectangle and square grid



Simulation

- Erdos-Renyi random graph G(N, p)
 - Any pair of vertices are connected w.p. p, in total N=1000 (a)
 nodes, each testing one hypothesis

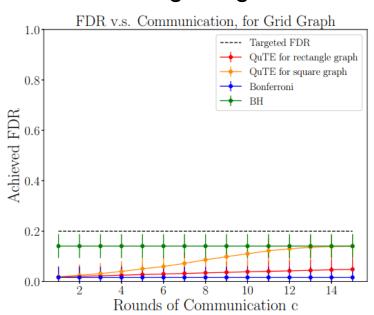


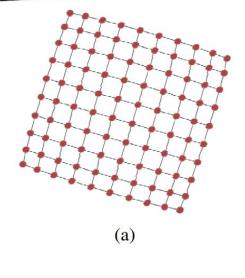


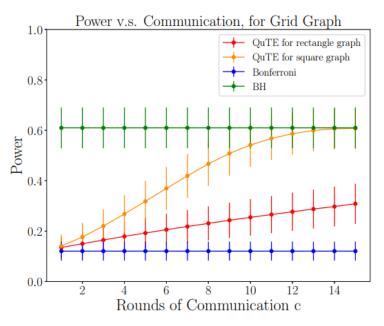


Simulation

- Planar grid
 - N=256 nodes, each testing one hypothesis
 - Square grid 16*16
 - Rectangular grid 2*128







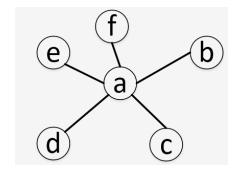
(b)

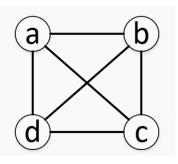
(c)



Discussion

- Quantization error of p-values
 - >> Sensors with limited computation ability--> accuracy
 - >> Wireless channels are noisy
 - >> What is the effect of quantization?
- Computation redundancy
 - >> As long as there exists one single node that are connected to all the other nodes, QuTE becomes centralized BH procedure
 - >> For a complete graph, all nodes perform the same BH multiple times--> unnecessary power consumption
 - >> Sampling







References

[1] Barber, Rina Foygel, and Aaditya Ramdas. "The p-filter: multilayer false discovery rate control for grouped hypotheses." *Journal of the Royal Statistical Society: Series B (Statistical Methodology)* 79.4 (2017): 1247-1268.

[2] Blanchard, Gilles, and Etienne Roquain. "Two simple sufficient conditions for FDR control." *Electronic journal of Statistics* 2 (2008): 963-992.



Thank you

• Questions?