

**ON BOUNDING THE UNION PROBABILITY USING PARTIAL
WEIGHTED INFORMATION – SUPPLEMENTARY MATERIAL**

JUN YANG, FADY ALAJAJI, AND GLEN TAKAHARA

1. Relation to the Cohen-Merhav bound. Let $f_i(B) > 0$ and $m_i(\omega_B)$ be non-negative real functions. Then by the Cauchy-Schwarz inequality,

$$\left[\sum_{B:i \in B} f_i(B) p_B \right] \left[\sum_{B:i \in B} \frac{p_B}{f_i(B)} m_i^2(\omega_B) \right] \geq \left[\sum_{B:i \in B} p_B m_i(\omega_B) \right]^2. \quad (1)$$

Thus, using

$$P \left(\bigcup_{i=1}^N A_i \right) = \sum_{B \in \mathcal{B}} \left(\sum_{i=1}^N f_i(B) \right) p_B = \sum_{i=1}^N \sum_{B \in \mathcal{B}: i \in B} f_i(B) p_B. \quad (2)$$

we have

$$P \left(\bigcup_{i=1}^N A_i \right) = \sum_{i=1}^N \sum_{B:i \in B} f_i(B) p_B \geq \sum_{i=1}^N \frac{[\sum_{B:i \in B} p_B m_i(\omega_B)]^2}{\sum_{B:i \in B} \frac{p_B}{f_i(B)} m_i^2(\omega_B)}. \quad (3)$$

If we define $f_i(B)$ by

$$f_i(B) = \begin{cases} \frac{1}{|B|} = \frac{1}{\deg(\omega_B)} & \text{if } i \in B \\ 0 & \text{if } i \notin B \end{cases} \quad (4)$$

so that

$$P \left(\bigcup_{i=1}^N A_i \right) = \sum_{i=1}^N \sum_{B \in \mathcal{B}: i \in B} \frac{p_B}{\deg(\omega_B)} = \sum_{i=1}^N \sum_{\omega \in A_i} \frac{p(\omega)}{\deg(\omega)}. \quad (5)$$

then the inequality reduces to

$$P \left(\bigcup_{i=1}^N A_i \right) \geq \sum_{i=1}^N \frac{[\sum_{B:i \in B} p_B m_i(\omega_B)]^2}{\sum_{B:i \in B} p_B m_i^2(\omega_B) |B|} = \sum_i \frac{[\sum_{\omega \in A_i} p(\omega) m_i(\omega)]^2}{\sum_j \sum_{\omega \in A_i \cap A_j} p(\omega) m_i^2(\omega)}, \quad (6)$$

where the equality holds when $m_i(\omega) = \frac{1}{\deg(\omega)}$ (i.e., $m_i(\omega_B) = \frac{1}{|B|}$), which was first shown by Cohen and Merhav [1, Theorem 2.1].

When $m_i(\omega) = c_i > 0$, (6) reduces to the DC bound

$$P \left(\bigcup_{i=1}^N A_i \right) \geq \sum_i \frac{[c_i P(A_i)]^2}{\sum_j c_j^2 P(A_i \cap A_j)} = \sum_i \frac{P(A_i)^2}{\sum_j P(A_i \cap A_j)} = \ell_{\text{DC}}. \quad (7)$$

Note that as remarked in [2], the DC bound can be seen as a special case of the lower bound

$$P \left(\bigcup_{i=1}^N A_i \right) \geq \frac{[\sum_i c_i P(A_i)]^2}{\sum_i \sum_j c_i^2 P(A_i \cap A_j)}, \quad (8)$$

when $c_i = \frac{P(A_i)}{\sum_j P(A_i \cap A_j)}$. This is because

$$\begin{aligned} \frac{\left[\sum_i \left(\frac{P(A_i)}{\sum_j P(A_i \cap A_j)} \right) P(A_i) \right]^2}{\sum_i \sum_j \left(\frac{P(A_i)}{\sum_j P(A_i \cap A_j)} \right)^2 P(A_i \cap A_j)} &= \frac{\left(\sum_i \frac{P(A_i)^2}{\sum_j P(A_i \cap A_j)} \right)^2}{\sum_i \left\{ \left(\frac{P(A_i)}{\sum_j P(A_i \cap A_j)} \right)^2 \sum_j P(A_i \cap A_j) \right\}} \quad (9) \\ &= \frac{\ell_{\text{DC}}^2}{\ell_{\text{DC}}} = \ell_{\text{DC}}. \end{aligned}$$

Note that although $c_i > 0$ is not assumed in (8), one can always replace c_i by $|c_i|$ in (8) if $c_i < 0$ to get a sharper bound.

However, the lower bound in (8) is looser than the following two (left-most) lower bounds (which we later derive in (16) and (18)):

$$\sum_{i=1}^N \frac{c_i^2 P(A_i)^2}{c_i \sum_k c_k P(A_i \cap A_k)} \geq \frac{[\sum_i c_i P(A_i)]^2}{\sum_i \sum_k c_i c_k P(A_i \cap A_k)} \geq \frac{[\sum_i c_i P(A_i)]^2}{\sum_i \sum_j c_i^2 P(A_i \cap A_j)}, \quad (10)$$

where $c_i > 0$ for all i and the last inequality can be proved using $2c_i c_j \leq c_i^2 + c_j^2$.

2. Relation to the Gallot-Kounias bound. By the Cauchy-Schwarz inequality, or assuming $m_i(\omega) = 1$ in (1), we have

$$\left[\sum_{B:i \in B} f_i(B) p_B \right] \left[\sum_{B:i \in B} \frac{p_B}{f_i(B)} \right] \geq \left[\sum_{B:i \in B} p_B \right]^2 = P(A_i)^2. \quad (11)$$

Using $f_i(B)$ defined using \mathbf{c} (note that $f_i(B) > 0$ is equivalent to $c_i > 0$ for all i), we have

$$\left[\sum_{B:i \in B} \frac{c_i p_B}{\sum_{k \in B} c_k} \right] \left[\sum_{B:i \in B} \left(\frac{\sum_{k \in B} c_k}{c_i} \right) p_B \right] \geq P(A_i)^2. \quad (12)$$

Note that

$$\sum_{B:i \in B} \left(\frac{\sum_{k \in B} c_k}{c_i} \right) p_B = \frac{1}{c_i} \sum_{k=1}^N \sum_{B:i \in B, k \in B} c_k p_B = \frac{\sum_k c_k P(A_i \cap A_k)}{c_i}. \quad (13)$$

Therefore, we have

$$\left[\sum_{B:i \in B} \frac{c_i p_B}{\sum_{k \in B} c_k} \right] \left[\frac{\sum_k c_k P(A_i \cap A_k)}{c_i} \right] \geq P(A_i)^2. \quad (14)$$

Then for all i ,

$$\sum_{B:i \in B} \frac{c_i p_B}{\sum_{k \in B} c_k} \geq \frac{c_i^2 P(A_i)^2}{c_i \sum_k c_k P(A_i \cap A_k)} \quad (15)$$

By summing (15) over i , we get another new lower bound:

$$P \left(\bigcup_i A_i \right) \geq \sum_{i=1}^N \frac{c_i^2 P(A_i)^2}{c_i \sum_k c_k P(A_i \cap A_k)}. \quad (16)$$

Note that we can use Cauchy-Schwarz inequality again:

$$\left[\sum_{i=1}^N \frac{c_i^2 P(A_i)^2}{c_i \sum_k c_k P(A_i \cap A_k)} \right] \left[\sum_i c_i \sum_k c_k P(A_i \cap A_k) \right] \geq \left[\sum_i c_i P(A_i) \right]^2, \quad (17)$$

which yields

$$P\left(\bigcup_i A_i\right) \geq \sum_{i=1}^N \frac{c_i^2 P(A_i)^2}{c_i \sum_k c_k P(A_i \cap A_k)} \geq \frac{[\sum_i c_i P(A_i)]^2}{\sum_i \sum_k c_i c_k P(A_i \cap A_k)}. \quad (18)$$

Since the above inequality holds for any positive \mathbf{c} , we have

$$P\left(\bigcup_i A_i\right) \geq \max_{\mathbf{c} \in \mathbb{R}_+^N} \sum_{i=1}^N \frac{c_i^2 P(A_i)^2}{c_i \sum_k c_k P(A_i \cap A_k)} \geq \max_{\mathbf{c} \in \mathbb{R}_+^N} \frac{[\sum_i c_i P(A_i)]^2}{\sum_i \sum_k c_i c_k P(A_i \cap A_k)}. \quad (19)$$

One can show that by computing the partial derivative with respect to c_i and set it to zero that

$$\max_{\mathbf{c} \in \mathbb{R}^N} \sum_{i=1}^N \frac{c_i^2 P(A_i)^2}{c_i \sum_k c_k P(A_i \cap A_k)} = \max_{\mathbf{c} \in \mathbb{R}^N} \frac{[\sum_i c_i P(A_i)]^2}{\sum_i \sum_k c_i c_k P(A_i \cap A_k)} =: \ell_{\text{GK}}, \quad (20)$$

where ℓ_{GK} is the Gallot-Kounias bound (see [2]), and the optimal $\tilde{\mathbf{c}}$ can be obtained from

$$\mathbf{\Sigma} \tilde{\mathbf{c}} = \boldsymbol{\alpha}, \quad (21)$$

where $\boldsymbol{\alpha} = (P(A_1), P(A_2), \dots, P(A_N))^T$ and $\mathbf{\Sigma}$ is a $N \times N$ matrix whose (i, j) -th element equals to $P(A_i \cap A_j)$. Thus, we conclude that the lower bounds in (19) are equal to the GK bound as shown in [2] if $\tilde{\mathbf{c}} \in \mathbb{R}_+^N$; otherwise, the lower bounds in (19) are weaker than the GK bound.

3. Complete Results and Proof of Theorem 1.

Theorem. For any given \mathbf{c} that satisfies

$$\sum_{k \in B} c_k \neq 0, \quad \text{for all } B \in \mathcal{B} \quad (22)$$

a new lower bound on the union probability is given by

$$P\left(\bigcup_{i=1}^N A_i\right) \geq \sum_{i=1}^N \ell_i(\mathbf{c}) =: \ell_{\text{NEW-I}}(\mathbf{c}), \quad (23)$$

where

$$\ell_i(\mathbf{c}) = P(A_i) \left(\frac{c_i}{\sum_{k \in B_1^{(i)}} c_k} + \frac{c_i}{\sum_{k \in B_2^{(i)}} c_k} - \frac{c_i \sum_k c_k P(A_i \cap A_k)}{P(A_i) \left(\sum_{k \in B_1^{(i)}} c_k \right) \left(\sum_{k \in B_2^{(i)}} c_k \right)} \right), \quad (24)$$

where $B_1^{(i)}$ and $B_2^{(i)}$ are subsets of $\{1, \dots, N\}$ that satisfy the following conditions.

1. If $\frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)} \geq 0$ and $\min_{\{B: i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} < 0$, then

$$\begin{aligned} B_1^{(i)} &= \arg \max_{\{B: i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} \quad \text{s.t.} \quad \frac{\sum_{k \in B} c_k}{c_i} < 0, \\ B_2^{(i)} &= \arg \max_{\{B: i \in B\}} \frac{\sum_{k \in B} c_k}{c_i}. \end{aligned} \quad (25)$$

2. If $\frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)} \geq 0$ and $\min_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} \geq 0$, then

$$\begin{aligned} B_1^{(i)} &= \arg \max_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} \text{ s.t. } \frac{\sum_{k \in B} c_k}{c_i} \leq \frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)}, \\ B_2^{(i)} &= \arg \min_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} \text{ s.t. } \frac{\sum_{k \in B} c_k}{c_i} \geq \frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)}. \end{aligned} \quad (26)$$

3. If $\frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)} < 0$ and $\frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)} < \left\{ \max_{\{B:i \in B, \frac{\sum_{k \in B} c_k}{c_i} < 0\}} \frac{\sum_{k \in B} c_k}{c_i} \right\}$, then

$$\begin{aligned} B_1^{(i)} &= \arg \max_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i}, \quad \text{s.t. } \frac{\sum_{k \in B} c_k}{c_i} < 0, \\ B_2^{(i)} &= \arg \min_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i}. \end{aligned} \quad (27)$$

4. If $\frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)} < 0$ and $\frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)} \geq \left\{ \max_{\{B:i \in B, \frac{\sum_{k \in B} c_k}{c_i} < 0\}} \frac{\sum_{k \in B} c_k}{c_i} \right\}$, then

$$\begin{aligned} B_1^{(i)} &= \arg \max_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i}, \\ B_2^{(i)} &= \arg \max_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} \text{ s.t. } \frac{\sum_{k \in B} c_k}{c_i} \leq \frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)}. \end{aligned} \quad (28)$$

Proof. Note that for the third and fourth cases, under the condition $\frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)} < 0$, the elements of \mathbf{c} cannot be all positive or negative, so the set $\{B : i \in B, \frac{\sum_{k \in B} c_k}{c_i} < 0\}$ is not empty. Therefore, the solutions of $B_1^{(i)}$ and $B_2^{(i)}$ always exist.

We note that $\ell_i(\mathbf{c})$ is the solution of

$$\begin{aligned} \min_{\{p_B:i \in B\}} \sum_{B:i \in B} \frac{c_i p_B}{\sum_{k \in B} c_k} \quad \text{s.t.} \quad \sum_{B:i \in B} p_B &= P(A_i), \\ \sum_{B:i \in B} \left(\frac{\sum_{k \in B} c_k}{c_i} \right) p_B &= \frac{1}{c_i} \sum_k c_k P(A_i \cap A_k), \\ p_B &\geq 0, \quad \text{for all } B \in \mathcal{B} \text{ such that } i \in B. \end{aligned} \quad (29)$$

From (29) we have that

$$\sum_{B:i \in B} \frac{c_i p_B}{\sum_{k \in B} c_k} \geq \ell_i(\mathbf{c}). \quad (30)$$

Summing (30) over i and using

$$P\left(\bigcup_{i=1}^N A_i\right) = \sum_{i=1}^N \sum_{B \in \mathcal{B}: i \in B} \frac{c_i p_B}{\sum_{k \in B} c_k} = \sum_{i=1}^N \sum_{\omega \in A_i} \frac{c_i p(\omega)}{\sum_{\{k:\omega \in A_k\}} c_k}. \quad (31)$$

we directly obtain $P\left(\bigcup_{i=1}^N A_i\right) \geq \sum_{i=1}^N \ell_i(\mathbf{c})$.

Note that we can solve (29) using the same technique used in [4, 5]. Consider two subsets B_1 and B_2 such that $p_{B_1} \geq 0$ and $p_{B_2} \geq 0$, then denoting

$$b := \frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)}, \quad b_1 := \frac{\sum_{k \in B_1} c_k}{c_i}, \quad b_2 := \frac{\sum_{k \in B_2} c_k}{c_i}, \quad (32)$$

then problem (29) reduces to

$$\begin{aligned} \ell_i(\mathbf{c}) = \min_{\{p_{B_1}, p_{B_2}\}} \quad & \frac{p_{B_1}}{b_1} + \frac{p_{B_2}}{b_2} \quad \text{s.t.} \quad p_{B_1} + p_{B_2} = P(A_i), \\ & b_1 p_{B_1} + b_2 p_{B_2} = bP(A_i), \\ & p_{B_1} \geq 0, \quad p_{B_2} \geq 0. \end{aligned} \quad (33)$$

According to [4, Appendix B], one can get that

$$\ell_i(\mathbf{c}) = \min_{\{b_1, b_2: b_1 \leq b \leq b_2\}} P(A_i) \left(\frac{1}{b_1} + \frac{1}{b_2} - \frac{b}{b_1 b_2} \right), \quad (34)$$

and the partial derivative of $P(A_i) \left(\frac{1}{b_1} + \frac{1}{b_2} - \frac{b}{b_1 b_2} \right)$ with respect to b_1 and b_2 are (see [4, Appendix B, Eq. (B.3)]):

$$\begin{aligned} \frac{\partial \left[P(A_i) \left(\frac{1}{b_1} + \frac{1}{b_2} - \frac{b}{b_1 b_2} \right) \right]}{\partial b_1} &= \frac{P(A_i)}{b_1^2} \left(\frac{b - b_2}{b_2} \right), \\ \frac{\partial \left[P(A_i) \left(\frac{1}{b_1} + \frac{1}{b_2} - \frac{b}{b_1 b_2} \right) \right]}{\partial b_2} &= \frac{P(A_i)}{b_2^2} \left(\frac{b - b_1}{b_1} \right). \end{aligned} \quad (35)$$

Note that the partial derivatives are not continuous at $b_1 = 0$ and $b_2 = 0$. Therefore, the solution depends on the following different scenarios.

1. If $b \geq 0$ and $\min_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} < 0$, the solutions of (34) are given by

$$\begin{aligned} b_1 &= \max_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} \quad \text{s.t.} \quad \frac{\sum_{k \in B} c_k}{c_i} < 0, \\ b_2 &= \max_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i}. \end{aligned} \quad (36)$$

2. If $b \geq 0$ and $\min_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} \geq 0$, the solutions of (34) are given by

$$\begin{aligned} b_1 &= \max_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} \quad \text{s.t.} \quad \frac{\sum_{k \in B} c_k}{c_i} \leq b, \\ b_2 &= \min_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} \quad \text{s.t.} \quad \frac{\sum_{k \in B} c_k}{c_i} \geq b. \end{aligned} \quad (37)$$

3. If $b < 0$ and $b < \left\{ \max_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i}, \quad \text{s.t.} \quad \frac{\sum_{k \in B} c_k}{c_i} < 0 \right\}$, the solutions of (34) are given by

$$\begin{aligned} b_1 &= \max_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i}, \quad \text{s.t.} \quad \frac{\sum_{k \in B} c_k}{c_i} < 0, \\ b_2 &= \min_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i}. \end{aligned} \quad (38)$$

4. If $b < 0$ and $b \geq \left\{ \max_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i}, \quad \text{s.t.} \quad \frac{\sum_{k \in B} c_k}{c_i} < 0 \right\}$, the solutions of (34) are given by

$$\begin{aligned} b_1 &= \max_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i}, \\ b_2 &= \max_{\{B:i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} \quad \text{s.t.} \quad \frac{\sum_{k \in B} c_k}{c_i} \leq b. \end{aligned} \quad (39)$$

□

4. Proof of Lemma 2.

Lemma. *When $\mathbf{c} \in \mathbb{R}_+^N$, the lower bound $\ell_{NEW-I}(\mathbf{c})$ can be computed in pseudo-polynomial time, and can be arbitrarily closely approximated by an algorithm running in polynomial time.*

Proof. The problems in (25) to (28) are exactly the 0/1 knapsack problem with mass equals to value (see [3], the corresponding decision problem is also called subset sum problem). Unfortunately, the 0/1 knapsack problem is NP-hard in general.

However, if $\mathbf{c} \in \mathbb{R}_+^N$, i.e., the case (26), there exists a dynamic programming solution which runs in pseudo-polynomial time, i.e., polynomial in N , but exponential in the number of bits required to represent $\frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)}$ (see [3]). Furthermore, there is a fully polynomial-time approximation scheme (FPTAS), which finds a solution that is correct within a factor of $(1 - \epsilon)$ of the optimal solution (see [3]). The running time is bounded by a polynomial and $1/\epsilon$ where ϵ is a bound on the correctness of the solution.

Therefore, if $\mathbf{c} \in \mathbb{R}_+^N$, one can get a lower bound for $\ell_i(\mathbf{c})$ in polynomial time which can be arbitrarily close to $\ell_i(\mathbf{c})$ by setting ϵ small enough, i.e.,

$$\ell_i(\mathbf{c}) \geq \ell_i^L(\mathbf{c}, \epsilon), \quad \lim_{\epsilon \rightarrow 0^+} \ell_i^L(\mathbf{c}, \epsilon) = \ell_i(\mathbf{c}). \quad (40)$$

The details are as follows. First, assume \hat{B}_1 and \hat{B}_2 are obtained by the FPTAS which satisfy

$$(1 - \epsilon) \sum_{k \in B_1^{(i)}} c_k \leq \sum_{k \in \hat{B}_1^{(i)}} c_k \leq \sum_{k \in B_1^{(i)}} c_k, \quad \sum_{k \in B_2^{(i)}} c_k \leq \sum_{k \in \hat{B}_2^{(i)}} c_k \leq (1 + \epsilon) \sum_{k \in B_2^{(i)}} c_k. \quad (41)$$

Then we have

$$\begin{aligned} \sum_{k \in B_1^{(i)}} c_k &\leq \min \left\{ \frac{\sum_{k \in \hat{B}_1^{(i)}} c_k}{1 - \epsilon}, \frac{\sum_k c_k P(A_i \cap A_k)}{P(A_i)} \right\} =: b_1^{(i)}, \\ \sum_{k \in B_2^{(i)}} c_k &\geq \max \left\{ \frac{\sum_{k \in \hat{B}_2^{(i)}} c_k}{1 + \epsilon}, \frac{\sum_k c_k P(A_i \cap A_k)}{P(A_i)} \right\} =: b_2^{(i)}. \end{aligned} \quad (42)$$

Then one can get the arbitrarily close lower bound for $\ell_i(\mathbf{c})$ as

$$\ell_i(\mathbf{c}) \geq \ell_i^L(\mathbf{c}, \epsilon) := P(A_i) \left(\frac{c_i}{b_1^{(i)}} + \frac{c_i}{b_2^{(i)}} - \frac{c_i \sum_k c_k P(A_i \cap A_k)}{P(A_i) b_1^{(i)} b_2^{(i)}} \right). \quad (43)$$

Therefore, we can get a lower bound for $P\left(\bigcup_{i=1}^N A_i\right)$ that is arbitrarily close to $\ell_{NEW-I}(\mathbf{c})$ in polynomial time: $P\left(\bigcup_{i=1}^N A_i\right) \geq \sum_i \ell_i(\mathbf{c}) \geq \sum_i \ell_i^L(\mathbf{c}, \epsilon)$. \square

5. Proof of Corollary 1.

Corollary. *(New class of upper bounds $\hat{h}_{NEW-I}(\mathbf{c})$): We can derive an upper bound for any given $\mathbf{c} \in \mathbb{R}_+^N$ by*

$$\begin{aligned} P\left(\bigcup_i A_i\right) &\leq \left(\frac{1}{\min_k c_k} + \frac{1}{\sum_k c_k} \right) \sum_i c_i P(A_i) \\ &\quad - \frac{1}{(\min_k c_k) \sum_k c_k} \sum_i \sum_k c_i c_k P(A_i \cap A_k) =: \hat{h}_{NEW-I}(\mathbf{c}). \end{aligned} \quad (44)$$

Proof. We get the upper bound by maximizing, instead of minimizing, the objective function of (29). More specifically, for any given $\mathbf{c} \in \mathbb{R}^+$, a upper bound can be obtained by

$$h(\mathbf{c}) = \sum_{i=1}^N h_i(\mathbf{c}), \quad (45)$$

where $h_i(\mathbf{c})$ is defined by

$$\begin{aligned} h_i(\mathbf{c}) := \max_{\{p_B: i \in B\}} \sum_{B: i \in B} \frac{c_i p_B}{\sum_{k \in B} c_k} \quad \text{s.t.} \quad \sum_{B: i \in B} p_B = P(A_i), \\ \sum_{B: i \in B} \left(\frac{\sum_{k \in B} c_k}{c_i} \right) p_B = \frac{1}{c_i} \sum_k c_k P(A_i \cap A_k), \\ p_B \geq 0, \quad \text{for all } B \in \mathcal{B} \text{ such that } i \in B. \end{aligned} \quad (46)$$

The resulting upper bound is given by

$$\begin{aligned} P\left(\bigcup_i A_i\right) &\leq \sum_i \left\{ P(A_i) \left[\frac{c_i}{\min_k c_k} + \frac{c_i}{\sum_k c_k} \right. \right. \\ &\quad \left. \left. - \frac{c_i^2}{(\min_k c_k) \sum_k c_k} \frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)} \right] \right\} \\ &= \left(\frac{1}{\min_k c_k} + \frac{1}{\sum_k c_k} \right) \sum_i c_i P(A_i) \\ &\quad - \frac{1}{(\min_k c_k) \sum_k c_k} \sum_i \sum_k c_i c_k P(A_i \cap A_k). \end{aligned} \quad (47)$$

□

6. Proof of Theorem 2.

Theorem. Defining $\mathcal{B}^- = \mathcal{B} \setminus \{1, \dots, N\}$, $\tilde{\gamma}_i := \sum_k c_k P(A_i \cap A_k)$, $\tilde{\alpha}_i := P(A_i)$ and

$$\tilde{\delta} := \max_i \left[\frac{\tilde{\gamma}_i - (\sum_k c_k - \min_k c_k) \tilde{\alpha}_i}{\min_k c_k} \right]^+, \quad (48)$$

where $\mathbf{c} \in \mathbb{R}_+^N$, another class of lower bounds is given by

$$P\left(\bigcup_{i=1}^N A_i\right) \geq \tilde{\delta} + \sum_{i=1}^N \ell'_i(\mathbf{c}, \tilde{\delta}) =: \ell_{NEW-II}(\mathbf{c}), \quad (49)$$

where

$$\begin{aligned} \ell'_i(\mathbf{c}, x) &= [P(A_i) - x] \cdot \\ &\quad \left(\frac{c_i}{\sum_{k \in B_1^{(i)}} c_k} + \frac{c_i}{\sum_{k \in B_2^{(i)}} c_k} - \frac{c_i \sum_k c_k [P(A_i \cap A_k) - x]}{[P(A_i) - x] \left(\sum_{k \in B_1^{(i)}} c_k \right) \left(\sum_{k \in B_2^{(i)}} c_k \right)} \right), \end{aligned} \quad (50)$$

and

$$\begin{aligned} B_1^{(i)} &= \arg \max_{\{B \in \mathcal{B}^- : i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} \text{ s.t. } \frac{\sum_{k \in B} c_k}{c_i} \leq \frac{\sum_k c_k [P(A_i \cap A_k) - x]}{c_i [P(A_i) - x]}, \\ B_2^{(i)} &= \arg \min_{\{B \in \mathcal{B}^- : i \in B\}} \frac{\sum_{k \in B} c_k}{c_i} \text{ s.t. } \frac{\sum_{k \in B} c_k}{c_i} \geq \frac{\sum_k c_k [P(A_i \cap A_k) - x]}{c_i [P(A_i) - x]}. \end{aligned} \quad (51)$$

Proof. Let $x = p_{\{1,2,\dots,N\}}$ and consider $\sum_i \ell'_i(\mathbf{c}, x) + x$ as a new lower bound where where $\ell'_i(\mathbf{c}, x)$ equals to the objective value of the problem

$$\begin{aligned} & \min_{\{p_B : i \in B, B \in \mathcal{B}^-\}} \sum_{B : i \in B, B \in \mathcal{B}^-} \frac{c_i p_B}{\sum_{k \in B} c_k} \\ \text{s.t. } & \sum_{B : i \in B, B \in \mathcal{B}^-} p_B = P(A_i) - x, \\ & \sum_{B : i \in B, B \in \mathcal{B}^-} \left(\frac{\sum_{k \in B} c_k}{c_i} \right) p_B = \frac{1}{c_i} \sum_k c_k [P(A_i \cap A_k) - x], \\ & p_B \geq 0, \quad \text{for all } B \in \mathcal{B}^- \text{ such that } i \in B. \end{aligned} \quad (52)$$

The solution of (52) exists if and only if $\min_k c_k \leq \frac{\tilde{\gamma}_i - (\sum_k c_k)x}{\tilde{\alpha}_i - x} \leq \sum_k c_k - \min_k c_k$, which gives $\max_i \left[\frac{\tilde{\gamma}_i - (\sum_k c_k - \min_k c_k)\tilde{\alpha}_i}{\min_k c_k} \right]^+ \leq x \leq \min_i \left[\frac{\tilde{\gamma}_i - (\min_k c_k)\tilde{\alpha}_i}{\sum_k c_k - \min_k c_k} \right]$. Therefore, the new lower bound can be written as

$$\min_x \left[x + \sum_{i=1}^N \ell'_i(\mathbf{c}, x) \right] \text{ s.t. } \left[\frac{\tilde{\gamma}_i - (\sum_k c_k - \min_k c_k)\tilde{\alpha}_i}{\min_k c_k} \right]^+ \leq x \leq \frac{\tilde{\gamma}_i - (\min_k c_k)\tilde{\alpha}_i}{\sum_k c_k - \min_k c_k}, \forall i. \quad (53)$$

Next, we can prove that the objective function of (53) is non-decreasing with x . First, we prove $\ell'_i(\mathbf{c}, x)$ is continuous when $\exists B' \in \mathcal{B}^-$ such that $\frac{\sum_k c_k [P(A_i \cap A_k) - x]}{c_i [P(A_i) - x]} = \frac{\sum_{k \in B'} c_k}{c_i}$. This can be proved by $\lim_{h \rightarrow 0^+} \ell'_i(\mathbf{c}, x + h) = \lim_{h \rightarrow 0^+} \ell'_i(\mathbf{c}, x - h) = \frac{c_i}{\sum_{k \in B'} c_k}$. Then one can prove that when $\frac{\sum_{k \in B_2^{(i)}} c_k}{c_i} < \frac{\sum_k c_k [P(A_i \cap A_k) - x]}{c_i [P(A_i) - x]} < \frac{\sum_{k \in B_1^{(i)}} c_k}{c_i}$, the partial derivative of $\ell'_i(\mathbf{c}, x) + \frac{c_i}{\sum_k c_k} x$ w.r.t. x is non-negative:

$$\begin{aligned} \frac{\partial \left(\ell'_i(\mathbf{c}, x) + \frac{c_i}{\sum_k c_k} x \right)}{\partial x} &= \frac{c_i}{\sum_k c_k} - \frac{c_i}{\sum_{k \in B_1^{(i)}} c_k} - \frac{c_i}{\sum_{k \in B_2^{(i)}} c_k} \\ &\quad + \frac{c_i \sum_k c_k}{\left(\sum_{k \in B_1^{(i)}} c_k \right) \left(\sum_{k \in B_2^{(i)}} c_k \right)} \\ &= \frac{c_i \left(\sum_k c_k - \sum_{k \in B_1^{(i)}} c_k \right) \left(\sum_k c_k - \sum_{k \in B_2^{(i)}} c_k \right)}{\left(\sum_k c_k \right) \left(\sum_{k \in B_1^{(i)}} c_k \right) \left(\sum_{k \in B_2^{(i)}} c_k \right)} \\ &= \frac{c_i \left(\sum_{k \notin B_1^{(i)}} c_k \right) \left(\sum_{k \notin B_2^{(i)}} c_k \right)}{\left(\sum_k c_k \right) \left(\sum_{k \in B_1^{(i)}} c_k \right) \left(\sum_{k \in B_2^{(i)}} c_k \right)} \geq 0. \end{aligned} \quad (54)$$

Therefore, the objective function of (53), $\sum_i \ell'_i(\mathbf{c}, x) + x = \sum_i \left(\ell'_i(\mathbf{c}, x) + \frac{c_i}{\sum_k c_k} x \right)$, is non-decreasing with x .

Finally, defining $\tilde{\delta}$ as in (48), the new lower bound can be written as $P\left(\bigcup_{i=1}^N A_i\right) \geq \tilde{\delta} + \sum_{i=1}^N \ell'_i(\mathbf{c}, \tilde{\delta})$, where $\ell'_i(\mathbf{c}, \tilde{\delta})$ can be obtained using the solution for $\ell_i(\mathbf{c})$ with $b = \frac{\sum_k c_k P(A_i \cap A_k)}{c_i P(A_i)}$ replaced by $\tilde{b} = \frac{\sum_k c_k [P(A_i \cap A_k) - \tilde{\delta}]}{c_i [P(A_i) - \tilde{\delta}]}$. \square

7. Proof of Corollary 2.

Corollary. (Improved class of upper bounds $\hbar_{NEW-II}(\mathbf{c})$): We can improve the upper bound $\hbar_{NEW-I}(\mathbf{c})$ in (44) by

$$\begin{aligned} P\left(\bigcup_i A_i\right) &\leq \min_i \left\{ \frac{\sum_k c_k P(A_i \cap A_k) - (\min_k c_k) P(A_i)}{\sum_k c_k - \min_k c_k} \right\} \\ &\quad + \left(\frac{1}{\min_k c_k} + \frac{1}{\sum_k c_k - \min_k c_k} \right) \sum_i c_i P(A_i) \\ &\quad - \frac{1}{(\min_k c_k)(\sum_k c_k - \min_k c_k)} \sum_i \sum_k c_i c_k P(A_i \cap A_k), \\ &=: \hbar_{NEW-II}(\mathbf{c}). \end{aligned} \tag{55}$$

Note that the upper bound $\hbar_{NEW-II}(\mathbf{c})$ in (55) is always sharper than \hbar_{NEW-I} in (44).

Proof. Letting $x = p_{\{1, \dots, N\}}$. Defining $\mathcal{B}^- = \mathcal{B} \setminus \{1, \dots, N\}$, then

$$\hbar'(\mathbf{c}) = \max_x \left[x + \sum_{i=1}^N \hbar'_i(\mathbf{c}, x) \right], \tag{56}$$

where $\hbar'_i(\mathbf{c}, x)$ is defined by

$$\begin{aligned} \hbar'_i(\mathbf{c}, x) &:= \max_{\{p_B: i \in B, B \in \mathcal{B}^-\}} \sum_{B: i \in B, B \in \mathcal{B}^-} \frac{c_i p_B}{\sum_{k \in B} c_k} \\ \text{s.t.} \quad &\sum_{B: i \in B, B \in \mathcal{B}^-} p_B = P(A_i) - x, \\ &\sum_{B: i \in B, B \in \mathcal{B}^-} \left(\frac{\sum_{k \in B} c_k}{c_i} \right) p_B = \frac{1}{c_i} \sum_k c_k [P(A_i \cap A_k) - x], \\ &p_B \geq 0, \quad \text{for all } B \in \mathcal{B}^- \text{ such that } i \in B. \end{aligned} \tag{57}$$

The solution of $\hbar'_i(\mathbf{c}, x)$ is independent with x :

$$\begin{aligned} \hbar'_i(\mathbf{c}, x) &= (P(A_i) - x) \left(\frac{c_i}{\min_k c_k} + \frac{c_i}{\sum_k c_k - \min_k c_k} \right) \\ &\quad - \frac{c_i}{(\min_k c_k)(\sum_k c_k - \min_k c_k)} \sum_k c_k (P(A_i \cap A_k) - x), \\ &= P(A_i) \left(\frac{c_i}{\min_k c_k} + \frac{c_i}{\sum_k c_k - \min_k c_k} \right) \\ &\quad - \frac{c_i}{(\min_k c_k)(\sum_k c_k - \min_k c_k)} \sum_k c_k P(A_i \cap A_k), \end{aligned} \tag{58}$$

and the solution exists if and only if for all i

$$\min_k c_k \leq \frac{\sum_k c_k P(A_i \cap A_k) - (\sum_k c_k)x}{P(A_i) - x} \leq \sum_k c_k - \min_k c_k. \quad (59)$$

Thus, we get

$$\begin{aligned} & \left\{ \max_i \frac{\sum_k c_k P(A_i \cap A_k) - (\sum_k c_k - \min_k c_k) P(A_i)}{\min_k c_k} \right\}^+ \\ & \leq x \leq \min_i \frac{\sum_k c_k P(A_i \cap A_k) - (\min_k c_k) P(A_i)}{\sum_k c_k - \min_k c_k} \end{aligned} \quad (60)$$

Therefore, we get the upper bound

$$\begin{aligned} P\left(\bigcup_i A_i\right) & \leq \min_i \left\{ \frac{\sum_k c_k P(A_i \cap A_k) - (\min_k c_k) P(A_i)}{\sum_k c_k - \min_k c_k} \right\} \\ & + \left(\frac{1}{\min_k c_k} + \frac{1}{\sum_k c_k - \min_k c_k} \right) \sum_i c_i P(A_i) \\ & - \frac{1}{(\min_k c_k)(\sum_k c_k - \min_k c_k)} \sum_i \sum_k c_i c_k P(A_i \cap A_k). \end{aligned} \quad (61)$$

□

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E-mail address: jun@utstat.toronto.edu

E-mail address: fady@mast.queensu.ca

E-mail address: takahara@mast.queensu.ca