

Joint Scheduling and Multiflow Maximization in Wireless Networks

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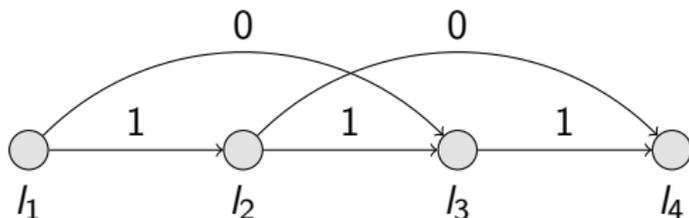
Maximum Multiflow Problem: Background

- We consider the **maximum multiflow** problem.
- To maximize the total (or concurrent) flow values between multiple sources and sinks, we usually have two steps:
 - ① Model the wireless interference as a conflict graph to compute the scheduling rates (Jain et al., 2005);
 - ② Find the maximum throughput supported by achievable scheduling rate.
- However, the first step is NP-hard and also difficult to approximate (Baker, 1994).
- The MMF problem is NP-hard even in simple settings (Wan, 2009).
- Existing joint optimization methods usually focus on either *approximate* solutions or *restricted* networks.

Maximum Multiflow Problem: Our Contribution

- We investigate the exact-optimal (rather than approximate) solutions for general multi-hop networks.
- We *jointly* calculate the maximum multiflow and the scheduling rates by employing a *decomposition* method.
- Usually only a small subset of the scheduling rate region is needed.
- We prove the output solution is optimal.
- We consider the most general setting: *multiple multicast* with network coding (Ahlsvede et al., 2000) allowed.

- The network is modeled by $\mathcal{N} = (\mathcal{V}, \mathcal{L}, \mathcal{I}, D)$, where
 - \mathcal{V} is the *node set*;
 - $\mathcal{L} \subseteq \mathcal{V}^2$ is the *link set*; each link l is associated with a *collision set* $\mathcal{I}(l)$;
 - $D : \mathcal{L}^2 \rightarrow \mathbb{Z}$ is a link-wise delay matrix.
 - We assume each link has a unit bandwidth and allow parallel links between nodes.
- $(\mathcal{L}, \mathcal{I}, D)$ can form a weighted, directed graph \mathcal{N} where (l, l') is an edge if $l' \in \mathcal{I}(l)$, and $D(l, l')$ is the weight.
- For example, for a 4-link line network with unit delay and 1-hop interference model, we can draw:



Scheduling Rate Region

- Assuming time is slotted, when link l is active at time slot t and link $l' \in \mathcal{I}(l)$ is active at time slot $t + D(l, l')$, a *collision occurs*.
- We use a binary matrix $S : \mathcal{L} \times \mathbb{N} \rightarrow \{0, 1\}$ to specify a *schedule*: $S(l, t) = 1$ indicates that link l is active in time slot t .
- $S(l, t)$ has collision if $S(l, t) = S(l', t + D(l, l')) = 1$ for some $l' \in \mathcal{I}(l)$.
- S is collision free if $S(l, t)$ is collision free for all (l, t) .
- For a collision-free schedule S and a link l , the link rate is

$$R_S(l) = \lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=0}^{T-1} S(l, t). \quad (1)$$

- $R_S = (R_S(l), l \in \mathcal{L})$ is the *rate vector* of S , and it is *achievable* if there is a collision-free schedule achieves S it.
- The collection $\mathcal{R}(\mathcal{N})$ of all the achievable rate vectors is called the *scheduling rate region*.

- We now define the maximum multiflow (MMF) problem.
- We say $F = (F(l) \in \mathbb{N}_{\geq 0} : l \in \mathcal{L})$ is a valid flow from source $s \in \mathcal{V}$ to sink $t \in \mathcal{V}$ w.r.t. R if
 - $0 \leq F(l) \leq R(l)$ for all $l \in \mathcal{L}$;
 - $\sum_{l \in \text{In}(v)} F(l) = \sum_{l \in \text{Out}(v)} F(l)$ for all $v \in \mathcal{V} \setminus \{s, t\}$.
- Flow $\sum_{l \in \text{Out}(s)} F(l)$ out of s equals to $\sum_{l \in \text{In}(t)} F(l)$ into t , and it is called the *value* of F , denoted by $\text{val}(F)$.
- **MMF problem:** Consider a source s_i multicasts a message to nodes in the set $\mathcal{D}_{s_i} = \{t_{i,1}, \dots, t_{i,k_i}\}$ (with network coding) at a rate

$$v_i = \min_j \text{val}(F_{i,j}),$$

where $F_{i,j}$ is a flow from s_i to $t_{i,j}$ for each j .

- Fix rate vector R . Link l has to accommodate all k flow requirements simultaneously, i.e., $\sum_{i=1}^k \max_j F_{i,j}(l) \leq R(l)$ for all $l \in \mathcal{L}$.

Problem Setting

- LP-MMF($\tilde{\mathcal{R}}$): maximize the sum of rates of multicasting k sources:

$$\text{maximize } \sum_{i=1}^k v_i \quad (2)$$

subject to $F_{i,j}$ is a valid flow, $\forall i \in [k], j \in [k_i]$,

$$\sum_{l \in \text{Out}(s_i)} F_{i,j}(l) = \sum_{l \in \text{In}(t_{i,j})} F_{i,j}(l) = v_i, \forall i \in [k], j \in [k_i],$$

$$G_i(l) \geq F_{i,j}(l), \forall l \in \mathcal{L}, i \in [k], j \in [k_i],$$

$$\sum_{i=1}^k G_i(l) \leq R(l), \forall l \in \mathcal{L}, \quad (3)$$

$$R \in \tilde{\mathcal{R}},$$

where $[k] := \{1, \dots, k\}$, $k, k_i \in \mathbb{N}_+, \forall i$.

- It takes polytope $\tilde{\mathcal{R}}$ as an input.
- (3) gives a dual vector μ ; $\mu(l)$ indicates how sensitive the optimization objective is to the rate constraint $R(l)$.

Algorithm

- Our joint algorithm computes the optimal MMF while only requires a subset of the scheduling rate region.
- We attach $a_i \geq 0$ to link l_i , and with $\mathbf{a} = (a_i)_{i=1, \dots, |\mathcal{L}|}$ we solve

$$\operatorname{argmax}_{R \in \mathcal{R}} \langle \mathbf{a}, R \rangle, \quad (4)$$

- This corresponds to a weighted maximal independent set problem

$$\begin{aligned} \text{ILP:} \quad & \text{maximize} \quad \sum_{i=1}^{|\mathcal{L}|} a_i S(l_i) \\ & \text{subject to} \quad S(l_i) + S(l_j) \leq 1, \quad \forall l_i, l_j : l_j \in \mathcal{I}(l_i). \end{aligned}$$

- The solution gives us an achievable rate vector, i.e., a vertex of \mathcal{R} .
- The ILP can have high complexity; we deal with this later.
- Based on (4), we iteratively search the MMF and the scheduling rates.

Algorithm 1

Input: a network $(\mathcal{V}, \mathcal{L}, \mathcal{I})$ **Output:** maximum multiflow v

- 1: Start with any rate vector $R_1 \in \mathcal{R}$, $\mathcal{R}_1 \leftarrow \{R_1\}$
 - 2: **for** $i = 1, 2, \dots$ **do**
 - 3: Run $v_i \leftarrow \text{LP-MMF}(\mathcal{R}_i)$ and obtain the dual vector μ_i
 - 4: Run ILP to find $R_{i+1} \leftarrow \text{argmax}_{R \in \mathcal{R}} \langle \mu_i, R \rangle$
 - 5: $\mathcal{R}_{i+1} \leftarrow \text{conv}(\mathcal{R}_i \cup \{R_{i+1}\})$
 - 6: **if** $\langle \mu_i, R_{i+1} \rangle = \max_{R \in \mathcal{R}_i} \langle \mu_i, R \rangle$ **then**
 - 7: **return** v_i
 - 8: **end if**
 - 9: **end for**
-

From $i = 1$, in each iteration, the algorithm works as follows:

- 1 We start with a known subset of the scheduling region \mathcal{R}_i . E.g., we start with $\mathcal{R}_1 = \{R_1\}$.
- 2 By the ILP, we find a new rate vector R_{i+1} by

$$R_{i+1} = \operatorname{argmax}_{R \in \mathcal{R}} \langle \mu_i, R \rangle. \quad (5)$$

- 3 We update the subset by $\mathcal{R}_{i+1} = \operatorname{conv}(\mathcal{R}_i \cup \{R_{i+1}\})$.
- 4 If $\langle \mu_i, R_{i+1} \rangle = \max_{R \in \mathcal{R}_i} \langle \mu_i, R \rangle$, the algorithm terminates and outputs the last optimal value of $\operatorname{LP-MMF}(\mathcal{R}_i)$.

The algorithm can terminate quite quickly; the remaining question is whether the output value is optimal.

Algorithm: Proof Sketch of Optimality

- 1 Since the scheduling region is a polytope with finite number of vertices, the Algorithm 1 will eventually terminate.
- 2 $\text{flow}(R)$ is the function that outputs the MMF w.r.t. a rate vector R .
- 3 \mathcal{R} is the entire scheduling region and \mathcal{R}_i is the subset until iteration i .
- 4 If the algorithm ends at iteration i' , $\langle \mu_{i'}, R_{i'+1} \rangle = \max_{R \in \mathcal{R}_{i'}} \langle \mu_{i'}, R \rangle$.
- 5 Combining with $R_{i+1} = \operatorname{argmax}_{R \in \mathcal{R}} \langle \mu_i, R \rangle$, we have

$$\max_{R \in \mathcal{R}} \langle \mu_{i'}, R \rangle = \max_{R \in \mathcal{R}_{i'}} \langle \mu_{i'}, R \rangle. \quad (6)$$

- 6 Suppose the optimal R in $\text{LP-MMF}(\mathcal{R}_{i'})$ is R^* .
- 7 We can verify R^* maximizes $\text{flow}(R) - \langle \mu_{i'}, R \rangle$ for $R \in \mathbb{R}_{\geq 0}^{|\mathcal{L}|}$, hence

$$\begin{aligned} \text{flow}(R^*) - \langle \mu_{i'}, R^* \rangle &\geq \text{flow}(R') - \langle \mu_{i'}, R' \rangle \\ &\stackrel{(a)}{\geq} \text{flow}(R') - \langle \mu_{i'}, R^* \rangle, \end{aligned}$$

where (a) is by (6) and the R^* maximizes $\langle \mu_{i'}, R \rangle$ for $R \in \mathcal{R}_{i'}$.

- 8 Therefore R^* maximizes $\text{flow}(R)$ for $R \in \mathcal{R}$.

Algorithm

- The algorithm requires integer linear programming, hence it does not have a polynomial time complexity in theory, which is expected since this problem is NP-hard.
- We use simulation on random networks to show the efficiency.

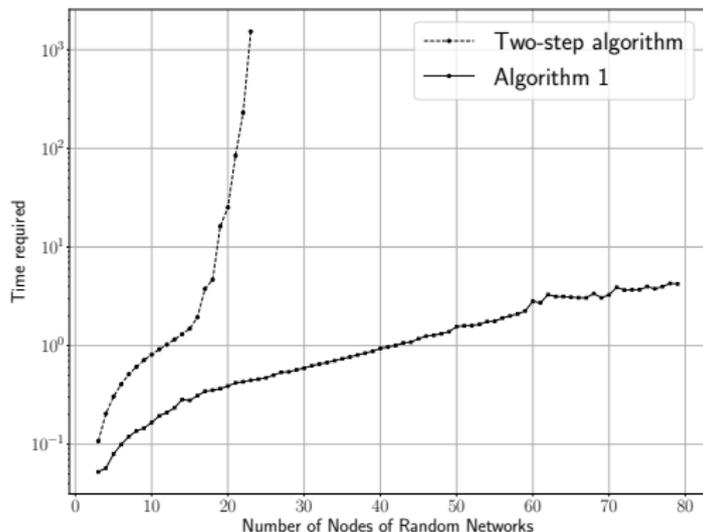
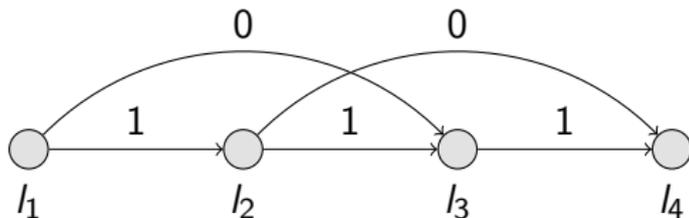


Figure: Results of 1-hop interference and multicast case of 1 source and 2 sinks.

Networks with Non-negligible Delays

- Recent results show that by utilizing non-negligible propagation delays, the scheduling region can be greatly improved (Yang et al., 2023).
- However, the scheduling problem suffers doubly-exponential complexity.
- We adopt a similar graphical formulation and extend our algorithm.
- We first review the scheduling problem.
- Recall for $|\mathcal{L}| = 4$ and unit delay and 1-hop interference, we have:



Networks with Non-negligible Delays

- For a collision-free schedule matrix S and integers $T \in \mathbb{N}_+$, $k \in \mathbb{Z}$, $S[T, k]$ denotes the submatrix of S consisting of columns $kT, kT + 1, \dots, (k + 1)T - 1$.

Definition (Scheduling Graph (Ma et al., 2021))

Given a network \mathcal{N} and an integer $T > 0$, a *scheduling graph* $(\mathcal{M}_T, \mathcal{E}_T)$ satisfies: the vertex set \mathcal{M}_T includes all the $|\mathcal{L}| \times T$ binary matrices A such that $A = S[T, 0]$ for a certain collision-free schedule S . The edge set \mathcal{E}_T includes all the vertex pairs (A, B) such that $A = S[T, 0]$ and $B = S[T, 1]$ for a certain collision-free schedule S .

- By choosing $T \geq \max_{l \in \mathcal{L}} \max_{l' \in \mathcal{I}(l)} |D(l, l')|$, the scheduling problem is searching all the cycles in the scheduling graph, which is NP-hard.

Networks with Non-negligible Delays

- To solve the MMF problem, the key is to have a function similar to $\operatorname{argmax}_{R \in \mathcal{R}} \langle \mathbf{a}, R \rangle$ that gives a rate vector fast.
- We hence define a *weighted* scheduling graph:

Definition

Given a weight vector $\mathbf{a} \in \mathbb{R}^{|\mathcal{L}|}$ and a scheduling graph $(\mathcal{M}_T, \mathcal{E}_T)$, a *weighted scheduling graph* $(\mathcal{M}_T, \mathcal{E}_T, w_{\mathbf{a}})$ is a directed, weighted graph such that: the vertex set is \mathcal{M}_T , and for each directed edge (v_1, v_2) in $(\mathcal{M}_T, \mathcal{E}_T)$, there is a weighed, directed edge (v_1, v_2) in $(\mathcal{M}_T, \mathcal{E}_T, w_{\mathbf{a}})$ with weight $w_{\mathbf{a}}(v_1, v_2) = \mathbf{a}^T v_2 \mathbf{1}$.

- Given a vector \mathbf{a} , finding a vector that solves $\operatorname{argmax}_{R \in \mathcal{R}} \langle \mathbf{a}, R \rangle$ is equivalent to finding the **maximum-mean-cycle** in $(\mathcal{M}_T, \mathcal{E}_T, w_{\mathbf{a}})$.
- This can be solved with a polynomial complexity $\Theta(|\mathcal{M}_T| |\mathcal{E}_T|)$ by the Karp's algorithm (Karp, 1978).

Algorithm 2 Algorithm for Networks with Delays

Input: a network $\mathcal{N} = (\mathcal{V}, \mathcal{L}, \mathcal{I}, D)$

Output: maximum multiflow v

- 1: Start with any rate vector $R_1 \in \mathcal{R}$, $\mathcal{R}_1 \leftarrow \{R_1\}$
 - 2: Construct $(\mathcal{M}_T, \mathcal{E}_T)$ from \mathcal{N}
 - 3: **for** $i = 1, 2, \dots$ **do**
 - 4: Run $v_i \leftarrow \text{LP-MMF}(\mathcal{R}_i)$ and obtain the dual vector μ_i
 - 5: Construct $(\mathcal{M}_T, \mathcal{E}_T, w_{\mu_i})$ by $(\mathcal{M}_T, \mathcal{E}_T)$ and μ_i
 - 6: Find maximum-mean-cycle in $(\mathcal{M}_T, \mathcal{E}_T, w_{\mu_i})$ and obtain $R_{i+1} \leftarrow \operatorname{argmax}_{R \in \mathcal{R}} \langle \mu_i, R \rangle$
 - 7: $\mathcal{R}_{i+1} \leftarrow \operatorname{conv}(\mathcal{R}_i \cup \{R_{i+1}\})$
 - 8: **if** $\langle \mu_i, R_{i+1} \rangle = \max_{R \in \mathcal{R}_i} \langle \mu_i, R \rangle$ **then**
 - 9: **return** v_i
 - 10: **end if**
 - 11: **end for**
-

Networks with Non-negligible Delays

- We can similar prove the optimality of the output.
- We present simulation results on simple settings:

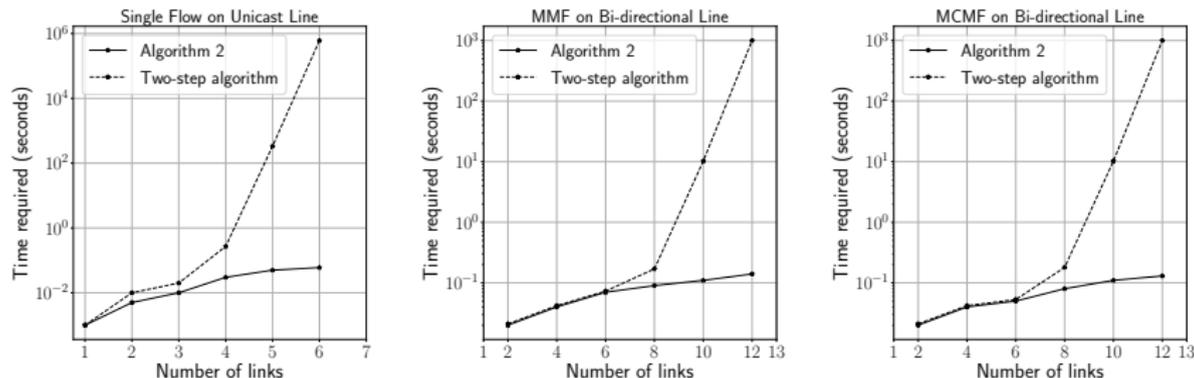


Figure: Line networks with unit delay

- For $\mathcal{N}_{4,1}^{D=1}$, the algorithm in (Ma et al., 2021) needs to search 7653 rate vectors, while we only need to find 2 of them for solving the MMF problem.

- We revisited the maximum multiframe problem that is NP-hard in theory.
- We provided an **efficient joint algorithm** that outputs optimal value in a finite number of iterations, while only requires a subset of the scheduling region.
- We studied the general **multi-source multi-sink** network with network coding.
- Our framework involves networks with non-negligible delay.
- We used simulation results to demonstrate our advantages.
- Future work: our maximum-mean-cycle algorithm could improve the calculation of the scheduling rate region (Ma et al., 2021; Yang et al., 2023).

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