



Submodular Interval Scheduling

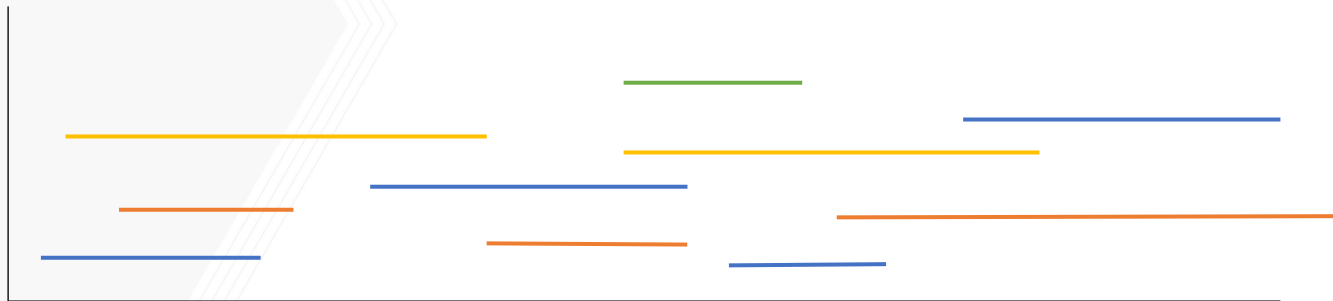
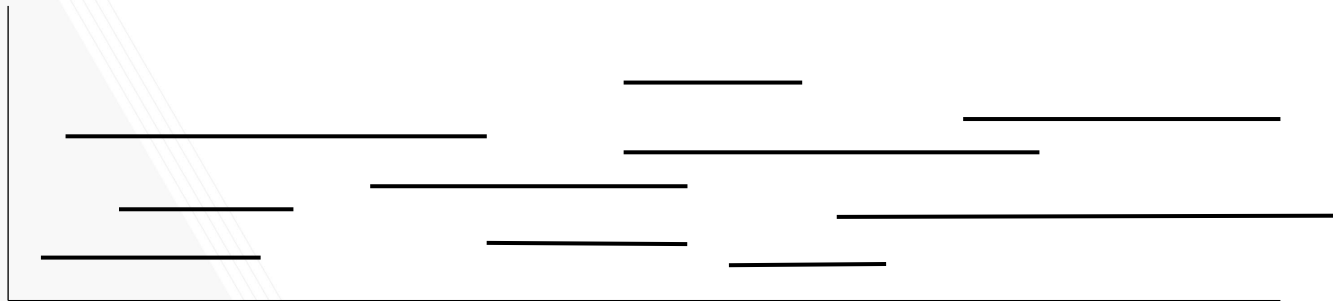
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Joint with Chris Muir

Submodular Interval Scheduling



$$cost = f(Y) + f(B) + f(O) + f(G)$$

Submodular Interval Scheduling

- **Input:** Set J of n jobs and a function $f: 2^J \rightarrow \mathbb{R}_+$
 - Job j : start time s_j , end time $e_j > s_j$, and weight $w_j > 0$
 - Monotone: $f(S) \leq f(T), S \subseteq T$
 - Submodular: $f(S \cap T) + f(S \cup T) \leq f(S) + f(T)$
 - Weight-defined: $f(S) = g(w_j, j \in S)$
- **Output:** Partition S_1, S_2, \dots, S_k of J minimizing $\sum f(S_i)$
 - S_i feasible if $[s_j, e_j] \cap [s_k, e_k] = \emptyset \quad \forall j, k \in S_i$
- **Applications:** crew scheduling, channel assignment, airline gate assignment, **cloud scheduling**

Related Work

- **Previous work**

- Submodular coloring
- Fukunaga et al 2007: 4-approximation for concave functions,
 - Prove NP-Hard for strictly concave functions
- Correa et al 2015: 5-approximation for value monotone-submodular functions
- Escoffier et al 2006: Special case max-weight function
 - Prove NP-Hard
- Mehrotra et al 1996: Set covering model for graph coloring

- **Our contributions**

- Propose set cover formulation
- Exact branch-and-price algorithm for max-weight and concave cost functions
 - Max-weight: $f(S) = \max_{i \in S} w_i$
 - Concave: $f(S) = h(\sum_{i \in S} w_i)$, h monotone, non-negative, concave

Set Cover Formulation

$$\begin{aligned} \min \quad & \sum_{S \in \mathcal{S}} f(S) z_S \\ \text{s. t.} \quad & \sum_{S \in \mathcal{S}(i)} z_S \geq 1 \quad \forall i \in J \\ & z_S \geq 0 \quad \forall S \in \mathcal{S} \end{aligned}$$

- \mathcal{S} : set of feasible schedules
- $\mathcal{S}(i)$: set of schedules containing i
- $O(2^{|J|})$ variables, solve with column generation

Pricing Problem

$$\max_{S \in \mathcal{S}} \left\{ \sum_{i \in S} \pi_i - f(S) \right\}$$

- π : current dual prices
- Interval packing with supermodular maximization objective

Pricing: Max-Weight Function

- Pricing problem: $\max_{S \in \mathcal{S}} \left\{ \sum_{i \in S} \pi_i - \max_{i \in S} w_i \right\}$

- For fixed value of $\max_{i \in S} w_i = \bar{w}$,
simplifies to an *interval packing* problem

$$\max_{S \in \mathcal{S}} \sum_{i \in S: w_i \leq \bar{w}} \pi_i$$

- Interval Packing DP: $v_i = \max\{\pi_i + v_{\eta_i}, v_{i+1}\} \forall i \in J: w_i \leq \bar{w}$
 - $\eta_i = \min\{j: j > i, s_j \geq e_i, w_j \leq \bar{w}\}$
 - $\Theta(n)$ running time
- Similar DP with pairs has $O(n^2)$ running time
- **Pricing concave function:** $O(n \sum_{i \in J} w_i)$ or $O(n^2 \sum_{i \in J} w_i)$ running time

Branch-and-Price

- Branching: i, j scheduled consecutively in some schedule
 - $\theta_{i,j} = \sum_{S \in \mathcal{S}} \mathbb{1}_{(i,j) \in S} z_S$
- Modified DP, select what job follows next in schedule
 - Max-weight function: $O(n^3)$ complexity
 - Concave functions: $O(n^2 \sum_{i \in J} w_i)$ complexity
- Implementation Details
 - Branch on most fractional $\theta_{i,j}$
 - Iterative depth-first exploration, explore $\theta_{i,j} = 1$ branches first
 - Initial solution: greedy heuristic + local search

Computational Study

- Random Instances
 - Start times $\sim \text{UniformInt}(0, t_{max})$
 - Lengths $\sim \text{UniformInt}(1, 10)$
 - Weights $\sim \text{UniformInt}(1, 100)$
 - Average of 10 replications
- Heuristics: greedy, IP-based, local search
- Max-weight: Compare B&P vs. assignment IP
- Use $f(S) = \sqrt{\sum_{i \in S} w_i}$ for concave function
 - Solve LP + primal heuristics

Max-Weight Function

$ J $	t_{\max}	Root Gap	B&P Gap	Sol. Time	% Sol.	Assign Gap	Sol. Time	% Sol.
100	100	0.94%	0.00%	41.56	100%	0.00%	436.45	100%
100	50	3.25%	0.00%	43.04	100%	0.75%	1199.34	60%
100	20	1.66%	0.00%	15.01	100%	0.68%	405.86	30%
250	250	1.54%	0.00%	5892.44	100%	2.01%	2228.56	50%
250	125	2.64%	0.00%	2779.20	100%	3.02%	-	0%
250	50	3.11%	0.00%	1006.74	100%	4.46%	-	0%

- Time limit: 3 hours
- Assignment IP solved with Gurobi
- B&P scales well to ~550 jobs

Square Root Function

$ J $	t_{\max}	CG Gap	CG Time	% Conv.	% Sol.
100	20	0.00%	252.55	100%	100%
250	50	0.30%	4959.18	70%	70%
400	80	1.30%	-	0%	0%

- Time limit: 6 hours
- If CG did not converge, used dual bounds
 - In-out column generation see Ben-Ameur et al 2007

Conclusion

- Proposed a set cover model for submodular interval scheduling
- Designed branch-and-price algorithm for two important cases

Future Work

- Extend to other submodular functions
- Extend to branch-price-and-cut

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