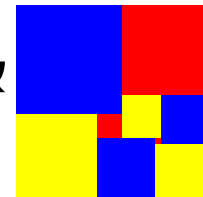


Packing via Covering and LP-Relative Approximation

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Linear Program–Based Combinatorial Optimization in a Nutshell

- LPs used for exact algorithms since mid-century (e.g., flow, matching, matroids)
 - Literature developed lots of *proof techniques* & *algorithmic techniques* (LP duality, total unimodularity, uncrossing, ellipsoid algorithm)
- Broad base of knowledge applied also to approximation algorithms ~30 years ago
 - Spurred more techniques, e.g. primal–dual schema, randomized rounding, scaling, grouping, iterated rounding/relaxation

Technique-Based Theory

- LP-based combinatorial optimization thereby exhibits the best & worst of mathematical problem-solving:

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- LP-based combinatorial optimization thereby exhibits the best & worst of mathematical problem-solving:
 - Best: techniques are elementary when considered individually, and combining them gives some very great results
 - Worst: it can be very hard to figure out how to combine them!

Approximation algorithms

- For NP-hard optimization problems optimal value OPT cannot be found by a poly-time algorithm (unless $P=NP$)
- Thus in poly-time the best we can do is find an *approximately* optimal answer
- Our convention: An α -*approximation algorithm* for a “max” problem always returns feasible solution with value at least OPT/α . For “min” problem, always returns value at most $\text{OPT} \cdot \alpha$. ($\alpha \geq 1$; exact $\equiv \alpha = 1$)

Iterated LP Relaxation

- Background: Iterated relaxation (Lau, Naor, Salavatipour, Singh, STOC 2007) finds solutions with *additive violation of constraints* instead of multiplicative factor
 - Bounded degree spanning tree: can find a spanning tree with super-optimal weight but violating degree bounds by up to +1 (SL '07)
 - Integral multicommodity flow in a tree: can find a flow with super-optimal weight but violating edge capacities by up to +2 (KPP '08)



Contribution of Our Work

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- Can iterated rounding help when constraints are inflexible?
- Our contribution: Techniques that remove violation (at expense of value)
 - General approach that works in most iterated rounding situations (where a “counting lemma” exists)
- Gives new algorithmic results
 - Ultimate goal: understanding best possible approximation ratios for problems

Rest of Talk

- Defⁿ: Multicommodity Flow in a Tree (MFT)
- Defⁿ: LP-Relative Approximation Ratio
- Counting lemma for MFT and consequences
- Results in case of MFT:
 - $1 + 1/O(\textit{minimum capacity})$ approx algorithms
- Mini-Technique: LP-Relative Covers Help Cover
- Technique: LP-Relative Covers Help Pack
- General Form of Results, Open Problems

Integer multicommodity flow in a tree (MFT)

- **Input:** tree with edge capacities c_e ; pairs of terminals, profit w_i for each commodity i .
 - Let $path(i)$ denote path between terminals for i
- **Goal:** integers $x_i \geq 0$ such that for each e

$$\sum_{i: e \in path(i)} x_i \leq c_e$$

such that $\sum_i w_i x_i$ is maximized

- E.g. ship kegs on tree network
- APX-complete (GVY '93),
4-*apx* (CMS '03)



Key Notion: LP-Relative Approximation

- *Natural LP* for MFT: flow values x can be fractional, otherwise the same. LP-OPT can be found in poly-time. Note $\text{LP-OPT} \geq \text{OPT}$.
- Defⁿ. An *LP-relative α -approx. algorithm* for a “max” problem always returns value at least $\text{LP-OPT}/\alpha$; “min” is analogous
 - Same as “ α -approx.” defⁿ with $\text{OPT} \Rightarrow \text{LP-OPT}$
 - Stronger notion
 - Abundant in papers but no common term?
 - E.g. 4-approx for MFT, min degree+1 spanning tree

Counting Lemma for MFT

- (KPP 08) “Let x be an extreme solution to the natural LP for MFT. If $x_i < 1$ for every commodity i , then some edge e^* has the following property:
 - At most 3 commodities i have $x_i > 0$ and i in $path(e^*)$ ”
- Same works for *covering version* of MFT
- (‘09) Also holds for arc- and vertex-capacitated versions of MFT, with 3 replaced by 7
 - Won’t state these versions explicitly from now on but all results in this talk go through

Counting Lemma Consequences

- i) Capacitated covering-MFT: some var has value $\geq 1/3 \Rightarrow 3$ -apx by iterated rounding

ratio \swarrow violation \nwarrow

- ii) $(1, +2)$ -approximation for MFT by iterated relaxation as mentioned earlier
- iii) $(1, -2)$ -approximation for covering-MFT
- Key: these results are all LP-relative!

Results as Applied To MFT

- Let μ be minimum (edge/arc/ v^x) capacity
- We get $1+O(1/\mu)$ approximation algorithm
 - Previous best ratio: constant (4/4/5)
 - Asymptotically optimal in terms of μ : $\exists \epsilon > 0 \forall \mu$, we get $1+O(\epsilon/\mu)$ inapproximability
 - Also get $1+O(1/\mu)$ approximation algorithm for covering-MFT

- Next up: sketches of the techniques

Mini-Technique: LP-Relative Covers Help Cover (e.g.: MFT)

□ *There is a $(1+2/\mu)$ -apx alg for covering-MFT*

□ Proof

- Artificially increase all requirements c_e by 2
- Scaling: if x is an optimum to the old LP, then $x^*(2+\mu)/\mu$ is a solution to the new LP
- So $\text{LP-OPT}' \leq \text{LP-OPT}^*(2+\mu)/\mu$
- Now apply the LP-relative $(1, -2)$ approximation to the new requirements, gives a solution which meets old requirements and has cost at most

$$1 * \text{LP-OPT}' \leq \text{LP-OPT}^*(2+\mu)/\mu$$

Technique: LP-Relative Covers Help Pack (e.g.: MFT)

- *There is a $1+O(1/\mu)$ -apx alg for MFT*
- **Proof**
 - Let x be output of $(1, +2)$ approx. algorithm
 - For each e let f_e be overload of e by x ($0 \leq f_e \leq 2$)
 - Look @ capacitated MFT-covering w/ requirements f and capacities x :
 - Scaling: $x \cdot 2/(2+\mu)$ is a feasible fractional covering
 - So $\text{LP-OPT}' \leq c(x) \cdot 2/(2+\mu)$
 - Run 3-apx alg for capacitated cover-MFT $\Rightarrow y$
 - $x - y$ is a solution to original MFT instance and $c(x-y) \geq c(x) \cdot (1 - 3 \cdot 2/(2+\mu)) \geq \text{LP-OPT}/(1+O(1/\mu))$

Generalized Results

- Consider a family of integer linear programs (A, b)
- Define $\mu := b_{\min}$
- If “counting lemma” \Rightarrow known LP-rel consequences
 - $(\alpha', -v')$ for uncapacitated covering $\min \{cx \mid Ax \geq b, 0 \leq x\}$
 - $(\alpha, +v)$ for capacitated packing $\max \{cx \mid Ax \leq b, 0 \leq x \leq d\}$
 - β for capacitated covering $\min \{cx \mid Ax \geq b, 0 \leq x \leq d\}$
- Our techniques give in addition:
 - $\alpha'(1+v'/\mu) = \alpha'(1+O(1/\mu))$ for covering
 - $\alpha/(1-\beta v/(\mu+v)) =^* \alpha(1+O(1/\mu))$ for capacitated packing
 - *Remark:* can't hope to $1+O(1/\mu)$ -approx *capacitated covering* in general since we can increase μ artificially

Related Open Problems

- Smallest k -edge-connected subgraph [GG 08]: best known apx ratio is $1 + 1/2k + O(1/k^2) + \dots$
 - What about *min-cost* k -edge-connected multisubgraph?
 - Best known apx is 2, can we get $1+O(1/k)$?
- [*] Our packing techniques give no result when μ is too small. Is this avoidable or not?
 - E.g. *column-sparse packing ILPs* admit a counting lemma but need ad-hoc techniques to get const ratio for small μ
- Demand multicommodity flow in a tree?
 - Case of a star (*demand matching*) is well-studied and also a special case of column-sparse integer programs
 - $1+O(\text{dem}_{\max}/\mu)^{1/2}$ best known, can we get $1+O(\text{dem}_{\max}/\mu)$?

Merci!

