PARTIAL DEGREE BOUNDED EDGE PACKING PROBLEM WITH ARBITRARY BOUNDS

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PROBLEM DEFINITION

PDBEP PROBLEM

• Input: Graph G = (V, E) and degree-bound function $c: V \to \mathbb{Z}^*$.

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- Zhang showed PDBEP is NP-Hard even for uniform $c_v = 1$.

Zhang, 2012

For uniform $c_v = k$:

• 2-approx for k = 1.

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For uniform $c_v = k$:

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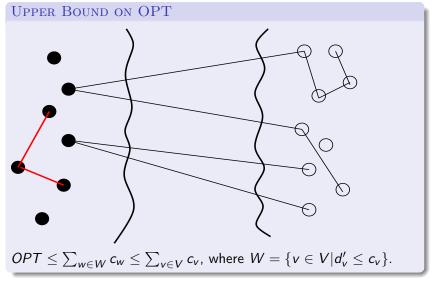
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- 2 + log₂ *n*-approx with weights on edges.

UPPER BOUND



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Edge Deletion based Algorithm

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- Observe that $d_Y(v) \ge c_v \ \forall \ v \in V$.
- Hence $|Y| \ge \sum_{v} c_{v}/2 \ge OPT/2$.

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• Let H(v) be the heaviest c_v edges incident on vertex v.

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- Then the heaviest solution gives a $2 + \log_2 |V|$ approximation of the problem.

EXACT ALGORITHM

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- In order to compute the values for internal nodes we need to sort the child nodes with respect to some key values.
- Thus at each vertex we incur $O(|Ch| \log |Ch|)$ cost.
- That gives an overall time complexity of $O(n \log n)$.

THE NATURAL IP FORMULATION OF THE PROBLEM IP1: max $\psi = \sum_{e \in E} y_e$, subject to

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- Consider K_n with uniform degree constraint $c_v = 1$.
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 ψ ≥ (n 1)²/4.

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- $\psi \ge (n-1)^2/4$.
- However, $OPT \leq n$.
- \implies integrality gap = $\Omega(n)$.

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c) $y_e \in \{0, 1\} \ \forall e \in E$.

LEMMA

Every maximal solution of IP2 is also a feasible solution of PDBEP.

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If $c_v \ge (1 - \beta)d_v \forall v$, then any α approximate solution of IP2, which is also maximal, is a $2\alpha/(2 - (1 + \epsilon)\beta)$ approximation of PDBEP.

LP RELAXATION

LP RELAXATION OF IP2 LP2: max $\phi = 2 \sum_{e \in E} y_e - (1 + \epsilon) \sum_{v \in V} z_v$, subject to $\sum_{e \in \delta(v)} y_e \le c_v + z_v \ \forall v \in V$. $z_v \ge 0 \ \forall v \in V$. $y_e \ge 0 \ \forall e \in E$. $-y_e \ge -1 \ \forall e \in E$.

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In a corner solution of LP2 on a non-empty graph there is at least one edge e with $y_e = 0$ or $y_e \ge 1/2$.

Algorithm using LP2

ITERATIVE ROUNDING ALGORITHM

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Lemma

The Iterative Rounding Algorithm returns a feasible solution of PDBEP.

LEMMA

The Iterative Rounding Algorithm gives a $1.5/(1-\epsilon)$ approximation of IP2.



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Theorem

The Iterative Rounding Algorithm approximates PDBEP with approximation factor $3/(1-\epsilon)^2$.

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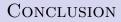
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THEOREM

The Iterative Rounding Algorithm approximates PDBEP with approximation factor $3/(1-\epsilon)^2$.

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If $c_v \ge (1 - \beta)d_v$ for all v, then The Iterative Rounding Algorithm approximates PDBEP with approximation factor $\frac{3}{(2-(1+\epsilon)\beta)(1-\epsilon)}$.



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• LP based solution for the weighted case.

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• Establishing lower bounds for both the weighted and unweighted PDBEP.

- LP based solution for the weighted case.
- Generalization to hypergraphs.

Thanks!