3. Branching Algorithms

COMP6741: Parameterized and Exact Computation

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Semester 2, 2015

Outline

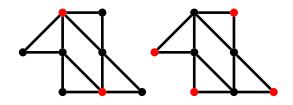
- Introduction
- Maximum Independent Set
 - Simple Analysis
 - Search Trees and Branching Numbers
 - Measure Based Analysis
 - Optimizing the measure
 - Exponential Time Subroutines
 - Structures that arise rarely
 - State Based Measures
- Exercise on Max 2-CSP
- Further Reading

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Recall: Maximal Independent Sets

- A vertex set $S \subseteq V$ of a graph G = (V, E) is an independent set in G if there is no edge $uv \in E$ with $u, v \in S$.
- An independent set is maximal if it is not a subset of any other independent set.
- Examples:



Enumeration problem: Enumerate all maximal independent sets

ENUM-MIS

Input: graph G

Output: all maximal independent sets of G



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Maximal independent sets: $\{a, d\}, \{b\}, \{c\}$

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Maximal independent sets: $\{a, d\}, \{b\}, \{c\}$

Note: Let v be a vertex of a graph G. Every maximal independent set contains a vertex from $N_G[v]$.

```
Algorithm enum-mis(G,I)
Input : A graph G=(V,E), an independent set I of G.
Output: All maximal independent sets of G that are supersets of I.

1 G' \leftarrow G - N_G[I]
2 if V(G') = \emptyset then // G' has no vertex 3 \bigcup Output I
4 else
5 Select v \in V(G') such that d_{G'}(v) = \delta(G') // v has min degree in G'
Run enum-mis(G,I \cup \{u\}) for each u \in N_{G'}[v]
```

Running Time Analysis

Define L(n)= largest number of leaves in any search tree of **enum-mis** for an instance with $|V(G')| \leq n$.

Note: L(n) is non-decreasing.

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Note: L(n) is non-decreasing.

Suppose $d_{G'}(v) = d$ generates a maximum number of leaves. Then,

$$L(n) \le (d+1) \cdot L(n-(d+1)) = O\left((d+1)^{n/(d+1)}\right)$$

For s>0, the function $f(s)=s^{1/s}$ has its maximum value for s=e and for integer s the maximum value of f(s) is when s = 3.

Running Time Analysis II

Since the height of the search trees is $\leq |V(G')|$, we obtain:

Theorem 1

Algorithm enum-mis has running time $O^*(3^{n/3}) \subseteq O(1.4423^n)$, where n = |V|.

Corollary 2

A graph on n vertices has $O(3^{n/3})$ maximal independent sets.

Constraints Based Analysis

Suppose
$$L(n) = 2^{\alpha \cdot n}$$
, $\alpha > 0$.

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We constrain for each $d \geq 0$, that

$$2^{\alpha \cdot n} \ge (d+1) \cdot 2^{\alpha \cdot (n - (d+1))},$$

or, equivalently,

$$1 \ge (d+1) \cdot 2^{\alpha \cdot (-(d+1))},$$

and, since we would like to prove a small running time bound, we **minimize** α subject to these constraints.

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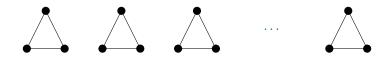
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and, since we would like to prove a small running time bound, we **minimize** α subject to these constraints.

This amounts to solving a convex program, which gives $\alpha=(1/3)\cdot\log_23$ and $L(n)=2^{(n/3)\cdot\log_23}=3^{n/3}.$

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Running Time Lower Bound



Theorem 3

There is an infinite family of graphs with $\Omega(3^{n/3})$ maximal independent sets.

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Branching Algorithm

Branching Algorithm

- Selection: Select a local configuration of the problem instance
- Recursion: Recursively solve subinstances
- Combination: Compute an optimal solution of the instance based on the optimal solutions of the subinstances
- Simplification rule: 1 recursive call
- Branching rule: ≥ 2 recursive calls

Outline

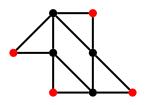
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MAXIMUM INDEPENDENT SET

MAXIMUM INDEPENDENT SET

Input: graph G

Output: A largest independent set of G.



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Exercise

Suppose there exists a $O^*(1.2^n)$ time algorithm, which, given a graph G on n vertices, computes **the size** of a largest independent set of G.

Design an algorithm, which, given a graph G, **finds** a largest independent set of G in time $O^*(1.2^n)$.

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Solution Idea

- Compute k, the size of a largest independent set of G
- ullet Find a vertex v belonging to an independent set of size k
 - \bullet We can do this by going through each vertex u of G , and checking whether $G-N_G[u]$ has an independent set of size k-1
- Recurse on $(G N_G[v], k 1)$

```
Algorithm mis(G)
 Input : A graph G = (V, E).
 Output: The size of a maximum i.s. of G.
1 if \Delta(G) \leq 2 then
                                                  // G has max degree \leq 2
return the size of a maximum i.s. of G in polynomial time
3 else if \exists v \in V : d(v) = 1 then
                                                          //v has degree 1
4 return 1 + \min(G - N[v])
5 else if G is not connected then
6 Let G_1 be a connected component of G
7 return mis(G_1) + mis(G - V(G_1))
8 else
9 | Select v \in V s.t. d(v) = \Delta(G) // v has max degree
o return \max(1 + \min(G - N[v]), \min(G - v))
```

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Lemma 4 (Simple Analysis Lemma)

Let

- A be a branching algorithm
- $\alpha > 0, \ c \geq 0$ be constants

such that on input I, A calls itself recursively on instances I_1, \ldots, I_k , but, besides the recursive calls, uses time $O(|I|^c)$, such that

$$(\forall i: 1 \le i \le k) \quad |I_i| \le |I| - 1, \text{ and}$$
 (1)

$$2^{\alpha \cdot |I_1|} + \dots + 2^{\alpha \cdot |I_k|} \le 2^{\alpha \cdot |I|}. \tag{2}$$

Then A solves any instance I in time $O(|I|^{c+1}) \cdot 2^{\alpha \cdot |I|}$.

Simple Analysis II

 $< |I|^{c+1} 2^{\alpha \cdot |I|}$.

Proof.

By induction on |I|.

W.l.o.g., replace the hypotheses' O statement with a simple inequality, and for the base case assume that the algorithm returns the solution to an empty instance in time $1 \leq |I|^{c+1}2^{\alpha \cdot |I|}$.

Suppose the lemma holds for all instances of size at most $|I|-1\geq 0$, then the running time of algorithm A on instance I is

$$\begin{split} T_A(I) &\leq |I|^c + \sum_{i=1}^k T_A(I_i) & \text{(by definition)} \\ &\leq |I|^c + \sum_{i=1}^k |I_i|^{c+1} 2^{\alpha \cdot |I_i|} & \text{(by the inductive hypothesis)} \\ &\leq |I|^c + (|I|-1)^{c+1} \sum_{i=1}^k 2^{\alpha \cdot |I_i|} & \text{(by (1))} \\ &\leq |I|^c + (|I|-1)^{c+1} 2^{\alpha \cdot |I|} & \text{(by (2))} \end{split}$$

The final inequality uses that $\alpha \cdot |I| > 0$ and holds for any $c \ge 0$.

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Simple Analysis for mis

- At each node of the search tree: $O(n^2)$
- G disconnected:

$$(\forall s: 1 \le s \le n-1) \quad 2^{\alpha \cdot s} + 2^{\alpha \cdot (n-s)} \le 2^{\alpha \cdot n}. \tag{3}$$

always satisfied by convexity of the function 2^x

• Branch on vertex of degree $d \geq 3$

$$(\forall d: 3 \le d \le n-1) \quad 2^{\alpha \cdot (n-1)} + 2^{\alpha \cdot (n-1-d)} \le 2^{\alpha n}.$$
 (4)

Dividing all these terms by $2^{\alpha n}$, the constraints become

$$2^{-\alpha} + 2^{\alpha \cdot (-1-d)} \le 1. \tag{5}$$

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Compute optimum a

The minimum α satisfying the constraints is obtained by solving a convex mathematical program minimizing α subject to the constraints (the constraint for d=3 is sufficient as all other constraints are weaker).

Compute optimum α

The minimum α satisfying the constraints is obtained by solving a convex mathematical program minimizing α subject to the constraints (the constraint for d=3 is sufficient as all other constraints are weaker).

Alternatively, set $x:=2^{\alpha}$, compute the unique positive real root of each of the characteristic polynomials

$$c_d(x) := x^{-1} + x^{-1-d} - 1,$$

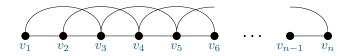
and take the maximum of these roots [Kullmann '99].

d	\boldsymbol{x}	α
3	1.3803	0.4650
4	1.3248	0.4057
5	1.2852	0.3620
6	1.2555	0.3282
7	1.2321	0.3011

Simple Analysis: Result

- use the Simple Analysis Lemma with c=2 and $\alpha=0.464959$
- running time of Algorithm **mis** upper bounded by $O(n^3) \cdot 2^{0.464959 \cdot n} = O(2^{0.4650 \cdot n}) \text{ or } O(1.3803^n)$

Lower bound



$$T(n) = T(n-5) + T(n-3)$$

- \bullet for this graph, P_n^2 , the worst case running time is $1.1938\dots^n\cdot\operatorname{poly}(n)$
- \bullet Run time of algo \mathbf{mis} is $\Omega(1.1938^n)$

Worst-case running time — a mystery

Mystery

What is the worst-case running time of Algorithm mis?

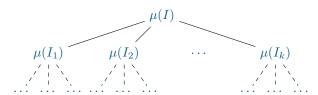
- lower bound $\Omega(1.1938^n)$
- upper bound $O(1.3803^n)$

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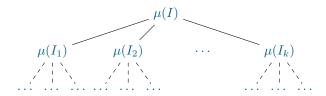
Search Trees

Denote $\mu(I) := \alpha \cdot |I|$.

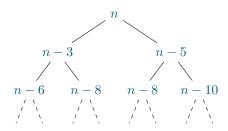


Search Trees

Denote $\mu(I) := \alpha \cdot |I|$.



Example: execution of **mis** on a P_n^2



Branching number: Definition

Consider a constraint

$$2^{\mu(I)-a_1} + \dots + 2^{\mu(I)-a_k} \le 2^{\mu(I)}$$
.

Its branching number is

$$2^{-a_1} + \cdots + 2^{-a_k}$$
,

and is denoted by

$$(a_1,\ldots,a_k)$$
.

Clearly, any constraint with branching number at most 1 is satisfied.

Branching numbers: Properties

Dominance For any a_i, b_i such that $a_i \geq b_i$ for all $i, 1 \leq i \leq k$,

$$(a_1,\ldots,a_k)\leq (b_1,\ldots,b_k)\,,$$

as $2^{-a_1} + \dots + 2^{-a_k} \le 2^{-b_1} + \dots + 2^{-b_k}$. In particular, for any a, b > 0,

$$\text{either} \qquad (a,a) \leq (a,b) \qquad \text{or} \qquad (b,b) \leq (a,b) \, .$$

Balance If $0 < a \le b$, then for any ε such that $0 \le \varepsilon \le a$,

$$(a,b) \le (a-\varepsilon,b+\varepsilon)$$

by convexity of 2^x .

Exercises

Exercises

1. Let A be a branching algorithm, such that, on any input of size at most n its search tree has height at most n and for the number of leaves L(n), we have

$$L(n) \le 3 \cdot L(n-2)$$

Upper bound the running time of A, assuming it spends only polynomial time at each node of the search tree.

Exercises

1. Let A be a branching algorithm, such that, on any input of size at most n its search tree has height at most n and for the number of leaves L(n), we have

$$L(n) \le 3 \cdot L(n-2)$$

Upper bound the running time of A, assuming it spends only polynomial time at each node of the search tree.

2. Same question, except that

$$L(n) \le \max \begin{cases} 2 \cdot L(n-3) \\ L(n-2) + L(n-4) \\ 2 \cdot L(n-2) \\ L(n-1) \end{cases}$$

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Measure based analysis

- Goal, idea
 - capture more structural changes when branching into subinstances
- Means
- potential-function method, a.k.a., Measure & Conquer
- Example: Algorithm mis
 - advantage when degrees of vertices decrease

Multivariate recurrences

• Model running time of mis by

$$T(n_1, n_2, \ldots)$$
, short $T\left(\left\{n_i\right\}_{i \geq 1}\right)$,

where
$$n_i := |\{v \in V : d(v) = i\}|.$$

- G-v: neighbors' degrees decrease
- ullet G-N[v]: a vertex in $N^2[v]$ has its degree decreased

Multivariate recurrences (2)

• We obtain the following recurrence where the maximum ranges over all $d \geq 3$, all $p_i, 2 \leq i \leq d$ such that $\sum_{i=2}^d p_i = d$ and all k such that $2 \leq k \leq d$:

$$T\left(\left\{n_{i}\right\}_{i\geq1}\right) = \max_{d,p_{2},\dots,p_{d},k} \begin{cases} T\left(\left\{n_{i} - p_{i} + p_{i+1} - [d=i]\right\}_{i\geq1}\right) \\ +T\left(\left\{n_{i} - p_{i} - [d=i] - [k=i]\right\}_{i\geq1}\right) \\ + [k=i+1]\right\}_{i\geq1} \end{cases}$$
(6)

where the Iverson bracket $[F] = \begin{cases} 1 \text{ if } F \text{ true} \\ 0 \text{ otherwise} \end{cases}$

Solve multivariate recurrence

- restrict to max degree 5
- [Eppstein 2004]: there exists a set of weights $w_1, \ldots, w_5 \in \mathbb{R}^+$ such that a solution to (6) is within a polynomial factor of a solution to the corresponding univariate weighted model $(T(\sum_{i=1}^5 \omega_i n_i) = \max \ldots)$.

Definition 5

A measure μ for a problem P is a function from the set of all instances for P to the set of non negative reals

$$\begin{split} \mu(G) := \sum_{i=1}^5 w_i n_i \\ (i \ge 1) \quad w_i \ge 0 \\ (i \ge 2) \quad w_i \ge w_{i-1} \\ (\forall d: 2 \le d \le 5) \quad h_d := \min_{2 \le i \le d} \{w_i - w_{i-1}\} \end{split}$$

By [Eppstein 2004], there exist weights w_i such that a solution to (6) corresponds to a solution to the following recurrence, where the maximum ranges over all $d, 3 \leq d \leq 5$, and all $p_i, 2 \leq i \leq d$, such that $\sum_{i=2}^d p_i = d$,

$$T\left(\mu(G)\right) = \max_{d, p_2, \dots, p_d, k} \left\{ T\left(\mu(G) - w_d - \sum_{i=2}^d p_i \cdot (w_i - w_{i-1})\right) + T\left(\mu(G) - w_d - \sum_{i=2}^d p_i \cdot w_i - h_d\right). \right.$$

... to constraints

$$T(\mu(G)) \ge T\left(\mu(G) - w_d - \sum_{i=2}^d p_i \cdot (w_i - w_{i-1})\right) + T\left(\mu(G) - w_d - \sum_{i=2}^d p_i \cdot w_i - h_d\right)$$

for all $d, 3 \le d \le 5$, and all $p_i, 2 \le i \le d$, such that $\sum_{i=2}^d p_i = d$.

Measure Based Analysis

Lemma 6 (Measure Analysis Lemma)

Let

- A be a branching algorithm
- c > 0 be a constant, and
- $\mu(\cdot), \eta(\cdot)$ be two measures for the instances of A,

such that on input I, A calls itself recursively on instances I_1, \ldots, I_k , but, besides the recursive calls, uses time $O(|I|^c)$, such that

$$(\forall i) \quad \eta(I_i) \le \eta(I) - 1, \text{ and} \tag{7}$$

$$2^{\mu(I_1)} + \ldots + 2^{\mu(I_k)} \le 2^{\mu(I)}.$$
(8)

Then A solves any instance I in time $O(\eta(I)^{c+1}) \cdot 2^{\mu(I)}$.

Applying the lemma

$$\begin{aligned} w_i &\geq 0 \\ w_i &\geq w_{i-1} \\ 2^{\mu(G)} &\geq 2^{\mu(G) - w_d - \sum_{i=2}^d p_i \cdot (w_i - w_{i-1})} + 2^{\mu(G) - w_d - \sum_{i=2}^d p_i \cdot w_i - h_d} \\ &\Leftrightarrow \\ 1 &\geq 2^{-w_d - \sum_{i=2}^d p_i \cdot (w_i - w_{i-1})} + 2^{-w_d - \sum_{i=2}^d p_i \cdot w_i - h_d} \end{aligned}$$

Applying the lemma

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i	w_i	h_i
1	0	0
2	0.25	0.25
3	0.35	0.10
4	0.38	0.03
5	0.40	0.02

These values for w_i satisfy all the constraints and $\mu(G) \leq 2n/5$ for any graph of max degree ≤ 5 .

Taking c=2 and $\eta(G)=n$, the Measure Analysis Lemma shows that **mis** has run time $O(n^3)2^{2n/5}=O(1.3196^n)$ on graphs of max degree ≤ 5 .

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Compute optimal weights

• By convex programming [Gaspers, Sorkin 2009]

All constraints are already convex, except conditions for h_{d}

$$(\forall d: 2 \leq d \leq 5) \quad h_d := \min_{2 \leq i \leq d} \left\{ w_i - w_{i-1} \right\}$$

$$\downarrow \downarrow$$

$$(\forall i, d: 2 \leq i \leq d \leq 5) \quad h_d \leq w_i - w_{i-1}.$$

Use existing convex programming solvers to find optimum weights.

```
param maxd integer >= 3;
set DEGREES := 0..maxd;
var W {DEGREES} >= 0: # weight for vertices according to their degrees
var g {DEGREES} >= 0; # weight for degree reductions from deg i
var h {DEGREES} >= 0; # weight for degree reductions from deg \le i
                       # maximum weight of W[d]
var Wmax:
minimize Obj: Wmax; # minimize the maximum weight
subject to MaxWeight {d in DEGREES}:
  Wmax >= W[d]:
subject to gNotation {d in DEGREES : 2 <= d}:
  g[d] \le W[d] - W[d-1];
subject to hNotation {d in DEGREES. i in DEGREES : 2 <= i <= d}:
  h[d] \le W[i] - W[i-1];
subject to Deg3 {p2 in 0..3, p3 in 0..3 : p2+p3=3}:
  2^{-}(-W[3] - p2*g[2] - p3*g[3]) + 2^{-}(-W[3] - p2*W[2] - p3*W[3] - h[3]) <=1;
subject to Deg4 {p2 in 0..4, p3 in 0..4, p4 in 0..4 : p2+p3+p4=4}:
  2^{-(-W[4] - p2*g[2] - p3*g[3] - p4*g[4])}
+ 2^{-4} - p2*W[2] - p3*W[3] - p4*W[4] - h[4]) <=1;
subject to Deg5 {p2 in 0..5, p3 in 0..5, p4 in 0..5, p5 in 0..5 :
                 p2+p3+p4+p5=5}:
  2^{-}(-W[5] - p2*g[2] - p3*g[3] - p4*g[4] - p5*g[5])
+ 2^{(-W[5]} - p2*W[2] - p3*W[3] - p4*W[4] - p5*W[5] - h[5]) <=1:
```

Optimal weights

i	w_i	h_i
1	0	0
2	0.206018	0.206018
3	0.324109	0.118091
4	0.356007	0.031898
5	0.358044	0.002037

- use the Measure Analysis Lemma with $\mu(G)=\sum_{i=1}^5 w_i n_i \leq 0.358044 \cdot n$, c=2, and $\eta(G)=n$
- \bullet mis has running time $O(n^3)2^{0.358044 \cdot n} = O(1.2817^n)$

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Exponential time subroutines

Lemma 7 (Combine Analysis Lemma)

Let

- ullet A be a branching algorithm and B be an algorithm,
- $c \geq 0$ be a constant, and
- $\mu(\cdot), \mu'(\cdot), \eta(\cdot)$ be three measures for the instances of A and B,

such that $\mu'(I) \leq \mu(I)$ for all instances I, and on input I, A either solves I by invoking B with running time $O(\eta(I)^{c+1}) \cdot 2^{\mu'(I)}$, or calls itself recursively on instances I_1, \ldots, I_k , but, besides the recursive calls, uses time $O(|I|^c)$, such that

$$(\forall i) \quad \eta(I_i) \le \eta(I) - 1, \text{ and}$$
 (9)

$$2^{\mu(I_1)} + \ldots + 2^{\mu(I_k)} \le 2^{\mu(I)}. \tag{10}$$

Then A solves any instance I in time $O(\eta(I)^{c+1}) \cdot 2^{\mu(I)}$.

Algorithm **mis** on general graphs

- use the Combine Analysis Lemma with A=B= mis, c=2, $\mu(G)=0.35805n$, $\mu'(G)=\sum_{i=1}^5 w_i n_i$, and $\eta(G)=n$
- for every instance G, $\mu'(G) \leq \mu(G)$ because $\forall i, w_i \leq 0.35805$
- for each $d \geq 6$,

$$(0.35805, (d+1) \cdot 0.35805) \le 1$$

ullet Thus, Algorithm **mis** has running time $O(1.2817^n)$ for graphs of arbitrary degrees

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Rare Configurations

- Branching on a local configuration C does not influence overall running time
 if C is selected only a constant number of times on the path from the root to
 a leaf of any search tree corresponding to the execution of the algorithm
- Can be proved formally by using measure

$$\mu'(I) := \begin{cases} \mu(I) + c & \text{if } C \text{ may be selected in the current subtree} \\ \mu(I) & \text{otherwise}. \end{cases}$$

Avoid branching on regular instances in mis

else

Select $v \in V$ such that

- (1) v has maximum degree, and
- (2) among all vertices satisfying (1), v has a neighbor of minimum degree

return
$$\max (1 + \min(G - N[v]), \min(G - v))$$

New measure:

$$\mu'(G) = \mu(G) + \sum_{d=3}^5 [G \text{ has a d-regular subgraph}] \cdot C_d$$

where C_d , $3 \le d \le 5$, are constants.

Resulting Branching numbers

For each $d, 3 \leq d \leq 5$ and all $p_i, 2 \leq i \leq d$ such that $\sum_{i=2}^d p_i = d$ and $p_d \neq d$,

$$\left(w_d + \sum_{i=2}^d p_i \cdot (w_i - w_{i-1}), w_d + \sum_{i=2}^d p_i \cdot w_i + h_d\right).$$

All these branching numbers are at most 1 with the optimal set of weights on the next slide

S. Gaspers (UNSW)

i	w_i	h_i
1	0	0
2	0.207137	0.207137
3	0.322203	0.115066
4	0.343587	0.021384
5	0.347974	0.004387

Thus, the modified Algorithm \mathbf{mis} has running time $O(2^{0.3480 \cdot n}) = O(1.2728^n).$

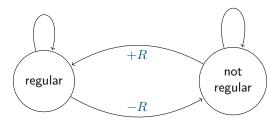
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State based measures

- "bad" branching always followed by "good" branchings
- amortize over branching numbers

$$\mu'(I) := \mu(I) + \Psi(I),$$

where $\Psi: \mathcal{I} \to \mathbb{R}^+$ depends on global properties of the instance.



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Exercise: Max 2-CSP

Max 2-CSP

Input: A graph G = (V, E) and a set S of score functions containing

- a score function $s_e: \{0,1\}^2 \to \mathbb{N}_0$ for each edge $e \in E$,
- ullet a score function $s_v:\{0,1\} o\mathbb{N}_0$ for each vertex $v\in V$, and
- a score "function" $s_{\emptyset}: \{0,1\}^0 \to \mathbb{N}_0$ (which takes no arguments and is just a constant convenient for bookkeeping).

Output: The maximum score $s(\phi)$ of an assignment $\phi: V \to \{0,1\}$:

$$s(\phi) := s_{\emptyset} + \sum_{v \in V} s_v(\phi(v)) + \sum_{uv \in E} s_{uv}(\phi(u), \phi(v)).$$

- **1** Design simplification rules for vertices of degree ≤ 2 .
- ② Using the simple analysis, design and analyze an $O^*(2^{m/4})$ time algorithm, where m=|E|.
- ① Use the measure $\mu:=w_e\cdot m-\left(\sum_{v\in V}w_{d_G(v)}\right)$ to improve the analysis to $O^*(2^{m/5}).$

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Solution sketch

Simplification rules

- S0 If there is a vertex y with d(y)=0, then set $s_\emptyset=s_\emptyset+\max_{C\in\{0,1\}}s_y(C)$ and delete y from G.
- S1 If there is a vertex y with d(y)=1, then denote $N(y)=\{x\}$ and replace the instance with (G',S') where G'=(V',E')=G-y and S' is the restriction of S to V' and E' except that for all $C\in\{0,1\}$ we set

$$s_x'(C) = s_x(C) + \max_{D \in \{0,1\}} \{s_{xy}(C,D) + s_y(D)\}.$$

S2 If there is a vertex y with d(y)=2, then denote $N(y)=\{x,z\}$ and replace the instance with (G',S') where $G'=(V',E')=(V-y,(E\setminus\{xy,yz\})\cup\{xz\})$ and S' is the restriction of S to V' and E', except that for $C,D\in\{0,1\}$ we set

$$s'_{xz}(C,D) = s_{xz}(C,D) + \max_{F \in \{0,1\}} \{s_{xy}(C,F) + s_{yz}(F,D) + s_y(F)\}$$

if there was already an edge xz, discarding the first term $s_{xz}(C,D)$ if there was not.

Solution sketch II

Branching rules

B Let y be a vertex of maximum degree. There is one subinstance (G',s^C) for each color $C\in\{0,1\}$, where G'=(V',E')=G-y and s^C is the restriction of s to V' and E', except that we set

$$(s^C)_{\emptyset} = s_{\emptyset} + s_y(C),$$

and, for every neighbor x of y and every $D \in \{0, 1\}$,

$$(s^C)_x(D) = s_x(D) + s_{xy}(D, C).$$

- \bullet Branching on a vertex of degree ≥ 4 removes ≥ 4 edges from both subinstances
- Branching on a vertex of degree 3 removes ≥ 6 edges from both subinstances since G is 3-regular.

The recurrence $T(m) \leq 2 \cdot T(m-4)$ solves to $2^{m/4}$

Solution sketch III

Using the measure

$$\mu := w_e \cdot m + \left(\sum_{v \in V} w_{d_G(v)}\right)$$

we constrain that

$$\begin{array}{ll} w_d \leq 0 & \text{for all } d \geq 0 \text{ to ensure that } \mu \leq w_e m \\ d \cdot w_e/2 + w_d \geq 0 & \text{for all } d \geq 0 \text{ to ensure that } \mu(G) \geq 0 \\ -w_0 \leq 0 & \text{constraint for S0} \\ -w_2 - w_e \leq 0 & \text{constraint for S2} \end{array}$$

$$1 - w_d - d \cdot w_e - d \cdot (w_j - w_{j-1}) \le 0$$

for all $d, j \geq 3$.

Using $w_e=0.2$, $w_0=0$, $w_1=-0.05$, $w_2=-0.2$, $w_3=-0.05$, and $w_d=0$ for $d\geq 4$, all constraints are satisfied and $\mu(G)\leq m/5$ for each graph G.

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Further Reading

- Chapter 2, Branching in Fedor V. Fomin and Dieter Kratsch. Exact Exponential Algorithms. Springer, 2010.
- Chapter 6, Measure & Conquer in Fedor V. Fomin and Dieter Kratsch. Exact Exponential Algorithms. Springer, 2010.
- Chapter 2, Branching Algorithms in Serge Gaspers. Exponential Time Algorithms: Structures, Measures, and Bounds. VDM Verlag Dr. Mueller, 2010.