# Fixed parameter and exponential algorithms Journée COMRED, 01/03/2010

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## Context

Many (combinatorial) optimisation problems are NP-hard = no polynomial-time algorithms to solve them.

#### WE NEED TO SOLVE THEM.

Several classical approaches:

- approximation algorithms;
- randomised algorithms;
- heuristics.

Drawback: do not give the optimal solution.

#### WE NEED EXACT ALGORITHMS.



## Two (new) approaches

Exact exponential time algorithm: running time in  $O(c^n)$  with c as small as possible.  $\Rightarrow$  if c is small, one can solve large size instances.

Fixed parameter algorithm: running time in f(k)P(n) with

- k parameter (well chosen),
- f function,
- P polynomial.
- $\Rightarrow$  if k is small, one can solve.



## Optimisation problem and parameterisation

#### NP minimisation problem Instance: $x \in \Sigma^*$ with $\Sigma$ a finite alphabet. Goal: Find min{ $cost(x, y) | y \in sol(x)$ } with • sol(x) set of solutions of x;

• 
$$cost : \{(x, y) \mid y \in sol(x)\}.$$

#### Associated parameterized problem Instance: $x \in \Sigma^*$ and an integer k. Parameter: k. Question: $\min{cost(x, y) | y \in sol(x)} \le k$ ?



## Example 1: Vertex cover

Vertex cover = set of vertices C such that every edge has an endvertex in C.



#### Minimum Vertex Cover Problem:

Input: Graph G.

Output: A vertex cover of G of minimum size.

#### Parameterised Vertex Cover Problem:

Input: Graph G and integer k.

Parameter: k.

Question: Does G have a vertex cover of size (at most) k ?



## Example 2: Maximum Independent Set

independent set = set of pairwise non-adjacent vertices.



#### Maximum Independent Set Problem:

Input: Graph G.

Output: An independent set of maximum size in G.

Parameterised Independent Set Problem:

Input: Graph G and integer k.

Parameter: k.

Question: Does G have an independent of size (at least) k ?



## Example 3: Minimum Dominating Set

dominating set = set D of vertices such that  $D \cup N(D) = V(G)$ .



#### Minimum Dominating Set Problem:

Input: Graph G.

Output: A dominating set of minimum size in G.

#### Parameterised Dominating Set Problem:

Input: Graph G and integer k.

Parameter: k.

Question: Does G have a dominating set of size (at most) k ?



Example 4: Chromatic Number

colouring = c:  $V(G) \rightarrow S$  s. t.  $c(u) \neq c(v)$ ,  $\forall uv \in E(G)$ .



#### Chromatic Number Problem:

Input: Graph G.

Output: A colouring with minimum number of colours G.

#### k-Colourability Problem:

Input: Graph G and integer k.

Parameter: k.

Question: Is G k-colourable?  $(\chi(G) \le k?)$ 



## **FPT** problems

A parameterized problem is Fixed Parameter Tractable if it is decidable in time  $f(k)n^c$ .

FPT implies polynomial-time solvable for any fixed k.

Vertex Cover, Independent Set, Dominating Set: OK Trivial algorithm in  $O(n^k)$ .

Chromatic number is not FPT since 3-colourability is NP-complete.



Parameterized complexity theory

$$\mathsf{P} \subseteq \mathsf{FPT} \subseteq \underbrace{\mathsf{W}[1] \subseteq \mathsf{W}[2] \subseteq \ldots \subseteq \mathsf{W}[\mathsf{P}] \subseteq \mathsf{XP}}_{\mathsf{presumably fixed-parameter intractable}}$$

Conjecture FPT  $\neq$  W[1], and more generally W[i]  $\neq$  W[i + 1]

$$\mathsf{P} = \mathsf{NP} \Rightarrow \mathsf{FPT} = \mathsf{W[1]}$$

but the converse seems not to hold.

Examples: Vertex Cover is FPT, Independent Set is W[1], Dominating Set is W[2]



## Difference between W[1] and W[2]

Independent Set:  $\exists (x_1, \dots, x_k), \forall i \neq j, x_i x_j \notin E$ Dominating Set:  $\exists (x_1, \dots, x_k), \forall v \in V, \exists i, x_i v \in E \text{ or } x_i = v$ 

#### One more level of quantifiers for Dominating Set.



## Algorithmic methods for fixed parameter algorithms

- Data reduction and problems kernels
- Depth-bounded search trees
- Color Coding
- Iterative compression
- Tree decomposition, minor theory



Data reduction and problems kernels

Observation: If a vertex is incident to more than k edges, it must be in every vertex cover of size at most k.

Buss' reduction for Vertex Cover: All vertices with degree > k are added to the vertex cover.

In the resulting graph G' each vertex has degree at most k. Then iG' has a vertex cover of size  $k' \leq k$ , then it contains at most  $k^2 + k$  vertices and at most  $k^2$  edges.

Brute force: check the  $\binom{k^2+k}{k'}$  possibilities  $\Rightarrow$  algo in time g(k) + n.



## Kernelization

f(k)-kernelization: polynomial-time algorithm instance  $(G, k) \longrightarrow$  instance (G', k') such that:

- (G', k') is equivalent to (G, k);
- $k' \leq k$  and  $|G'| \leq f(k)$ .

Kernelization + brute force =  $O(g(k) + n^c)$  time algo.

Theorem: A parameterized problem is Fixed Parameter Tractable if and only if it has a kernelization.



## Kernel race

We want kernel but also as small as possible kernels.

Techniques for finding small kernels:

- Integer Linear Programming,
- Crown decomposition,

• ...

Techniques for non-existence of small kernel (under complexity assumptions):

- Distillation,
- Coloured strengthening,

• ...



## ILP-formulation of Vertex Cover

The relaxation has an half-integral solution, i.e.  $x_v \in \{0, 1/2, 1\}$ .

For 
$$t \in \{0, 1/2, 1\}$$
, set  $V_t = \{v \in V \mid x_v = t\}$  and  $G_t = G\langle V_t \rangle$ .

Obs:  $vc(G_{1/2}) \ge \frac{1}{2}|V_{1/2}|$ .

[Nemhauser et Trotter '75] There is a minimum vertex cover C of G such that:  $V_0 \cap C = \emptyset$  and  $V_1 \subseteq C$ .

## LP-based kernelization of Vertex Cover

- 1. Find an optimal solution x of the fractional relaxation.
- 2. If the weight of x is greater than k, return a "no"-instance.
- 3. Else return  $(G_{1/2}, k |V_1|)$ .

By the Observation,  $|G_{1/2}| \le 2(k - |V_1|) \le 2k$ .

So, we have a 2k-kernel.



### Depth-bounded search trees

Idea: In polynomial time find a small subset s. t. at least one element of this subset is part of an optimal solution.

Vertex Cover: small subset = two endvertices of an edge.  $\Rightarrow$  binary tree of depth at most k. Time 2<sup>k</sup>.n.



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Hardness results for Maximum Independent Set

- NP-hard [Karp '72]
- Not approximable within  $O(n^{1-\epsilon})$  unless P = NP[Zucherman '06]
- No exact O(c<sup>o(n)</sup>) algorithm unless SNP ⊆ SUBEXP [Impagliazzo,Paturi,Zane '01]
- W[1]-hard [Downey & Fellows '92]

⇒ The best we can hope for is a  $O(c^n)$  exact algorithm for some small constant  $c \in ]1, 2]$ .

## Race for Maximum Independent Set

- $O(1.261^n)$  poly-space [Tarjan & Trojanowski '77]
- *O*(1.235<sup>*n*</sup>) poly-space [Jian '86]
- *O*(1.228<sup>*n*</sup>) poly-space, *O*(1.221<sup>*n*</sup>) exp-space [Robson '86]
- better results for sparse graphs [Beigel99, Chen,Kanj & Xia '03]
- simpler O(1.221<sup>n</sup>) exp-space [Fomin, Grandoni and Kratsch '06]

New techniques for exact exponential algorithms

Design techniques: Branch and Recharge, Inclusion-Exclusion, ...

Running time analysis techniques: Measure and Conquer, ...



Reduction rules for Maximum Independent Setcomponents: C connected component  $\alpha(G) = \alpha(C) + \alpha(G - C)$ .dominance:if  $N[w] \subset N[v]$  $\alpha(G) = \alpha(G - v)$ .folding:if v has 2 neighbours v and w $\alpha(G) = 1 + \alpha(\tilde{G})$ .



mirroring: M set of mirrors of v  $\alpha(G) = \max\{\alpha(G - v - M), 1 + \alpha(G - N[v])\}$ u is a mirror of v if d(u, v) = 2 and  $N(v) \setminus N(u)$  is a clique.



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Algorithm for Maximum Independent Set

mis(G)

- 1. if  $|V(G)| \leq 1$  return |V(G)|;
- 2. **if**  $\exists$  component *C* of *G* **return**  $\min(C) + \min(G C)$ ;
- 3. **if**  $\exists$  v and w s.t.  $N[w] \subset N[v]$  **return** mis(G v);
- 4. **if**  $\exists$  v s.t. d(v) = 2 **return**  $1 + \min(\tilde{G})$ ;
- 5. pick a vertex v of max. degree;
- 6. return max{mis(G v M), 1 + mis(G N[v])}

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FPT and exponential algo

## Standard analysis

1.5 & 6 Branching Step

d(v) = 3: when discarding v, we also discard a mirror of v, or at next step we fold a neighbour of v, so we remove at least 2 vertices; when selecting we remove at least 4 vertices.  $\Rightarrow B(n) \le B(n-2) + B(n-4)$ .

 $d(v) \ge 4$ : we remove at least one or five vertices  $\Rightarrow B(n) \le B(n-1) + B(n-5)$ .

 $B(n) = O^*(\lambda^n)$  with  $\lambda = 1.3247$  the largest root of  $x^4 - x^2 - 1$ and  $x^5 - x^4 - 1$ .

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## Measure and conquer

In the standard analysis, measure of a graph:  $\mu(G) =$  number of vertices.

Idea: take into account the fact that reducing the degree has a positive impact on the long term. (degree 2 vertices are removed without branching)

New measure:  $n_i$ : number of vertices of degree *i*.  $\alpha_i$ : weight (in [0, 1]) of every vertex of degree *i*.

$$\mu'(G) = \sum_{i \ge 0} \alpha_i n_i$$

We obtain new recurrence relations and by choosing adequate value of  $\alpha_i$ , we get  $B(n) = O^*(1.221^n)$ .

