# An Introduction to Parameterized Complexity

Lecture 1: Fixed-Parameter Tractability

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Ph.D. Program, Advanced Period Gran Sasso Science Institute L'Aquila, Italy

Monday, June 10, 2024

## Course overview

Monday, June 10 10.30 – 12.30	Tuesday, June 11	Wednesday, June 12 10.30 – 12.30	Thursday, June 13	Friday, June 14
Introduction & basic FPT results		Algorithmic Meta-Theorems		
motivation for FPT kernelization, Crown Lemma, Sunflower Lemma	GDA	1st-order logic, monadic 2nd-order logic, FPT-results by Courcelle's Theorems for tree and clique-width	GDA	GDA
Algorithmic Techniques		Formal-Method & Algorithmic Techniques		
	14.30 - 16.30			14.30 - 16.30
	Notions of bounded graph width			FPT-Intractability Classes & Hierarchies
	path-, tree-, clique width, FPT-results by dynamic programming, transferring FPT results betw. widths	GDA	GDA	motivation for FP-intractability results, FPT-reductions, class XP (slicewise polynomial), W- and A-Hierarchies, placing problems on these hierarchies

## Course developers



Hugo Gilbert course 2019/20 (Hugo & Clemens)



CG & Alessandro Aloisio course 2020/21 (Alessandro & C)

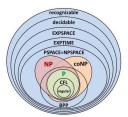
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## Motivation

### Classical complexity theory

- analyses problems by resource (space or time)
   needed to solve them on a reasonable machine model
- ▶ as a function of the input size n = |x| (Hartmanis/Stearns, 1965)
- ⇒ variety of complexity classes (P, LOGSPACE, NP, PSPACE, ...)
- ⇒ tractable problems
  - = polynomial-time computable (in P)
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#### Drawback

- measures problem size n = |x|
   only in terms of input instances x,
   and ignores structural information about instances
- sometimes problems are easier to solve for instances if additional structure information is available

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### Parameterized complexity

- measures complexity also in terms of a parameter  $k = \kappa(x)$  that may depend on the input x in an arbitrary way
- $\Rightarrow$  fixed-parameter tractable problems relaxes polynomial time solvability to algorithms whose non-polynomial behavior  $f(k) \cdot p(n)$  is restricted by parameter k
- ⇒ complexity classes (FPT, XP, W[P], W- and A-hierarchies)
- ⇒ theory of fixed-parameter intractability

# Parameterized (versus classical) problems

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#### Definition

The size of an instance  $\langle x, \kappa(x) \rangle$  of  $\langle Q, \kappa \rangle$  is

$$|\langle x, \kappa(x) \rangle| = |x| + \kappa(x)$$
.

# Parameterized problems (examples)

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**Given:** a graph G and an integer k,

**Question:** Does there exists a clique of size k in G?

Parameter: k.

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**Given:** a universe  $U = \{x_1, \dots, x_n\}$ , a collection of sets  $S = (S_1, \dots, S_m)$  where  $S_i \subseteq U$  and an integer k,

**Question:** Does there exists a set  $S \subseteq U$  such that  $|S| \le k$  and  $S \cap S_i \ne \emptyset$ ,  $\forall i \in \{1, ..., m\}$ .

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- ▶ NP-hard even if  $\max |S_i| = 2$ ,
- ▶ is fixed-parameter tractable.

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There is a hierarchy on parameters.

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There are many different types of parameters!

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- Some more structural property of the instance.
   E.g., the diameter of a graph.
- It can be a combination of values, a difference, ...

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- Social choice problems: number of voters, candidates, correlation of preferences...
- ▶ Boolean formulas: number of variables, number of clauses...
- Problems on strings: maximum length of a string, size of the alphabet...

## Fixed Parameter Tractability (Class FPT)

#### Definition

A parameterized problem  $(Q, \kappa)$  is *fixed-parameter tractable* if:

```
\exists f: \mathbb{N} \to \mathbb{N} \text{ computable } \exists p \in \mathbb{N}[X] \text{ polynomial} \\ \exists \mathbb{A} \text{ algorithm, takes inputs in } \Sigma^* \text{ and } \forall x \in \Sigma^* \\ \left[ \mathbb{A} \text{ decides if } x \in Q \text{ in time } \leq f(\kappa(x)) \cdot p(|x|) \right].
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### Assumption for a robust fpt-theory:

 $\kappa$  is polynomially computable, or itself fpt-computable.

### Goal in parameterized algorithmics:

- ⇒ design FPT algorithms,
- $\Rightarrow$  try to make both factors  $f(\kappa(x))$  and p(|x|) as small as possible.
- ⇒ or show (if possible) that finding such factors is impossible

# Slices of FPT problems are in P

The  $\ell$ -th slice of a parameterized problem  $(Q, \kappa)$ :

$$\langle Q, \kappa \rangle_{\ell} \coloneqq \{ x \in Q \mid \kappa(x) = \ell \}$$
 (as classical problem).

### Proposition

If  $\langle Q, \kappa \rangle \in \mathsf{FPT}$ , then  $\langle Q, \kappa \rangle_{\ell} \in \mathsf{P}$  for all  $\ell \in \mathbb{N}$ .

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# A problem not in FPT (unless P = NP)

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### Application

### p-Colorability

**Instance:** a graph  $\mathcal{G}$  and  $k \in \mathbb{N}$ .

Parameter: *k*.

**Problem:** Decide whether G is k-colorable.

Known: 3-COLORABILITY ∈ NP-complete (Lovàsz, Stockmeyer, 1973).

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Since 3-Colorability = p-Colorability<sub>3</sub>,

it follows that p-Colorability  $\notin$  FPT (unless P = NP).

# Slice-wise polynomial problems (Class XP)

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XP := complexity class of slice-wise polynomial problems.

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it follows that p-Colorability  $\notin XP$  (unless P = NP).

### Aims of the course

- Acquire a basic notions of parameterized complexity.
- Obtain an introduction to some techniques to derive FPT or XP results.
- Obtain an introduction to a variety of techniques to prove algorithmic lower bounds and in particular prove parameterized hardness results.

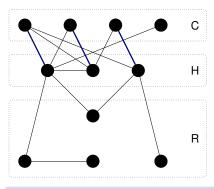
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				THOTALOTHES

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### From today's lecture



A **crown decomposition** of a graph G is a partitioning (C, H, R) of V(G), such that:

- C is nonempty.
- ② C is an independent set.
- $\bullet$  H separates C and R.
- 4 *G* contains a matching of *H* into *C*.

### Lemma (Crown lemma.)

Let G be a graph with no isolated vertices and with at least 3k + 1 vertices. There is a polynomial-time algorithm that:

- ▶ either finds a matching of size k + 1 in G;
- or finds a crown decomposition of G.

### **Tomorrow**

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## In tomorrow's lecture: a path decomposition of a graph



# Wednesday

Monday, June 10 10.30 – 12.30	Tuesday, June 11	Wednesday, June 12 10.30 – 12.30	Thursday, June 13	Friday, June 14
Introduction & basic FPT results motivation for FPT kernelization,		Algorithmic Meta-Theorems 1st-order logic, monadic 2nd-order		
Crown Lemma, Sunflower Lemma		logic, FPT-results by Courcelle's Theorems for tree and clique-width		
Algorithmic	Techniques	Formal-Method & Algorithmic Techniques		
	14.30 – 16.30  Notions of bounded graph width path-, tree-, clique width, FPT-results by dynamic programming, transferring FPT results betw. widths			14.30 – 16.30  FPT-Intractability Classes & Hierarchies motivation for FP-intractability results, FPT-reductions, class XP (slicewise polynomial), W- and A-Hierarchies, placing problems on these hierarchies

# In Wednesday's lecture: Monadic second-order logic

$$\psi_{\mathbf{3}} := \exists C_{\mathbf{1}} \exists C_{\mathbf{2}} \exists C_{\mathbf{3}} \big( \big( \forall x \bigvee_{i=1}^{3} C_{i}(x) \big) \\ \land \forall x \forall y \big( E(x,y) \to \bigwedge_{i=1}^{3} \neg \big( C_{i}(x) \land C_{i}(y) \big) \big) \big)$$

$$\mathcal{A}(\mathcal{G}) \vDash \psi_{\mathbf{3}} \iff \mathcal{G} \text{ has is 3-colorable}.$$

# Friday

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Introduction & basic FPT results motivation for FPT kernelization, Crown Lemma, Sunflower Lemma		Algorithmic Meta-Theorems 1st-order logic, monadic 2nd-order logic, FPT-results by Courcelle's Theorems for tree and clique-width		
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	14.30 – 16.30			14.30 – 16.30
	Notions of bounded			FPT-Intractability
	graph width			Classes & Hierarchies
	path-, tree-, clique width, FPT-results by dynamic programming, transferring FPT results betw. widths			motivation for FP-intractability results, FPT-reductions, class XP (slicewise polynomial), W- and A-Hierarchies, placing problems on these hierarchies

# From Friday's lecture: W-Hierarchy

'There is no definite single class that can be viewed as "the parameterized NP". Rather, there is a whole hierarchy of classes playing this role. (Flum, Grohe [FG06])



### Course overview

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Introduction & basic FPT results motivation for FPT		Algorithmic Meta-Theorems 1st-order logic,		
kernelization,		monadic 2nd-order logic, FPT-results by		
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	path-, tree-, clique			motivation for
	width, FPT-results			FP-intractability results,
	by dynamic			FPT-reductions, class
	programming,			XP (slicewise
	transferring FPT			polynomial), W- and
	results betw. widths			A-Hierarchies, placing
				problems on these hierarchies
				TiloraiGilles

### **Books**





- Marek Cygan, Fedor V. Fomin, Lukasz Kowalik, Daniel Lokshtanov, Daniel Marx, Marcin Pilipczuk, Michal Pilipczuk, and Saket Saurabh, *Parameterized Algorithms*, 1st ed., Springer, 2015.
- Jörg Flum and Martin Grohe, *Parameterized Complexity Theory*, Springer, 2006.

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- Sunflower lemma
  - kernel for hitting set problem

# Kernelization (formally)

#### Definition

Let  $\langle Q, \kappa \rangle$  be a parameterized problem over  $\Sigma$ .

A *kernelization* of  $(Q, \kappa)$  is a function  $K: \Sigma^* \to \Sigma^*$  such that:

- ▶ *K* is polynomial-time computable
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# The (parameterized) Point Line Cover Problem

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**Parameter:** The integer k.

**Question:** Do there exist k lines that cover all points?

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- i) include it in the solution;
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Observation: Let  $(x, \kappa)$  be a yes instance of the p-Point-Line-Cover such that Rule 1 cannot be applied. Then  $n \le k^2$  holds.

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### Proposition

p-POINT-LINE-COVER  $\in$  FPT: it admits a kernel of size with  $k^2$  points.

# The (parameterized) Vertex Cover Problem

### p-VERTEX-COVER:

Given: A graph G.

**Parameter**: The integer k.

**Question:** Does there exists a vertex cover of size at most k?

### Definition

Let G be a graph and  $S \subseteq V(G)$ . The set S is called a vertex cover if for every edge of G at least one of its endpoints is in S.

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### Exercise

Find an  $O(k^2)$  kernel for p-VERTEX-COVER.

# The (parameterized) Vertex Cover Problem (Samuel Buss)

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- **Rule 3**: Let (G,k) be an instance to which Rules 1 & 2 are not applicable. If G has  $> k^2 + k$  vertices, or  $> k^2$  edges, then (G,k) is a no-instance that can be replaced by a trivial no-instance.

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### Theorem

p-VERTEX-COVER  $\in$  FPT, because it admits a kernel with at most  $O(k^2)$  vertices and  $O(k^2)$  edges.

## Kernelization ⇒ FPT

### Exercise

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A parameterized problem  $\langle Q, \kappa \rangle$  is *fixed-parameter tractable* if:

```
\exists f: \mathbb{N} \to \mathbb{N} \text{ computable } \exists p \in \mathbb{N}[X] \text{ polynomial} \\ \exists \mathbb{A} \text{ algorithm, takes inputs in } \Sigma^* \text{ and } \forall x \in \Sigma^* \\ \left[ \mathbb{A} \text{ decides if } x \in Q \text{ in time } \leq f(\kappa(x)) \cdot p(|x|) \right].
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FPT := complexity class of all fixed-parameter tractable problems.

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```
(Q,K) a parameterized problem, Q < 2*
 Definition K: Z* > Z* a kernelization for (Q, K) if:
    (K1) YXE>* (XEQ (XK)EQ)
      Ka) K is polytime computable
      M3) ∃n: N→N Yx∈ Z*( | K(x)| ≤ L( k(x))).
Proposition: If <0,187 is decidable, and has kernelization K, then (Q,18) EFPT
Proof. Since < Q K) is decidable, there is an algorithm A) that decides instances xet in time = f(1x1) steps for some Computable function f: N > N.
Then assuming a polynomial algorialum Ax for k (time bounded by F(x))
  we construct on PPT algorishm Al(K) for
                                         K(x) E = * | Ruming Lime A(K) =
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```

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We define the polynomial-time computable function  $K: \Sigma^* \to \Sigma^*$  by:

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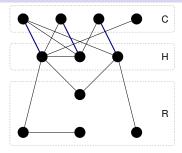
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$$K(x) \coloneqq \begin{cases} x_0 & \dots \text{ } \mathbb{A} \text{ accepts } x \text{ in } \leq p(|x|) \cdot p(|x|) \text{ steps,} \\ x_1 & \dots \text{ } \mathbb{A} \text{ rejects } x \text{ in } \leq p(|x|) \cdot p(|x|) \text{ steps,} \\ x & \dots \text{ } \mathbb{A} \text{ does not terminate in } \leq p(|x|) \cdot p(|x|) \text{ steps.} \end{cases}$$

In the last case (K(x) = x) we have  $p(|x|) \cdot p(|x|) \le f(\kappa(x)) \cdot p(|x|)$ , and hence  $|K(x)| = |x| \le p(|x|) \le f(\kappa(x))$ . Therefore K is a kernel.

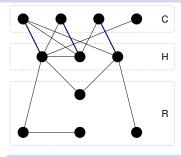
# Crown Decomposition and Crown Lemma



A **crown decomposition** of a graph G is a partitioning (C, H, R) of V(G), such that:

- C is nonempty.
- 2 C is an independent set.
- $\odot$  H separates C and R.
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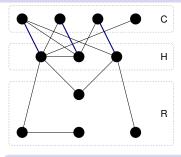
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### Lemma (Crown Lemma)

Let G be a graph with no isolated vertices and with at least 3k + 1 vertices. There is a polynomial-time algorithm that:

- either finds a matching of size k + 1 in G;
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#### Exercise

Apply the Crown Lemma to the Vertex Cover Problem.

# The (par.) Vertex Cover Problem (smaller kernel)

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Given: A graph G.

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- ▶ If it returns a matching of size k + 1, then conclude that (G, k) is a no-instance
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### **Theorem**

p-VERTEX-COVER admits a kernel with at most 3k vertices.

# The (parameterized) Dual-Coloring Problem

### p-COLORABILITY:

**Given:** A graph  $G = \langle V, E \rangle$  on n vertices and an integer k.

Parameter: The integer k. Question: Is G k-colorable?

#### Definition

Let  $k \in \mathbb{N}$ . A graph  $G = \langle V, E \rangle$  is k-colorable if there is a function  $C : V \to \{1, \dots, k\}$  such that  $C(u) \neq C(v)$  for all edges  $\{u, v\} \in E$ .

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### Exercise

Obtain a kernel with O(k) vertices using crown decomposition.

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#### **Theorem**

p-DUAL-COLORING admits a kernel with at most 3k vertices.

### Sunflower Lemma

#### Definition

A sunflower with k petals and a core Y is a collection of sets  $S_1, \ldots, S_k$  such that  $S_i \cap S_j = Y$  for all  $i \neq j$ . The sets  $S_i \setminus Y$  are petals and they must be non-empty.

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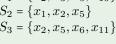
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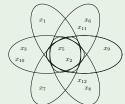
A sunflower with 6 petals and a core  $Y = \{x_2, x_5\}.$ 

$$S_1 = \{x_2, x_3, x_5, x_{10}\}$$

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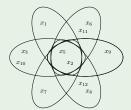
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### Lemma (Sunflower lemma (Erdős, Rado))

Let A be a family of sets (without duplicates) over a universe U such that each set in A has cardinality = d.

If  $|\mathcal{A}| > d!(k-1)^d$ , then  $\mathcal{A}$  contains a sunflower with k petals which can be computed in time polynomial in  $|\mathcal{A}|$ , |U|, and k.

## Application to *d*-Hitting Set

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### p-d-HITTING-SET:

**Given:** A family  $\mathcal{A}$  of sets over a universe U, where each set has cardinality  $\leq d$  and a positive integer k,

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**Question:** Does there exists a subset  $H \subseteq U$  of size at most

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#### Exercise

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#### **Theorem**

p-d-HITTING-SET has a kernel with  $\leq d!k^dd$  sets  $\& \leq d!k^dd^2$  elements.

# Application to *d*-Hitting Set

#### Observation

If  $\mathcal A$  contains a sunflower  $\mathcal S=\{S_1,\ldots,S_{k+1}\}$  of k+1 sets, then every hitting set H of  $\mathcal A$  with  $|H|\leq k$  must intersect the core Y of  $\mathcal S$ . Otherwise it is a no-instance, because H cannot intersect each of the k+1 petals  $S_i \smallsetminus Y$ .

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Proof (kernel of p-d-HITTING-SET with  $\leq d!k^dd$  sets and  $\leq d!k^dd^2$  elements).

If for some  $d' \in \{1, ..., d\}$ , the number of sets in  $\mathcal{A}$  of size = d' is more than  $d'!k^{d'}$ , then the sunflower lemma yields a sunflower of size k + 1.

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### **Tomorrow**

Monday, June 10 10.30 – 12.30	Tuesday, June 11	Wednesday, June 12 10.30 – 12.30	Thursday, June 13	Friday, June 14
Introduction & basic FPT results		Algorithmic Meta-Theorems		
motivation for FPT kernelization, Crown Lemma, Sunflower Lemma	GDA	1st-order logic, monadic 2nd-order logic, FPT-results by Courcelle's Theorems for tree and clique-width	GDA	GDA
Algorithmic Techniques		Formal-Method & Algorithmic Techniques		
	14.30 - 16.30 Notions of bounded graph width			14.30 – 16.30  FPT-Intractability  Classes & Hierarchies
	path-, tree-, clique width, FPT-results by dynamic programming, transferring FPT results betw. widths	GDA	GDA	motivation for FP-intractability results, FPT-reductions, class XP (slicewise polynomial), W- and A-Hierarchies, placing problems on these hierarchies