8. Parameterized intractability: the W-hierarchy COMP6741: Parameterized and Exact Computation

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Outline

- Reminder: Polynomial Time Reductions and NP-completeness
- Parameterized Complexity Theory
 - Parameterized reductions
 - Parameterized complexity classes
- Case studies
- Further Reading

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Polynomial-time reduction

Definition 1

A polynomial-time reduction from a decision problem Π_1 to a decision problem Π_2 is a polynomial-time algorithm, which, for any instance of Π_1 produces an equivalent instance of Π_2 .

If there exists a polynomial-time reduction from Π_1 to Π_2 , we say that Π_1 is polynomial-time reducible to Π_2 and write $\Pi_1 \leq_P \Pi_2$.

New polynomial-time algorithms via reductions

Lemma 2

If Π_1,Π_2 are decision problems such that $\Pi_1\leq_{\mathsf{P}}\Pi_2$, then $\Pi_2\in P$ implies $\Pi_1\in P$.

NP-completeness

Definition 3 (NP-hard)

A decision problem Π is NP-hard if $\Pi' \leq_P \Pi$ for every $\Pi' \in NP$.

Definition 4 (NP-complete)

A decision problem Π is NP-complete (in NPC) if

- \bullet $\Pi \in \mathsf{NP}$, and

Proving NP-completeness

Lemma 5

If Π is a decision problem such that $\Pi' \leq_P \Pi$ for some NP-hard decision problem Π' , then Π is NP-hard.

If, in addition, $\Pi \in \mathsf{NP}$, then $\Pi \in \mathsf{NPC}$.

Proving NP-completeness II

Method to prove that a decision problem Π is NP-complete:

- Prove $\Pi \in NP$
- extstyle ext
 - Select a known NP-hard decision problem Π' .
 - Describe an algorithm that transforms every instance I of Π' to an instance r(I) of Π .
 - Prove that for each instance I of Π' , we have that I is a YES-instance of Π' $\Leftrightarrow r(I)$ is a YES-instance of Π .
 - Show that the algorithm runs in polynomial time.

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Main Parameterized Complexity Classes

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n: instance sizek: parameter
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P: class of problems that can be solved in $n^{O(1)}$ time

FPT: class of parameterized problems that can be solved in $f(k) \cdot n^{O(1)}$ time

W[·]: parameterized intractability classes

XP: class of parameterized problems that can be solved in $f(k) \cdot n^{g(k)}$ time ("polynomial when k is a constant")

$$P \subseteq FPT \subseteq W[1] \subseteq W[2] \cdots \subseteq W[P] \subseteq XP$$

Note: We assume that f is computable and non-decreasing.

Polynomial-time reductions for parameterized problems?

A vertex cover in a graph G=(V,E) is a subset of vertices $S\subseteq V$ such that every edge of G has an endpoint in S.

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Vertex Cover
```

Input: Graph G, integer k

Parameter: *k*

Question: Does G have a vertex cover of size k?

An independent set in a graph G=(V,E) is a subset of vertices $S\subseteq V$ such that there is no edge $uv\in E$ with $u,v\in S$.

INDEPENDENT SET

Input: Graph G, integer k

Parameter: *k*

Question: Does G have an independent set of size k?

Polynomial-time reductions for parameterized problems?

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Vertex Cover
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Question: Does G have a vertex cover of size k?

An independent set in a graph G=(V,E) is a subset of vertices $S\subseteq V$ such that there is no edge $uv\in E$ with $u,v\in S$.

INDEPENDENT SET

Input: Graph G, integer k

Parameter: *k*

Question: Does G have an independent set of size k?

- We know: Independent Set ≤_P Vertex Cover
- However: VERTEX COVER ∈ FPT but INDEPENDENT SET is not known to be in FPT

11 / 34

We will need another type of reductions

• Issue with polynomial-time reductions: parameter can change arbitrarily

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- Issue with polynomial-time reductions: parameter can change arbitrarily
- We will want the reduction to produce an instance where the parameter is bounded by a function of the original instance
- Also: we can allow the reduction to take FPT time instead of only polynomial time.

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Parameterized reduction

Definition 6

A parameterized reduction from a parameterized decision problem Π_1 to a parameterized decision problem Π_2 is an algorithm, which, for any instance I of Π_1 with parameter k produces an instance I' of Π_2 with parameter k' such that

- I is a YES-instance for $\Pi_1 \Leftrightarrow I'$ is a YES-instance for Π_2 ,
- there exists a computable function g such that $k' \leq g(k)$, and
- there exists a computable function f such that the running time of the algorithm is $f(k) \cdot |I|^{O(1)}$.

If there exists a parameterized reduction from Π_1 to Π_2 , we write $\Pi_1 \leq_{\mathsf{FPT}} \Pi_2$.

Note: We can assume that f and g are non-decreasing.

New FPT algorithms via reductions

Lemma 7

If Π_1, Π_2 are parameterized decision problems such that $\Pi_1 \leq_{\mathsf{FPT}} \Pi_2$, then $\Pi_2 \in \mathsf{FPT}$ implies $\Pi_1 \in \mathsf{FPT}$.

Proof sketch.

To obtain an FPTalgorithm for Π_1 , perform the reduction and then use an FPTalgorithm for Π_2 on the resulting instance.



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Boolean Circuits

Definition 8

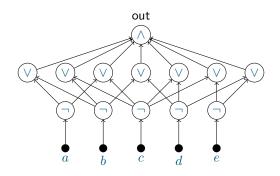
A Boolean circuit is a directed acyclic graph with the nodes labeled as follows:

- every node of in-degree 0 is an input node,
- every node with in-degree 1 is a negation node (\neg) , and
- ullet every node with in-degree ≥ 2 is either an AND-node (\wedge) or an OR-node (\vee).

Moreover, exactly one node with out-degree 0 is also labeled the output node. The depth of the circuit is the maximum length of a directed path from an input node to the output node.

The weft of the circuit is the maximum number of nodes with in-degree ≥ 3 on a directed path from an input node to the output node.

Example



A depth-3, weft-1 Boolean circuit with inputs a,b,c,d,e.

Weighted Circuit Satisfiability

Given an assignment of Boolean values to the input gates, the circuit determines Boolean values at each node in the obvious way.

If the value of the output node is 1 for an input assignment, we say that this assignment satisfies the circuit.

The weight of an assignment is its number of 1s.

WEIGHTED CIRCUIT SATISFIABILITY (WCS)

Input: A Boolean circuit C, an integer k

Parameter: k

Question: Is there an assignment with weight k that satisfies C?

Exercise: Show that WEIGHTED CIRCUIT SATISFIABILITY ∈ XP.

WCS for special circuits

Definition 9

The class of circuits $C_{t,d}$ contains the circuits with weft $\leq t$ and depth $\leq d$.

For any class of circuits C, we can define the following problem.

WCS[C]

Input: A Boolean circuit $C \in \mathcal{C}$, an integer k

Parameter: k

Question: Is there an assignment with weight k that satisfies C?

W classes

Definition 10 (W-hierarchy)

Let $t \in \{1,2,\ldots\}$. A parameterized problem Π is in the parameterized complexity class $\mathbf{W}[t]$ if there exists a parameterized reduction from Π to $\mathrm{WCS}[\mathcal{C}_{t,d}]$ for some constant d > 1.

Independent Set and Dominating Set

Theorem 11

Independent Set $\in W[1]$.

Theorem 12

Dominating Set $\in W[2]$.

Recall: A dominating set of a graph G=(V,E) is a set of vertices $S\subseteq V$ such that $N_G[S]=V$.

Dominating Set

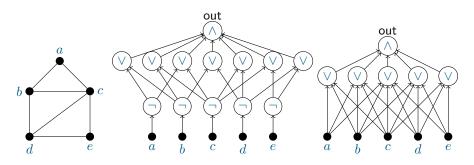
Input: A graph G = (V, E) and an integer k

Parameter: k

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

"Proof" by picture

Parameterized reductions from Independent Set to WCS[$\mathcal{C}_{1,3}$] and from Dominating Set to WCS[$\mathcal{C}_{2,2}$].



Setting an input node to 1 corresponds to adding the corresponding vertex to the independent set / dominating set.

W-hardness

Definition 13

Let $t \in \{1, 2, \dots\}$.

A parameterized decision problem Π is W[t]-hard if for every parameterized decision problem Π' in W[t], there is a parameterized reduction from Π' to Π . Π is W[t]-complete if $\Pi \in W[t]$ and Π is W[t]-hard.

It has been proved that INDEPENDENT SET is W[1]-hard and DOMINATING SET is W[2]-hard. Therefore,

Theorem 14

INDEPENDENT SET is W[1]-complete.

Theorem 15

DOMINATING SET is W[2]-complete.

Proving W-hardness

To show that a parameterized decision problem Π is W[t]-hard:

- Select a W[t]-hard problem Π'
- Show that $\Pi' \leq_{\mathsf{FPT}} \Pi$ by designing a parameterized reduction from Π' to Π
 - Design an algorithm, that, for any instance I' of Π' with parameter k', produces an equivalent instance I of Π with parameter k
 - Show that k is upper bounded by a function of k'
 - Show that there exists a function f such that the running time of the algorithm is $f(k') \cdot |I'|^{O(1)}$

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Clique

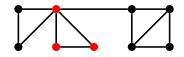
A clique in a graph G=(V,E) is a subset of its vertices $S\subseteq V$ such that every two vertices from S are adjacent in G.

CLIQUE

Input: Graph G = (V, E), integer k

Parameter: k

Question: Does G have a clique of size k?



• We will show that CLIQUE is W[1]-hard by a parameterized reduction from INDEPENDENT SET.

Lemma 16

INDEPENDENT SET \leq_{FPT} CLIQUE.

Proof.

Given any instance (G=(V,E),k) for INDEPENDENT SET, we need to describe an FPT algorithm that constructs an equivalent instance (G',k') for CLIQUE such that $k' \leq g(k)$ for some computable function g.

Lemma 16

Independent Set \leq_{FPT} Clique.

Proof.

Given any instance (G=(V,E),k) for INDEPENDENT SET, we need to describe an FPT algorithm that constructs an equivalent instance (G',k') for CLIQUE such that $k' \leq g(k)$ for some computable function g.

Construction. Set $k' \leftarrow k$ and $G' \leftarrow \overline{G} = (V, \{uv : u, v \in V, u \neq v, uv \notin E\}).$

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Proof.

Given any instance (G=(V,E),k) for INDEPENDENT SET, we need to describe an FPT algorithm that constructs an equivalent instance (G',k') for CLIQUE such that $k' \leq g(k)$ for some computable function g.

Construction. Set $k' \leftarrow k$ and $G' \leftarrow \overline{G} = (V, \{uv : u, v \in V, u \neq v, uv \notin E\})$. **Equivalence.** We need to show that (G, k) is a YES-instance for INDEPENDENT SET if and only if (G', k') is a YES-instance for CLIQUE.

Lemma 16

INDEPENDENT SET \leq_{FPT} CLIQUE.

Proof.

Given any instance (G=(V,E),k) for INDEPENDENT SET, we need to describe an FPT algorithm that constructs an equivalent instance (G',k') for CLIQUE such that $k' \leq g(k)$ for some computable function g.

Construction. Set $k' \leftarrow k$ and $G' \leftarrow \overline{G} = (V, \{uv : u, v \in V, u \neq v, uv \notin E\})$.

Equivalence. We need to show that (G, k) is a YES-instance for INDEPENDENT SET if and only if (G', k') is a YES-instance for CLIQUE.

 (\Rightarrow) : Let S be an independent set of size k in G. For every two vertices $u,v\in S$, we have that $uv\notin E$. Therefore, $uv\in E(\overline{G})$ for every two vertices in S. We conclude that S is a clique of size k in \overline{G} .

Lemma 16

Independent Set \leq_{FPT} Clique.

Proof.

Given any instance (G=(V,E),k) for INDEPENDENT SET, we need to describe an FPT algorithm that constructs an equivalent instance (G',k') for CLIQUE such that $k' \leq g(k)$ for some computable function g.

Construction. Set $k' \leftarrow k$ and $G' \leftarrow \overline{G} = (V, \{uv : u, v \in V, u \neq v, uv \notin E\})$.

Equivalence. We need to show that (G, k) is a YES-instance for INDEPENDENT SET if and only if (G', k') is a YES-instance for CLIQUE.

- (\Rightarrow) : Let S be an independent set of size k in G. For every two vertices $u,v\in S$, we have that $uv\notin E$. Therefore, $uv\in E(\overline{G})$ for every two vertices in S. We conclude that S is a clique of size k in \overline{G} .
- (\Leftarrow) : Let S be a clique of size k in \overline{G} . By a similar argument, S is an independent set of size k in G.

Lemma 16

Independent Set \leq_{FPT} Clique.

Proof.

Given any instance (G=(V,E),k) for INDEPENDENT SET, we need to describe an FPT algorithm that constructs an equivalent instance (G',k') for CLIQUE such that $k' \leq g(k)$ for some computable function g.

Construction. Set $k' \leftarrow k$ and $G' \leftarrow \overline{G} = (V, \{uv : u, v \in V, u \neq v, uv \notin E\}).$

Equivalence. We need to show that (G, k) is a YES-instance for INDEPENDENT SET if and only if (G', k') is a YES-instance for CLIQUE.

 (\Rightarrow) : Let S be an independent set of size k in G. For every two vertices $u,v\in S$, we have that $uv\notin E$. Therefore, $uv\in E(\overline{G})$ for every two vertices in S. We conclude that S is a clique of size k in \overline{G} .

 (\Leftarrow) : Let S be a clique of size k in \overline{G} . By a similar argument, S is an independent set of size k in G.

Parameter. $k' \leq k$.

Lemma 16

INDEPENDENT SET < FPT CLIQUE.

Proof.

Given any instance (G = (V, E), k) for INDEPENDENT SET, we need to describe an FPT algorithm that constructs an equivalent instance (G', k') for CLIQUE such that $k' \leq q(k)$ for some computable function q.

Construction. Set $k' \leftarrow k$ and $G' \leftarrow \overline{G} = (V, \{uv : u, v \in V, u \neq v, uv \notin E\}).$

Equivalence. We need to show that (G, k) is a YES-instance for INDEPENDENT SET if and only if (G', k') is a YES-instance for CLIQUE.

 (\Rightarrow) : Let S be an independent set of size k in G. For every two vertices $u, v \in S$, we have that $uv \notin E$. Therefore, $uv \in E(\overline{G})$ for every two vertices in S. We conclude that S is a clique of size k in G.

 (\Leftarrow) : Let S be a clique of size k in G. By a similar argument, S is an independent set of size k in G.

Parameter. k' < k.

Running time. The construction can clearly be done in FPT time, and even in polynomial time.

Corollary 17

CLIQUE is W[1]-hard

Exercise

Recall: A k-coloring of a graph G = (V, E) is a function $f : V \to \{1, 2, ..., k\}$ assigning colors to V such that no two adjacent vertices receive the same color.

Multicolor Clique

Input: A graph G = (V, E), an integer k, and a k-coloring of G

Parameter: k

Question: Does G have a clique of size k?

• Show that MULTICOLOR CLIQUE is W[1]-hard.

Exercise

A set system S is a pair (V, H), where V is a finite set of elements and H is a set of subsets of V.

A set cover of a set system $\mathcal{S}=(V,H)$ is a subset X of H such that each element of V is contained in at least one of the sets in X, i.e., $\bigcup_{Y\in X}Y=V$.

Set Cover

Input: A set system S = (V, H) and an integer k

Parameter: k

Question: Does S have a set cover of cardinality at most k?



• Show that SET COVER is W[2]-hard.

Exercise

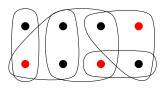
A hitting set of a set system $\mathcal{S}=(V,H)$ is a subset X of V such that X contains at least one element of each set in H, i.e., $X\cap Y\neq\emptyset$ for each $Y\in H$.

HITTING SET

Input: A set system S = (V, H) and an integer k

Parameter: k

Question: Does S have a hitting set of size at most k?



• Show that HITTING SET is W[2]-hard.

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Further Reading

- Chapter 13, Fixed-parameter Intractability in Marek Cygan, Fedor V. Fomin, Łukasz Kowalik, Daniel Lokshtanov, Dániel Marx, Marcin Pilipczuk, Michał Pilipczuk, and Saket Saurabh. Parameterized Algorithms. Springer, 2015.
- Chapter 13, Parameterized Complexity Theory in Rolf Niedermeier. Invitation to Fixed Parameter Algorithms. Oxford University Press, 2006.
- Elements of Chapters 20–23 in Rodney G. Downey and Michael R. Fellows. Fundamentals of Parameterized Complexity. Springer, 2013.