Courcelle’s Theorem for Triangulations

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In this lecture I have chosen some material which is relevant to the theme of this meeting.

Perhaps somewhat out of the comfort zone for some attendees, but I hope a nice illustration of logic “in practice.”

It involves a combination of logic, complexity theory and low dimensional topology, so I will explain each bit separately.
A mathematical idealization is to identify “Feasible” with P.

With this assumption, the theory of NP-hardness is an excellent vehicle for mapping an outer boundary of intractability, for all practical purposes.

Indeed, assuming the reasonable current working assumption that NTM acceptance is $\Omega(2^n)$, NP-hardness allows for practical lower bound for exact solution for problems.

A very difficult practical and theoretical problem is “How can we deal with P?”.

More importantly how can we deal with $P - \text{FEASIBLE}$, and map a further boundary of intractability.
When is the only thing you know about a problem is the input size

Answer: only cryptography, and this is by design.

For practical problems, the world comes equipped with many many additional parameters.

As we soon see, sensitizing the run times to parameters allows the development of a distinctive and often useful toolkit.

In particular, focusing on the parameters as a standard attack method for practice can be systematized, and sometimes even automated.

The theory equips us with both a positive and negative toolkit.

We are far from understanding the complexity of real data, and whether things like P vs NP even matters.
Without even going into details, think of all the graphs you have given names to and each has a relevant parameter: planar, bounded genus, bounded cutwidth, pathwidth, treewidth, degree, interval, etc, etc. In numerical analysis the degree of precision etc.

Also *nature* is kind in that for many practical problems the input (often designed by *us*) is nicely ordered.
Two Basic Examples

- VERTEX COVER
  Input: A Graph $G$.
  Parameter: A positive integer $k$.
  Question: Does $G$ have a size $k$ vertex cover? (Vertices cover edges.)

- DOMINATING SET
  Input: A Graph $G$.
  Parameter: A positive integer $k$.
  Question: Does $G$ have a size $k$ dominating set? (Vertices cover vertices.)
▶ VERTEX COVER is solvable by an algorithm $O$ in time $f(k)|G|$, a behaviour we call fixed parameter tractability, (Specifically $1.275^k k^2 + c|G|$, with $c$ a small absolute constant independent of $k$.)

▶ Whereas the only known algorithm for DOMINATING SET is complete search of the possible $k$-subsets, which takes time $\Omega(|G|^k)$. 
In the below I will mostly talk for convenience about graphs.
I could just as easily be talking about many other areas.
In the Computer Journal alone, there is biological, artificial intelligence, constraint satisfaction, geometric problems, scheduling, cognitive science, voting, combinatorial optimization, phylogeny.
Basic Definition(s)

- Setting: Languages $L \subseteq \Sigma^* \times \Sigma^*$.  
- Example (Graph, Parameter).
- We say that a language $L$ is **fixed parameter tractable** if there is an algorithm $M$, a constant $C$ and a function $f$ such that for all $x, k$,

  $$(x, k) \in L \text{ iff } M(x) = 1 \text{ and }$$

  $$\text{the running time of } M(x) = f(k)|x|^C.$$

- E.g. VERTEX COVER has $C = 1$. Vertex Cover has been implemented in computational biology for $k$ up to about 7000 and $n$ large.

OTHER POSSIBLE VERSIONS

VERTEX COVER: Vertices cover edges. Example: \{c, f, b, e, h\}.

GRAPH GENUS:
\(K_{3,3}\) has genus 1 by putting all lines except \(<b,2>\) on a sphere and \(<b,2>\) on a handle.

GRAPH LINKING NUMBER:
\(K_6\) has linking number 1.

\textbf{Figure}: Examples of FPT problems
There has been now over 20 years of research into this area.

What is not well known is that we now more or less can match upper and lower bounds for various algorithms, assuming a certain complexity hypothesis, like $P \neq NP$, (namely $M[1] \neq FPT$, no $2^{o(n)}$ algorithm for $n$-variable 3SAT. Impagliazzo, Paturi and Zane)

For example, (Cai and Juedes) Assuming this, there is no $2^{o(k)}|G|^c$ algorithm for $k$-PATH in a graph $G$.

There is known a $2^{O(k)}|G|^c$ algorithm!

That is, upper and lower bounds are matching up to big O.

In case you are interested, I have order forms for an excellent book....
Positive Techniques

- Elementary ones that work.
- Fancy metatechniques that almost work.
- Exotic methods that are even non-constructive. We know something is in P but don’t know the algorithm. E.g. embeddable on a torus, linkless embedding.
- Logical metatheorems that we thought did not work but now have some implementations.
- We care about the last one here. But I will begin with one easy algorithm.
- Limits
One of the key realizations is that many things we are interested in are constructed inductively using a finite set of “operators”.

This means that they can (i) be thought of as a “parse” formal language, and (ii) can attract “metrics” as parameters for the size of this “language”.

This allows for interactions complexity, automata, definability.

Ideas go back to Büchi, Rabin, Kleene, McNaughton etc.
(First order Logic)

1. **Atomic formulas**: $x = y$ and $R(x_1, ..., x_k)$, where $R$ is a $k$-ary relation symbol and $x, y, x_1, ..., x_k$ are individual variables, are FO-formulas.

2. **Conjunction, Disjunction**: If $\phi$ and $\psi$ are FO-formulas, then $\phi \land \psi$ is an FO-formula and $\phi \lor \psi$ is an FO-formula.

3. **Negation**: If $\phi$ is an FO-formula, then $\neg \phi$ is an FO-formula.

4. **Quantification**: If $\phi$ is an FO-formula and $x$ is an individual variable, then $\exists x \phi$ is an FO-formula and $\forall x \phi$ is an FO-formula.

**Eg** We can state that a graph has a clique of size $k$ using an FO-formula,

$$\exists x_1...x_k \bigwedge_{1\leq i\leq j\leq k} E(x_i, x_j)$$
Monadic Second Order Logic

- Two sorted structure with variables for sets of objects.
- **1. Additional atomic formulas:** For all set variables $X$ and individual variables $y$, $Xy$ is an MSO-formula.
- **2. Set quantification:** If $\phi$ is an MSO-formula and $X$ is a set variable, then $\exists X \phi$ is an MSO-formula, and $\forall X \phi$ is an MSO-formula.

- Eg $k$-colorability

\[
\exists X_1, \ldots, \exists X_k \left( \forall x \bigvee_{i=1}^{k} X_i x \land \forall x \forall y \left( E(x, y) \rightarrow \bigwedge_{i=1}^{k} \neg (X_i x \land X_i y) \right) \right)
\]

- Want to be careful as to what the primitives are. Eg. MSO graphs, MSO strings, etc. Many questions involve the translations between.
- For example, Büchi’s theorem for strings uses the ordering of strings, what about graphs of bounded treewidth (below)? No order there, etc.
Model Checking

Instance: A structure $A \in \mathcal{D}$, and a sentence (no free variables) $\phi \in \Phi$.
Question: Does $A$ satisfy $\phi$?

PSPACE-complete for FO and MSO.
Logicians such as Büchi, Rabin and others realized the connection between MSO and complexity, especially on formal languages. Below we will discuss this in the situation where objects of interest have parse languages, meaning that they are build up in an inductive manner.

MSO is central to program verification, and makes logic the calculus of computer science. See articles by Moshe Vardi.

That is, the connection is way beyond mere complexity.
Graphs constructed inductively. Treewidth, Pathwidth, Branschwidth, Cliquewidth mixed width etc. $k$-Inductive graphs, plus old favourites such as planarity etc, which can be viewed as local width.

Example:

**Definition**

[Tree decomposition and Treewidth] Let $G = (V, E)$ be a graph. A tree decomposition, $TD$, of $G$ is a pair $(T, \mathcal{X})$ where

1. $T = (I, F)$ is a tree, and
2. $\mathcal{X} = \{X_i | i \in I\}$ is a family of subsets of $V$, one for each node of $T$, such that

(i) $\bigcup_{i \in I} X_i = V$,

(ii) for every edge $\{v, w\} \in E$, there is an $i \in I$ with $v \in X_i$ and $w \in X_i$, and

(iii) for all $i, j, k \in I$, if $j$ is on the path from $i$ to $k$ in $T$, then $X_i \cap X_k \subseteq X_j$. 
This gives the following well-known definition.

**Definition**
The width of a tree decomposition \(((I, F), \{X_i \mid i \in I\})\) is \(\max_{i \in I} |X_i| - 1\). The treewidth of a graph \(G\), denoted by \(tw(G)\), is the minimum width over all possible tree decompositions of \(G\).
The following refers to any of these inductively defined graphs families. Notes that many commercial constructions of, for example chips are inductively defined.

1. Find a bounded-width tree (path) decomposition of the input graph that exhibits the underlying tree (path) structure.
2. Perform dynamic programming on this decomposition to solve the problem.
AN EXAMPLE FOR INDEPENDENT SET
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<th>b</th>
<th>c</th>
<th>ab</th>
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<th>bc</th>
<th>abc</th>
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<td>1</td>
<td>2</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
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</table>
The following theorem is shown that treewidth is FPT. Improves many earlier results showing this. The constant is about $2^{35k^3}$.

**Theorem (Bodlaender)**

$k$-TREewidth is linear time FPT

- Not practical because of large hidden $O$ term.
- Unknown if there is a practical FPT treewidth algorithm.
- Nevertheless approximation and algorithms specific to known decomps run well at least sometimes.
The model-checking problem for MSO restricted to graphs of bounded treewidth is linear-time fixed-parameter tractable.

Detlef Seese has proved a converse to Courcelle’s theorem.

Suppose that $\mathcal{F}$ is any family of graphs for which the model-checking problem for MSO is decidable, then there is a number $n$ such that, for all $G \in \mathcal{F}$, the treewidth of $G$ is less than $n$. 
The treewidth methodology has been applied in lots of areas. Finite model theory, braids, knots (Makowski, Rotics etc), matroids,

Can be used for hard problems like counting.

No general metheorem.

Generalized to other parse notions like CLIQUEWIDTH, \(d\)-DEGENERACY, NOWHERE DENSITY the idea being that formal languages and automata are basic.

Here I will look at a new application.
Triangulations

1. A *d-dimensional triangulation* consists of a collection of $d$-dimensional simplices $\Delta_1, \ldots, \Delta_n$ some or all of whose facets (ie. $d-1$-dimensional faces) are affinely identified.

2. Each facet $F$ may only be identified with at most one other facet $F'$ of a $d$-simplex. (This could be the same d-simplex but not the same facet.)

There are $\binom{d+1}{i+1}$ many $i$-faces of the $d$-simplices (where a 0-face is a vertex, a 1-face an edge etc).

A *d-manifold triangulation* is simply a $d$-dimensional triangulation whose underlying topological space is a $d$-manifold when using the quotient topology.

There are $\binom{d+1}{i+1}$ many $i$-faces of the $d$-simplices (where a 0-face is a vertex, a 1-face an edge etc).

The idea is to have a metatheorem for such triangulations.
An example-Klein Bottle

(a) A Klein bottle \( \mathcal{K} \)

(b) A one-tetrahedron solid torus

(c) The dual graph \( \mathcal{D}(\mathcal{K}) \)

\[ \Delta_1 : 02 \leftrightarrow \Delta_2 : 20, \quad \Delta_1 : 01 \leftrightarrow \Delta_1 : 12, \quad \Delta_2 : 01 \leftrightarrow \Delta_2 : 12. \]

Figure: Examples of triangulations
The resulting triangulation has one vertex (since all three vertices of $\Delta_1$ and all three vertices of $\Delta_2$ become identified together), and three edges (labelled $e, f, g$ in the diagram).

The dual graph $\mathcal{D}(T)$ of a triangulation $T$ is the multigraph whose nodes correspond to simplices and whose edges correspond to identified pairs of facets. Above we have the dual graph of the Klein bottle.
The language

- Standard boolean operations.
- for each \( i \in [0, d] \) variables for \( i \)-faces of a triangulation, and ones for sets of them.
- for each \( i \in [0, d] \), and each ordered sequence \( \pi_0, \ldots, \pi_i \) of distinct integers from \( \{0, \ldots, d\} \), a subface relation \( \leq_{\pi_0 \ldots \pi_i} \).
- The interpretation \( f \leq_{\pi_0 \ldots \pi_i} s \) indicates that \( f \) is a subface of the triangulation, \( s \) a simplex of the triangulation, and \( f \) is identified with the subface of \( s \) formed by the simplicial vertices \( \pi_0, \ldots, \pi_i \) in the way that vertices of the face \( 0, \ldots, i \) of the face \( f \) correspond to vertices \( \pi_0, \ldots, \pi_i \) of the simplex \( s \).
ORIENTABILITY

Recall: 2-dimensional triangulation is orientable if and only if each triangle can be assigned an orientation (clockwise or anticlockwise) so that adjacent triangles have compatible orientations, as illustrated in Figure ?? below.

**Figure:** Adjacent triangles have compatible orientations
**Theorem (Burton and D)**

For any fixed dimension $d$, let $K$ denote the class of $d$-dimensional triangulations whose dual graphs have universally bounded treewidth. Then for a fixed MSO $\phi$, and triangulation $T \in K$, we can decide if $T$ satisfies $\phi$ in linear time.

- We also show that the optimization problems can be solved.
- The method is one of reduction to the graph version.
- We apply this to various problems on these objects, including things called Taut Angle Structure, Discrete Morse Matching, Turaev-Viro Invariants, of which I know nothing.
Many Thanks