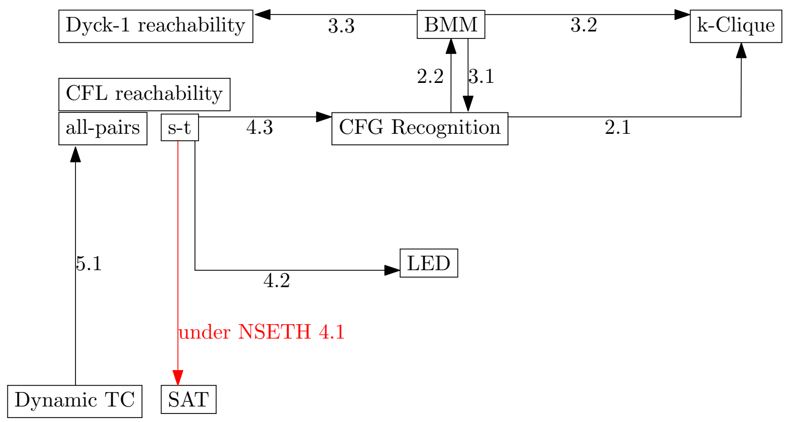
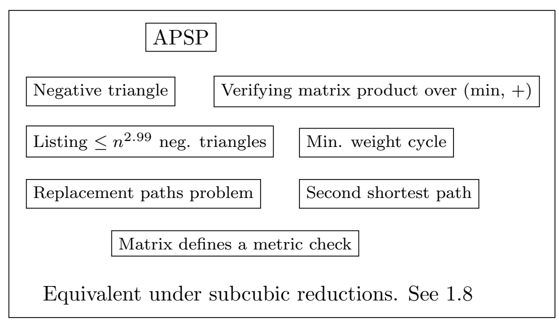
Reduction map





$A \to B \iff$ B has reduction to A

$\longrightarrow$ – no reduction

hypothesis

1. Problem definitions

1.1. **CFG recognition:** given a CFG $\mathcal{G}$ and a string $w \in \Sigma^*$ determine if w can be obtained from G (i.e. whether $w \in \mathcal{L}(\mathcal{G})$)

Best upper bound [12]: $\mathcal{O}(n^\omega \cdot |\mathcal{G}|^2)$

Best combinatorial algorithms: $\mathcal{O}(n^3/\log^3 n)$

CFG parsing problem: if $w \in \mathcal{L}(\mathcal{G})$, output a possible derivation sequence.

CFG recognition is as hard as CFG parsing up to logarithmic factors [11].

1.2. **CFL reachability:** given a CFG $\mathcal{L}$ and a labelled graph G

1.3. **k-Clique:** decide whether a given undirected unweighted graph on n nodes and $\mathcal{O}(n^2)$ edges contains a clique on k nodes.

It is a parameterized version of the NP-hard Max-Clique problem.

1.4. **SAT.**

1.4.1. *k-SAT:.* Given a k-CNF formula ϕ on n variables. Is there an assignment to the variables that satisfies ϕ?

1.4.2. *ETH (Exponential time hypothesis):* There is no algorithm that solves k-SAT in time $2^{o(n)}$ for any k.

1.4.3. *SETH (Strong exponential time hypothesis):* There is no $\epsilon > 0$ such that k-SAT can be solved in time $2^{(1-\epsilon)n}$ for any k.

1.4.4. *NSETH (Nondeterministic strong exponential time hypothesis):* There is no $\epsilon > 0$ such that k-SAT can be solved in time $2^{(1-\epsilon)n}$ co-nondeterministically for any k.

1.5. **BMM.** Given two $n \times n$ matrices A, B over $\{0, 1\}$, Boolean Matrix Multiplication (BMM) is $(AB)[i, j] = \vee_{k=1}^{n}(A(i, k) \wedge B(k, j))$.

Note that BMM can be computed using an algorithm for integer matrix multiplication in $\mathcal{O}(n^{\omega})$ time. Still, no combinatorial $\mathcal{O}(n^{3-\epsilon})$ algorithm is known.

1.5.1. *BMM conjecture [2].* In the Word RAM model with words of $\mathcal{O}(\log n)$ bits, any combinatorial algorithm requires $n^{3-o(1)}$ time in expectation to compute the boolean product of two $n \times n$ matrices.

1.6. **LED.** Given a string $w \in \Sigma^*$ and a context-free grammar $\mathcal{G}$ defined over the same alphabet, how many minimum number of repairs (insertions, deletions and substitutions) are required to map w into a valid member of $\mathcal{G}$?

1.7. **3SUM.** Determine whether a set $S \subset \{-n^3, \ldots, n^3\}$ of $|S| = n$ integers contains three distinct elements $a, b, c \in S$ with $a + b = c$.

1.7.1. *3SUM conjecture.* In the Word RAM model with words of $\mathcal{O}(\log n)$ bits, any algorithm requires $n^{2-o(1)}$ time in expectation to solve 3SUM problem.

1.8. **Triangle detection.**

1.8.1. *Triangle conjecture.* There is a constant $\delta > 0$, such that in the Word RAM model with words of $\mathcal{O}(\log n)$ bits, any algorithm requires $m^{1+\delta-o(1)}$ time in expectation to detect whether an m edge graph contains a triangle.

1.9. **APSP (all pairs shortest paths) and others [13]:**

- The all-pairs shortest paths problem on weighted digraphs (APSP).
- Detecting if a weighted graph has a triangle of negative total edge weight.
- Listing up to $n^{2.99}$ negative triangles in an edge-weighted graph.
- Finding a minimum weight cycle in a graph of non-negative edge weights.
- The replacement paths problem on weighted digraphs.
- Finding the second shortest simple path between two nodes in a weighted digraph.
- Checking whether a given matrix defines a metric.
- Verifying the correctness of a matrix product over the $(min, +)$-semiring (distance product).

1.9.1. *APSP conjecture.* There is a constant c, such that in the Word RAM model with words of $\mathcal{O}(\log n)$ bits, any algorithm requires $n^{3-o(1)}$ time in expectation to compute the distances between every pair of vertices in an n node graph with edge weights in $\{1, \ldots, n^c\}$.

1.9.2. *Negative Triangles Over Structures.* The negative triangle problem over $\mathcal{R}$ is defined on a weighted tripartite graph with parts I, J, K. Edge weights between I and J are from $\mathbb{Z}$, and all other edge weights are from $\mathcal{R}$. The problem is to detect if there are $i \in I, j \in J, k \in K$ so that $(w(i, k) \bigodot w(k, j)) + w(i, j) < 0$. Note that if one negates all weights of edges between I and J, the condition becomes $(w(i, k) \bigodot (k, j)) < w(i, j)$.

In the special case when $\bigodot = +$ and $\mathcal{R} \subseteq \mathbb{Z} \cup \{- - \infty, \infty\}$, the tripartiteness requirement is unnecessary, and the negative triangle problem is defined on an arbitrary graph with edge weights from $\mathbb{Z} \cup \{- - \infty, \infty\}$. This holds for the negative triangle problem over both the $(min, +)$ and Boolean semirings.

Note that detecting and finding one negative triangle in a graph problems reduce one to another:

Lemma (Folklore)**.** *Let $T(n)$ be a function so that $\frac{T(n)}{n}$ is nondecreasing. If there is a $T(n)$ time algorithm for negative triangle detection over $\mathcal{R}$ on a graph $G = (I \cup J \cup K, E)$, then there is an $\mathcal{O}(T(n))$ algorithm which returns a negative triangle over $\mathcal{R}$ in G if one exists.*

2. To CFG recongnition

2.1. From k-Clique problem [1]:

Summary: $T(n) \Rightarrow \mathcal{O}(T(n^{k/3+1})$ $(|G| = \mathcal{O}(1))$

2.2. From BMM [7]:

Summary: $\mathcal{O}(|G| \cdot n^{3-\epsilon}) \Rightarrow \mathcal{O}(m^{3-\epsilon/3})$ $(|G| = \Omega(n^6)!)$, combinatorial reduction

3. To BMM

3.1. From CFG recognition [12]:

Summary: $\mathcal{O}(n^{\omega}) \Rightarrow \mathcal{O}(n^{\omega})$

3.2. From k-Clique [9], [6]:

Summary: $\mathcal{O}(n^{\omega}) \Rightarrow \mathcal{O}(n^{i+\omega \cdot l}), k = 3 \cdot l + i, i = 0, 1, 2$

or $\mathcal{O}(n^{\omega(\lfloor k/3 \rfloor, \lfloor (k-1)/3 \rfloor, \lfloor k/3 \rfloor)}$, where $\mathcal{O}(n^{\omega(r,s,t)})$ denotes the running time of the multiplication of an $n^r \times n^s$ matrix by an $n^s \times n^t$ matrix.

Note that detecting arbitrary k-vertex (induced or not) subgraph in n-vertex graph is of the same complexity as detecting k-clique [9].

3.3. From Dyck-1 reachability [3], [8]:

Summary: $\mathcal{O}(n^{\omega}) \Rightarrow \mathcal{O}(n^{\omega} \cdot \log^2 n)$, combinatorial reduction [8]

Idea: find bell-shaped paths in $\mathcal{O}(\log n n^{\omega})$ by getting $\mathcal{O}(\log n)$ transitive closures; combine Dyck-1 path from bell-shaped paths: in each of $\mathcal{O}(\log n)$ iterations we remove bell-shaped paths of height at most $2^i, i = 1, \ldots, \log n$.

or $\mathcal{O}(n^{\omega}) \Rightarrow \mathcal{O}(n^{\omega} \cdot \log^3 n)$ in [3] via algebraic matrix encoding of flat Dyck1-Reachability to $\mathcal{O}(\log n)$ AGMY matrix multiplications.

4. To CFL reachability

4.1. To s-t reachability from SAT [5]:

Summary: subcubic certificates for CFL reachability (positive and negative), no reductions under NSETH from SAT (SETH) to CFL reachability.

Proof is based on the following lemma, where by instance (G, λ, s, t) of CFL reachability problem G is a graph, λ is edge-labelling function, s, t are vertices between which we search a path labelled with word from the language.

Lemma. *Let (G, λ, s, t) be an instance of the CFL reachability problem. There is a linear-time reduction (in the bit-size of the input) to an instance (G', λ', s', t') of the Dyck-2 reachability problem.*

4.2. **To s-t reachability from LED.** Naive reduction: build a path graph labelled with letters of w, add loops marked with every symbol of Σ on every vertex, add edges marked with every symbol of Σ and ϵ between pairs of adjacent vertices. Make original edges of zero weight, edges with Σ and ϵ symbols of positive weight (you can choose different weight-cost for different operations). Find path between end vertices marked with word formed from grammar with minimum weight.

4.3. **To s-t reachability from CFG recognition [4]:**
Summary: $\mathcal{O}(n^{3-\epsilon}) \Rightarrow \mathcal{O}(n^{3-\epsilon})$, combinatorial

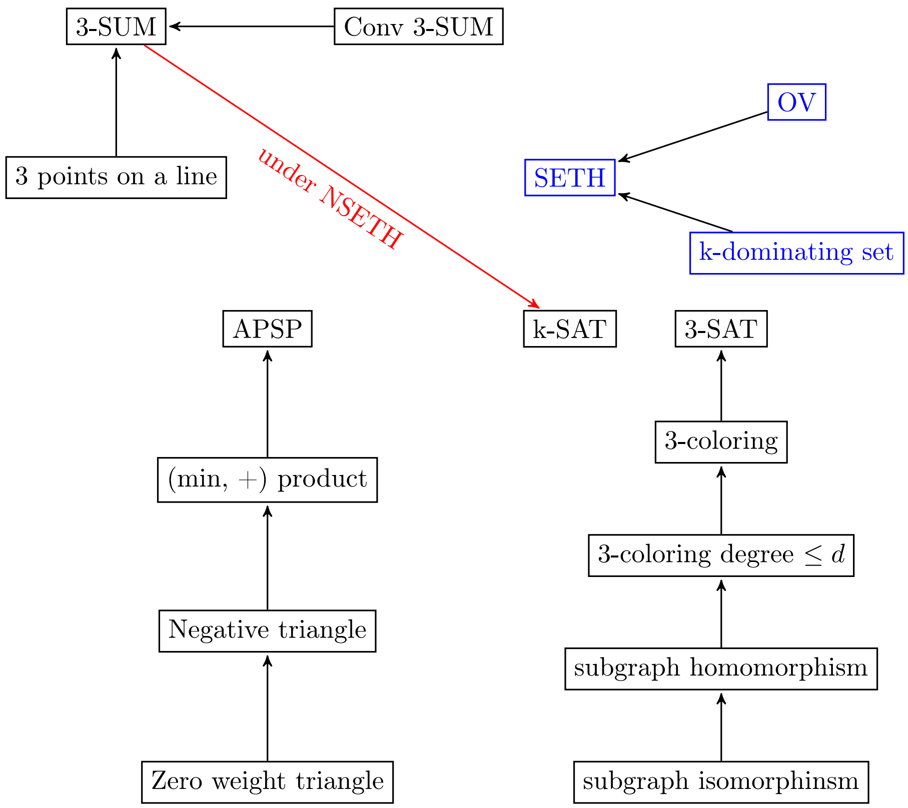
Lemma. *If there exists a combinatorial algorithm that solves the pair Dyck reachability problem in time $T(n)$, where n is the number of nodes of the input graph, then there exists a combinatorial algorithm that solves the CFL parsing problem in time $\mathcal{O}(n + T(n))$.*

In combination with reduction to BMM we have:

Theorem (BMM-hardness: Conditional cubic lower bound)**.** *For any fixed $\epsilon > 0$, if there is a combinatorial algorithm that solves the pair Dyck reachability problem in $\mathcal{O}(n^{3-\epsilon})$ time, then there is a combinatorial algorithm that solves Boolean Matrix Multiplication in $\mathcal{O}(n^{3-\epsilon})$ time.*

5. To Dynamic TC

5.1. **From all-pairs CFL reachability [10]:**
Summary: $\mathcal{O}(n^{3-\epsilon}) \Rightarrow \mathcal{O}(n^{3-\epsilon})$



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