New Inapproximability Bounds for TSP

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Abstract

In this paper, we study the approximability of the metric Traveling Salesman Problem (TSP) and prove new explicit inapproximability bounds for that problem. The best up to now known hardness of approximation bounds were $\frac{185}{184}$ for the symmetric case (due to Lampis) and $\frac{117}{116}$ for the asymmetric case (due to Papadimitriou and Vempala). We construct here two new bounded occurrence CSP reductions which improve these bounds to $\frac{123}{122}$ and $\frac{75}{74}$, respectively. The latter bound is the first improvement in more than a decade for the case of the asymmetric TSP. One of our main tools, which may be of independent interest, is a new construction of a bounded degree wheel amplifier used in the proof of our results.

1 Introduction

The Traveling Salesman Problem (TSP) is one of the best known and most fundamental problems in combinatorial optimization. Determining how well it can be approximated in polynomial time is therefore a major open problem, albeit one for which the solution still seems elusive. On the algorithmic side, the best known efficient approximation algorithm for the symmetric case is still a 35-year old algorithm due to Christofides [C76] which achieves an approximation ratio of $\frac{3}{2}$. However, recently there has been a string of improved results for the interesting special case of Graphic TSP, improving the ratio to $\frac{7}{5}$ [OSS11, MS11, M12, SV12]. For the asymmetric case (ATSP), it is not yet known if a constant-factor approximation is even possible, with the best known algorithm achieving a ratio of $O(\log n / \log \log n)$ [AGM+10].

Unfortunately, there is still a huge gap between the algorithmic results mentioned above and the best currently known hardness of approximation results for TSP and ATSP. For both problems, the known inapproximability thresholds are small constants ($\frac{185}{184}$}
and $117/116$ (cf. [L12, PV00], respectively). In this paper, we try to improve this situation somehow by giving modular hardness reductions that slightly improve the hardness bounds for both problems to $123/122$ and $75/74$, respectively. The latter bound is the first, for more than a decade now, improvement of Papadimitriou and Vempala bound [PV00] for the ATSP. The method of our solution differs essentially from that of [PV00] and uses some new paradigms of the bounded occurrence optimization which could be also of independent interest in other applications. Similarly to [L12], the hope is that the modularity of our construction, which goes through an intermediate stage of a bounded-occurrence Constraint Satisfaction Problem (CSP), will allow an easier analysis and simplify future improvements. Indeed, one of the main new ideas we rely on is a certain new variation of the wheel amplifiers first defined by Berman and Karpinski [BK01] to establish inapproximability for 3-regular CSPs. This construction, which may be of independent interest, allows us to establish inapproximability for a 3-regular CSP with a special structure. This special structure then makes it possible to simulate many of the constraints in the produced graph essentially “for free”, without using gadgets to represent them. Thus, even though for the remaining constraints we mostly reuse gadgets which have already appeared in the literature, we are still able to obtain improved bounds.

Let us now recall some of the previous work on the hardness of approximation of TSP and ATSP. Papadimitriou and Yannakakis [PY93] were the first to construct a reduction that, combined with the PCP Theorem [ALM+98], gave a constant inapproximability threshold, though the constant was not more than $1 + 10^{-6}$ for the TSP with distances either one or two. Engebretsen [E03] gave the first explicit approximation lower bound of $5381/5380$ for the problem. The inapproximability factor was improved to $3813/3812$ by Böckenhauer and Seibert [BS00], who studied the restricted version of the TSP with distances one, two and three. Papadimitriou and Vempala [PV00] proved that it is NP-hard to approximate the TSP with a factor better than $220/219$. Presently, the best known approximation lower bound is $185/184$ due to Lampis [L12].

The important restriction of the TSP, in which we consider instances with distances between cities being values in $\{1, \ldots, B\}$, is often referred to as the $(1, B)$-TSP. The best known efficient approximation algorithm for the $(1, 2)$-TSP has an approximation ratio $8/7$ and is due to Berman and Karpinski [BK06]. As for lower bounds, Engebretsen and Karpinski [EK06] gave inapproximability thresholds for the $(1, B)$-TSP problem of $741/740$ for $B = 2$ and $389/388$ for $B = 8$. More recently, Karpinski and Schmied [KS12, KS13] obtained improved inapproximability factors for the $(1, 2)$-TSP and the $(1, 4)$-TSP of $535/534$ and $337/336$, respectively.

For ATSP the currently best known approximation lower bound was $117/116$ due to Papadimitriou and Vempala [PV00]. When we restrict the problem to distances with values in $\{1, \ldots, B\}$, there is a simple approximation algorithm with approximation ratio $B$ that constructs an arbitrary tour as solution. Bläser [B04] gave an efficient approximation algorithm for the $(1, 2)$-ATSP with approximation ratio $5/4$. Karpinski and Schmied [KS12, KS13] proved that it is NP-hard to approximate the $(1, 2)$-ATSP and the $(1, 4)$-ATSP within any factor less than $207/206$ and $141/140$, respectively. For the case $B = 8$, Engebretsen
and Karpinski [EK06] gave an inapproximability threshold of $\frac{135}{134}$.

**Overview:** In this paper we give a hardness proof which proceeds in two steps. First, we start from the MAX-E3-LIN2 problem, in which we are given a system of linear equations mod 2 with exactly three variables in each equation and we want to find an assignment such as to maximize the number of satisfied equations. Optimal inapproximability results for this problem were shown by Håstad [H01]. We reduce this problem to a special case where variables appear exactly 3 times and the linear equations have a particular structure. The main tool here is a new variant of the wheel amplifier graphs of Berman and Karpinski [BK01].

In the second step, we reduce this 3-regular CSP to TSP and ATSP. The general construction is similar in both cases, though of course we use different gadgets for the two problems. The gadgets we use are mostly variations of gadgets which have already appeared in previous reductions. Nevertheless, we manage to obtain an improvement by exploiting the special properties of the 3-regular CSP. In particular, we show that it is only necessary to construct gadgets for roughly one third of the constraints of the CSP instance, while the remaining constraints are simulated without additional cost using the consistency properties of our gadgets. This idea may be useful in improving the efficiency of approximation-hardness reductions for other problems.

Thus, overall we follow an approach unlike that of [PV00], where the reduction is performed in one step, and closer to [L12]. The improvement over [L12] comes mainly from the idea mentioned above, which is made possible using the new wheel amplifiers, as well as several other tweaks. The end result is a more economical reduction which improves the bounds for both TSP and ATSP.

An interesting question may be whether our techniques can also be used to derive improved inapproximability results for variants of the ATSP and TSP (cf. [EK06],[KS13] and [KS12]) or other graph problems, such as the Steiner Tree problem.

## 2 Preliminaries

In the following, we give some definitions concerning directed (multi-)graphs and omit the corresponding definitions for undirected (multi-)graphs if they follow from the directed case. Given a directed graph $G = (V(G), E(G))$ and $E' \subseteq E(G)$, for $e = (x, y) \in E(G)$, we define $V(e) = \{x, y\}$ and $V(E') = \bigcup_{e \in E'} V(e)$. For convenience, we abbreviate a sequence of edges $(x_1, x_2), (x_2, x_3), \ldots, (x_{n-1}, x_n)$ by $x_1 \rightarrow x_2 \rightarrow x_3 \rightarrow \ldots \rightarrow x_{n-1} \rightarrow x_n$. In the undirected case, we use sometimes $x_1 - x_2 - x_3 - \ldots - x_{n-1} - x_n$ instead of $\{x_1, x_2\}, \{x_2, x_3\}, \ldots, \{x_{n-1}, x_n\}$. Given a directed (multi-)graph $G$, an Eulerian cycle in $G$ is a directed cycle that traverses all edges of $G$ exactly once. We refer to $G$ as Eulerian, if there exists an Eulerian cycle in $G$. For a multiset $E_T$ of directed edges and $v \in V(E_T)$, we define the outdegree (indegree) of $v$ with respect to $E_T$, denoted by $\text{outd}_T(v) (\text{ind}_T(v))$, to be the number of edges in $E_T$ that are outgoing of (incoming to) $v$. The balance of a vertex $v$ with respect to $E_T$ is defined as $\text{bal}_T(v) = \text{ind}_T(v) - \text{outd}_T(v)$. In the case of a multiset $E_T$ of undirected edges, we define the balance $\text{bal}_T(v)$ of a vertex $v \in V(E_T)$
to be one if the number of incident edges in $E_T$ is odd and zero otherwise. We refer to vertices $v \in V(E_T)$ with $bal_T(v) = 0$ as balanced with respect to $E_T$. It is well known that a (directed) (multi-)graph $G = (V(G), E(G))$ is Eulerian if and only if all edges are in the same (weakly) connected component and all vertices $v \in V(G)$ are balanced with respect to $E(G)$.

Given a multiset of edges $E_T$, we denote by $con_T$ the number of (weakly) connected components in the graph induced by $E_T$. A quasi-tour $E_T$ in a (directed) graph $G = (V(G), E(G))$ is a multiset of edges from $E(G)$ such that all vertices are balanced with respect to $E_T$ and $V(E_T) = V(G)$. We refer to a quasi-tour $E_T$ in $G$ as a tour if $con_T = 1$. Given a cost function $w : E(G) \to \mathbb{R}_+$, the cost of a quasi-tour $E_T$ in $G$ is defined by $\sum_{e \in E_T} w(e) + 2(con_T - 1)$.

In the Asymmetric Traveling Salesman problem (ATSP), we are given a directed graph $G = (V(G), E(G))$ with positive weights on edges and we want to find an ordering $v_1, \ldots, v_n$ of the vertices such as to minimize $d_G(v_n, v_1) + \sum_{i \in [n-1]} d_G(v_i, v_{i+1})$, where $d_G$ denotes the shortest path distance in $G$.

In this paper, we will use the following equivalent reformulation of the ATSP: Given a directed graph $G$ with weights on edges, we want to find a tour $E_T$ in $G$, that is, a spanning connected multi-set of edges that balances all vertices, with minimum cost.

The metric Traveling Salesman problem (TSP) is the special case of the ATSP, in which instances are undirected graphs with positive weights on edges.

### 3 Bi-Wheel Amplifiers

In this section, we define the bi-wheel amplifier graphs which will be our main tool for proving hardness of approximation for a bounded occurrence CSP with some special properties. Bi-wheel amplifiers are a simple variation of the wheel amplifier graphs given in [BK01]. Let us first recall some definitions (see also [BK03]).

If $G$ is an undirected graph and $X \subset V(G)$ a set of vertices, we say that $G$ is a $\Delta$-regular amplifier for $X$ if the following conditions hold:

- All vertices of $X$ have degree $\Delta - 1$ and all vertices of $V(G) \setminus X$ have degree $\Delta$.

- For every non-empty subset $U \subset V(G)$, we have the condition that $|E(U, V(G) \setminus U)| \geq \min\{|U \cap X|, |(V(G) \setminus U) \cap X|\}$, where $E(U, V(G) \setminus U)$ is the set of edges with exactly one endpoint in $U$.

We refer to the set $X$ as the set of contact vertices and to $V(G) \setminus X$ as the set of checker vertices. Amplifier graphs are useful in proving inapproximability for CSPs, in which every variable appears a bounded number of times. Here, we will rely on $3$-regular amplifiers. A probabilistic argument for the existence of such graphs was given in [BK01], with the definition of wheel amplifiers.

A wheel amplifier with $2n$ contact vertices is constructed as follows: first construct a cycle on $14n$ vertices. Number the vertices $1, \ldots, 14n$ and select uniformly at random a perfect matching of the vertices whose number is not a multiple of 7. The matched
vertices will be our checker vertices, and the rest our contacts. It is easy to see that the
degree requirements are satisfied.

Berman and Karpinski [BK01] gave a probabilistic argument to prove that with high
probability the above construction indeed produces an amplifier graph, that is, all parti-
tions of the sets of vertices give large cuts. Here, we will use a slight variation of this
construction, called a bi-wheel.

A bi-wheel amplifier with \(2n\) contact vertices is constructed as follows: first construct
two disjoint cycles, each on \(7n\) vertices and number the vertices of each \(1, \ldots, 7n\). The
contacts will again be the vertices whose number is a multiple of 7, while the remaining
vertices will be checkers. To complete the construction, select uniformly at random a
perfect matching from the checkers of one cycle to the checkers of the other.

Intuitively, the reason that amplifiers are a suitable tool here is that, given a CSP in-
estance, we can use a wheel amplifier to replace a variable that appears \(2n\) times with \(14n
new variables (one for each wheel vertex) each of which appears 3 times. Each appearance
of the original variable is represented by a contact vertex and for each edge of the wheel
we add an equality constraint between the corresponding variables. We can then use the
property that all partitions give large cuts to argue that in an optimal assignment all the
new vertices take the same value.

We use the bi-wheel amplifier in our construction in a similar way. The main differ-
ence is that while cycle edges will correspond to equality constraints, matching edges will
correspond to inequality constraints. The contacts of one cycle will represent the posi-
tive appearances of the original variable, and the contacts of the other the negative ones.
The reason we do this is that we can encode inequality constraints more efficiently than
equality with a TSP gadget, while the equality constraints that arise from the cycles will
be encoded in our construction “for free” using the consistency of the inequality gadgets.

Before we apply the construction however, we have to prove that the bi-wheel ampli-
fiers still have the desired amplification properties.

**Theorem 1.** With high probability, bi-wheels are 3-regular amplifiers.

*Proof.* Exploiting the similarity between bi-wheels and the standard wheel amplifiers of
[BK01], we will essentially reuse the proof given there. First, some definitions: We say
that \(U\) is a bad set if the size of its cut is too small, violating the second property of
amplifiers. We say that it is a minimal bad set if \(U\) is bad but removing any vertex from \(U
\)gives a set that is not bad.

Recall the strategy of the proof from [BK01]: for each partition of the vertices into \(U
\)and \(V(G)\setminus U\), they calculate the probability (over the random matchings) that this partition
gives a minimal bad set. Then, they take the sum of these probabilities over all poten-
tially minimal bad sets and prove that the sum is at most \(\gamma^{-n}\) for some constant \(\gamma < 1\). It follows
by union bound that with high probability, no set is a minimal bad set and therefore, the
graph is a proper amplifier.

Our first observation is the following: consider a wheel amplifier on \(14n\) vertices
where, rather than selecting uniformly at random a perfect matching among the check-
ers, we select uniformly at random a perfect matching from checkers with labels in the set
\{1, \ldots, 7n - 1\} to checkers with labels in the set \(\{7n + 1, \ldots, 14n - 1\}\). This graph is almost isomorphic to a bi-wheel. More specifically, for each bi-wheel, we can obtain a graph of this form by rewiring two edges, and vice-versa. It easily follows that properties that hold for this graph, asymptotically with high probability also hold for the bi-wheel.

Thus, we just need to prove that a wheel amplifier still has the amplification property if, rather than selecting a random perfect matching, we select a random matching from one half of the checker vertices to the other. We will show this by proving that, for each set of vertices \(S\), the probability that \(S\) is a minimal bad set is roughly the same in both cases. After establishing this fact, we can simply rely on the proof of [BK01].

Recall that the wheel has \(12n\) checker vertices. Given a set \(S\) with \(|S| = u\), what is the probability that exactly \(c\) edges have exactly one endpoint in \(S\)? In a standard wheel amplifier the probability is

\[
P(u, c) = \binom{u}{c} \binom{12n - u}{c} \frac{c!(u - c)!(12n - u - c)!!}{(12n)!!},
\]

where we denote by \(n!!\) the product of all odd natural numbers less than or equal to \(n\), and we assume without loss of generality that \(u - c\) is even. Let us explain this: the probability that exactly \(c\) edges cross the cut in this graph is equal to the number of ways we can choose their endpoints in \(S\) and in its complement, times the number of ways we can match the endpoints, times the number of matchings of the remaining vertices, divided by the number of matchings overall.

How much does this probability change if we only allow matchings from one half of the checkers to the other? Intuitively, we need to consider two possibilities: one is that \(S\) is a balanced set, containing an equal number of checkers from each side, while the other is that \(S\) is unbalanced. It is not hard to see that if \(S\) is unbalanced, then, we can easily establish that the cut must be large. Thus, the main interesting case is the balanced one (and we will establish this fact more formally).

Suppose that \(|S| = u\) and \(S\) contains exactly \(u/2\) checkers from each side. Then the probability that there are exactly \(c\) edges crossing the cut is

\[
P'(u, c) = \left(\frac{u}{2}\right)^2 \left(\frac{12n - u}{2}\right)^2 \left(\frac{c}{2}\right)! \frac{(u - c)!}{(u - c)/2}! \frac{(12n - u - c)/2)!}{(6n)!}
\]

Let us explain this. If \(S\) is balanced and there are \(c\) matching edges with exactly one endpoint in \(S\), then, exactly \(c/2\) of them must be incident on a vertex of \(S\) on each side, since the remaining vertices of \(S\) must have a perfect matching. Again, we pick the endpoints on each side, and on the complement of \(S\), select a way to match them, select matchings on the remaining vertices and divide by the number of possible perfect matchings.

Using Stirling formulas, it is not hard to see that \((\frac{n}{2})!^2 = \Theta(n!2^{-n}\sqrt{n})\). Also \(n!! = \Theta((\frac{n}{2})!^{12n/2})\). It follows that \(P'\) is roughly the same as \(P\) in this case, modulo some polynomial factors which are not significant since the probabilities we are calculating are exponentially small.
Let us now also show that if $S$ is unbalanced, the probability that it is a minimal bad set is even smaller. First, observe that if $S$ is a minimal bad set whose cut has $c$ edges, we have $c \leq u/6$. The reason for this is that since $S$ is bad, then, $c$ is smaller than the number of contacts in $S$ minus the number of cycle edges cut. It is not hard to see that, in each fragment, that is, each subset of $S$ made up of a contiguous part of the cycle, two cycle edges are cut. Thus, the extra edges we need for the contacts the fragment contains are at most $1/6$ of its checkers.

Suppose now that $S$ contains $u/2 + k$ checkers on one side and $u/2 - k$ checkers on the other. The probability that $c$ matching edges have one endpoint in $S$ is

$$P''(u, c, k) = \binom{\frac{u}{2} + k}{\frac{u}{2} - k} \binom{\frac{12n - u}{2} - k}{\frac{c}{2} + k} \binom{\frac{12n - u}{2} + k}{\frac{c}{2} - k} \frac{(u - c)!}{(6n)!}$$

The reasoning is the same as before, except we observe that we need to select more endpoints on the side where $S$ is larger, since after we remove checkers matched to outside vertices $S$ must have a perfect matching. Observe that for $k = 0$ this gives $P'$. We will show that for the range of values we care about $P''$ achieves a maximum for $k = 0$, and can thus be upper-bounded by (essentially) $P$, which is the probability that a set is bad in the standard amplifier. The rest of the proof follows from the argument given in [BK01]. In particular, we can assume that $k \leq c/2$, since $2k$ edges are cut with probability 1. To show that the maximum is achieved for $k = 0$, we look at $P''(u, c, k + 1)/P''(u, c, k)$. We will show that this is less than 1. Using the identity $\binom{n+1}{k+1}/\binom{n}{k} = \frac{n+1}{k+1}$, we get

$$\frac{P''(u, c, k + 1)}{P''(u, c, k)} = \left(1 - \frac{2k + 1}{\frac{c}{2} + k + 1}\right) \left(1 + \frac{2k + 1}{\frac{c}{2} - k}\right) \left(1 + \frac{2k + 1}{\frac{12n - u}{2} - k}\right)$$

Using the fact that $1 + x < e^x$, we end up needing to prove the following.

$$\frac{2k + 1}{\frac{c}{2} + k + 1} > \frac{2k + 1}{\frac{c}{2} - k} + \frac{2k + 1}{\frac{12n - u}{2} - k}$$

Combining that without loss of generality $u \leq 6n$ holds with the bounds of $c$ and $k$ we have already mentioned, the inequality (1) is straightforward to establish.

\[\square\]

4 Hybrid Problem

By using the bi-wheel amplifier from the previous section, we are going to prove hardness of approximation for a bounded occurrence CSP with very special properties. This particular CSP will be well-suited for constructing a reduction to the TSP given in the next section.

As the starting point of our reduction, we make use of the inapproximability result due to Håstad [H01] for the MAX-E3LIN2 problem, which is defined as follows: Given a system $I_1$ of linear equations mod 2, in which each equation is of the form $x_i \oplus x_j \oplus x_k = b_{ijk}$ with
b_{ijk} \in \{0,1\}, we want to find an assignment to the variables of \(I_1\) such as to maximize the number of satisfied equations.

Let \(I_1\) be an instance of the MAX-E3LIN2 problem and \(\{x_i\}_{i=1}^\nu\) the set of variables, that appear in \(I_1\). We denote by \(d(i)\) the number of appearances of \(x_i\) in \(I_1\).

**Theorem 2** (Håstad [H01]). For every \(\epsilon > 0\), there exists a constant \(B_\epsilon\) such that given an instance \(I_1\) of the MAX-E3LIN2 problem with \(m\) equations and \(\max_{i \in [\nu]} d(i) \leq B_\epsilon\), it is \(\mathsf{NP}\)-hard to decide whether there is an assignment that leaves at most \(\epsilon \cdot m\) equations unsatisfied, or all assignment leave at least \((0.5 - \epsilon)m\) equations unsatisfied.

Similarly to the work by Berman and Karpinski [BK99] (see also [BK01] and [BK03]), we will reduce the number of occurrences of each variable to 3. For this, we will use our amplifier construction to create special instances of the Hybrid problem, which is defined as follows: Given a system \(I_2\) of linear equations mod 2 with either three or two variables in each equation, we want to find an assignment such as to maximize the number of satisfied equations.

In particular, we are going to prove the following theorem.

**Theorem 3.** For every constant \(\epsilon > 0\) and \(b \in \{0,1\}\), there exist instances of the Hybrid problem with \(31m\) equations such that: (i) Each variable occurs exactly three times. (ii) \(21m\) equations are of the form \(x \oplus y = 0\), \(9m\) equations are of the form \(x \oplus y = 1\) and \(m\) equations are of the form \(x \oplus y \oplus z = b\). (iii) It is \(\mathsf{NP}\)-hard to decide whether there is an assignment to the variables leaves at most \(\epsilon \cdot m\) equations unsatisfied, or every assignment to the variables leaves at least \((0.5 - \epsilon)m\) equations unsatisfied.

**Proof.** Let \(\epsilon > 0\) be a constant and \(I_1\) an instance of the MAX-E3LIN2 problem with \(\max_{i \in [\nu]} d(i) \leq B_\epsilon\). For a fixed \(b \in \{0,1\}\), we can flip some of the literals such that all equations in the instance \(I_1\) are of the form \(x \oplus y \oplus z = b\), where \(x, y, z\) are variables or negations. By constructing three more copies of each equation, in which all possible pairs of literals appear negated, we may assume that each variable occurs the same number of times negated as unnegated.

Let us fix a variable \(x_i\) in \(I_1\). Then, we create \(7 \cdot d(i) = 2 \cdot \alpha\) new variables \(\text{Var}(i) = \{x_i^{ni}, x_i^{ni}\}_{j=1}^\alpha\). In addition, we construct a bi-wheel amplifier \(W_i\) on \(2 \cdot \alpha\) vertices (that is, a bi-wheel with \(d(i)\) contact vertices) with the properties described in Theorem 1. Since \(d(i) \leq B_\epsilon\) is a constant, this can be accomplished in constant time. In the remainder, we refer to contact and checker variables as the elements in \(\text{Var}(i)\), whose corresponding index is a contact and checker vertex in \(W_i\), respectively. We denote by \(M(W_i) \subseteq E(W_i)\) the associated perfect matching on the set of checker vertices of \(W_i\). In addition, we denote by \(C_n(W_i)\) and \(C_u(W_i)\) the set of edges contained in the first and second cycle of \(W_i\), respectively.

Let us now define the equations of the corresponding instance of the Hybrid problem. For each edge \(\{j, k\} \in M(W_i)\), we create the equation \(x_j^{ni} \oplus x_k^{ni} = 1\) and refer to equations of this form as matching equations. On the other hand, for each edge \(\{l, t\}\) in the cycle \(C_q(W_i)\) with \(q \in \{u, n\}\), we introduce the equation \(x_l^{qi} \oplus x_t^{qi} = 0\). Equations of this form will
be called cycle equations. Finally, we replace the \( j \)-th unnegated appearance of \( x_i \) in \( I_1 \) by the contact variable \( x_{\lambda}^{ui} \) with \( \lambda = 7 \cdot j \), whereas the \( j \)-th negated appearance is replaced by \( x_{\lambda}^{ni} \). The former construction yields \( m \) equations with three variables in the instance of the Hybrid problem, which we will denote by \( I_2 \). Notice that each variable appears in exactly 3 equations in \( I_2 \). Clearly, we have \( |I_2| = 31m \) equations, thereof \( 9m \) matching equations, \( 21m \) cycle equations and \( m \) equations of the form \( x \oplus y \oplus z = b \).

A consistent assignment to \( \text{Var}(i) \) is an assignment with \( x_{\lambda}^{ui} = b \) and \( x_{\lambda}^{ni} = (1 - b) \) for all \( j \in [\alpha] \), where \( b \in \{0, 1\} \). A consistent assignment to the variables of \( I_2 \) is an assignment that is consistent to \( \text{Var}(i) \) for all \( i \in [\nu] \). By standard arguments using the amplifier constructed in Theorem 1, it is possible to convert an assignment to a consistent assignment without decreasing the number of satisfied equations and the proof of Theorem 3 follows.

\[ \square \]

5 TSP

This section is devoted to the proof of the following theorem.

**Theorem 4.** It is \( \text{NP} \)-hard to approximate the TSP to within any constant approximation ratio less than \( 123/122 \).

Let us first sketch the high-level idea of the construction. Starting with an instance of the Hybrid problem, we will construct a graph, where gadgets represent the equations. We will design gadgets for equations of size three (Figure 1) and for equations of size two corresponding to matching edges of the bi-wheel (Figure 2). We will not construct gadgets for the cycle edges of the bi-wheel; instead, the connections between the matching edge gadgets will be sufficient to encode these extra constraints. This may seem counterintuitive at first, but the idea here is that if the gadgets for the matching edges are used in a consistent way (that is, the tour enters and exits in the intended way) then it follows that the tour is using all edges corresponding to one wheel and none from the other. Thus, if we prove consistency for the matching edge gadgets, we implicitly get the cycle edges “for free”. This observation, along with an improved gadget for size-three equations and the elimination of the variable part of the graph, are the main sources of improvement over the construction of [L12].

5.1 Construction

We are going to describe the construction that encodes an instance \( I_2 \) of the Hybrid problem into an instance of the TSP problem. Due to Theorem 3 we may assume that the equations with three variables in \( I_2 \) are all of the form \( x \oplus y \oplus z = 0 \).

In order to ensure that some edges are to be used at least once in any valid tour, we apply the following simple trick that was already used in the work by Lampis [L12]: Let \( e \) be an edge with weight \( w \) that we want to be traversed by every tour. We remove \( e \) and
replace it with a path of \( L \) edges and \( L - 1 \) newly created vertices each of degree two, where we think of \( L \) as a large constant. Each of the \( L \) edges has weight \( w/L \) and any tour that fails to traverse at least two newly created edges is not connected. On the other hand, a tour that traverses all but one of those edges can be extended by adding two copies of the unused edge increasing the cost of the underlying tour by a negligible value. In summary, we may assume that our construction contains forced edges that need to be traversed at least once by any tour. If \( x \) and \( y \) are vertices, which are connected by a forced edge \( e \), we write \( \{x, y\}_F \) or simply \( x - F y \). In the following, we refer to unforced edges \( e \) with \( w(e) = 1 \) as simple. All unforced edges in our construction will be simple.

Let us start with the description of the corresponding graph \( G_S \): For each bi-wheel \( W_p \), we will construct the subgraph \( G_p \) of \( G_S \). For each vertex of the bi-wheel, we create a vertex in the graph and for each cycle equation \( x \oplus y = 0 \), we create a simple edge \( \{x, y\} \). Given a matching equation between two checkers \( x^u_i \oplus x^n_j = 1 \), we connect the vertices \( x^u_i \) and \( x^n_j \) with two forced edges \( \{x^u_i, x^n_j\}_1 \) and \( \{x^u_i, x^n_j\}_2 \). We have \( w(\{x^u_i, x^n_j\}_F) = 2 \) for each \( i \in \{1, 2\} \).

Additionally, we create a central vertex \( s \) that is connected to gadgets simulating equations with three variables. Let \( x \oplus y \oplus z = 0 \) be the \( j \)-th equation with three variables in \( I_2 \). We now create the graph \( G^S_j \) displayed in Figure 1(a), where the (contact) vertices for \( x, y, z \) have already been constructed in the cycles. The edges \( \{\gamma^\alpha, \gamma\}_F \) with \( \alpha \in \{r, l\} \) and \( \gamma \in \{x, z, y\} \) are all forced edges with \( w(\{\gamma^\alpha, \gamma\}_F) = 1.5 \). Furthermore, we have \( w(\{e^\alpha_j, s\}_F) = 0.5 \) for all \( \alpha \in \{r, l\} \). \( \{e^r_j, s\}_F \) and \( \{e^l_j, s\}_F \) are both forced edges, whereas all remaining edges of \( G^S_j \) are simple. This is the whole description of \( G_S \).

![Figure 1](image.png)

Figure 1: Gadgets simulating equations with three variables in the symmetric case (a) and in the asymmetric case (b). Dotted and straight lines represent forced and simple edges, respectively.
5.2 Tour from Assignment

Given an instance $I_2$ of the Hybrid problem and an assignment $\phi$ to the variables in $I_2$, we are going to construct a tour in $G_S$ according to $\phi$ and give the proof of one direction of the reduction. In particular, we are going to prove the following lemma.

**Lemma 1.** If there is an assignment to the variables of a given instance $I_2$ of the Hybrid problem with $31m$ equations and $\nu$ bi-wheels, that leaves $k$ equations unsatisfied, then, there exists a tour in $G_S$ with cost at most $61m + 2\nu + k + 2$.

Before we proceed, let us give a useful definition. Let $G$ be an edge-weighted graph and $E_T$ a multi-set of edges of $E(G)$ that defines a quasi-tour. Consider a set $V' \subseteq V(G)$. The local edge cost of the set $V'$ is then defined as

$$c_T(V') = \sum_{u \in V'} \sum_{e \in E_T, e=\{u,v\}} \frac{w(e)}{2}$$

In words, for each vertex in $V'$, we count half the total weight of its incident edges used in the quasi-tour (including multiplicities). Observe that this sum contains half the weight of edges with one endpoint in $V'$ but the full weight for edges with both endpoints in $V'$ (since we count both endpoints in the sum). Also note that for two sets $V_1, V_2$, we have $c_T(V_1 \cup V_2) \leq c_T(V_1) + c_T(V_2)$ (with equality for disjoint sets) and that $c_T(V) = \sum_{e \in E_T} w(e)$.

**Proof.** First, note that it is sufficient to prove that we can construct a quasi-tour of the promised cost which uses all forced edges exactly once. Since all unforced edges have cost 1, if we are given a quasi-tour we can connect two disconnected components by using an unforced edge that connects them twice (this is always possible since the underlying graph we constructed is connected). This does not increase the cost, since we added two unit-weight edges and decreased the number of components. Repeating this results in a connected tour.

Let $\{W_a\}_{a=1}^{\nu}$ be the associated set of bi-wheels of $I_2$. For a fixed bi-wheel $W_a$, let $\{x_i^a, x_j^a\}_{i=1}^2$ be its associated set of variables. Due to the construction of instances of the Hybrid problem in Section 4, we may assume that all equations with two variables are satisfied by the given assignment. Thus, we have $x_i^a \neq x_j^a$, $x_i^a = x_j^a$ and $x_i^a = x_j^a$ for all $i, j \in [z]$.

Assuming $x_i^a = 1$ for some $\alpha \in \{u, n\}$, we use once all simple edges $\{x_i^a, x_{i+1}^a\}$ with $i \in [z - 1]$ and the edge $\{x_i^a, x_j^a\}$. We also use all forced edges corresponding to matching equations once. In other words, for each biwheel we select the cycle that corresponds to the assignment 1 and use all the simple edges from that cycle. This creates a component that contains all checker vertices from both cycles and all contacts from one cycle.

As for the next step, we are going to describe the tour traversing $G^{3S}_j$ with $j \in [m]$ given an assignment to contact variables. Let us assume that $G^{3S}_j$ simulates $x \oplus y \oplus z = 0$. According to the assignment to $x, y$ and $z$, we will traverse $G^{3S}_j$ as follows: In all cases, we will use all forced edges once.

**Case ($x + y + z = 2$):** Then, we use $\{\gamma', \gamma''\}$ for all $\alpha \in \{r, l\}$ and $\gamma \in \{x, y, z\}$ with $\gamma = 1$. 

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For $\delta \in \{x, z, y\}$ with $\delta = 0$, we use $\{e^\alpha_j, \delta^\alpha\}$ for all $\alpha \in \{r, l\}$.

**Case** ($x + y + z = b$ with $b \in \{0, 1\}$): In both cases, we traverse $\{\gamma^\alpha, e^\alpha_j\}$ for all $\gamma \in \{x, y, z\}$ and $\alpha \in \{r, l\}$.

**Case** ($x + y + z = 3$): We use $\{\gamma^r, \gamma^l\}$ with $\gamma \in \{y, z\}$. Furthermore, we include $\{x^\alpha, e^\alpha_j\}$ for both $\alpha \in \{r, l\}$.

Let us now analyze the cost of the edges of our quasi-tour given an assignment. For each matching edge $\{x^u_i, x^n_j\}$ consider the set of vertices made up of its endpoints. Its local cost is 5: we pay 4 for the forced edges and there are two used simple edges with one endpoint in the set. Let us also consider the local cost for a size-three equation gadget, where we consider the set to contain the contact vertices $\{x, y, z\}$ as well the other 8 vertices of the gadget. The local cost here is 9.5 for the forced edges. We also pay 6 more (for a total of 15.5) when the assignment satisfies the equation or 7 more when it does not.

Thus, we have given a covering of the vertices of the graph by $9m$ sets of size two, $m$ sets of size 11 and $\{s\}$. The total edge cost is thus at most $5 \cdot 9m + 15.5 \cdot m + 0.5 \cdot m + k = 61m + k$.

To obtain an upper bound on the cost of the quasi-tour, we observe that the tour has at most $\nu + 1$ components (one for each bi-wheel and one containing $s$). The lemma follows.

### 5.3 Assignment from Tour

In this section, we are going prove the other direction of our reduction. Given a tour in $G_S$, we are going to define an assignment to the variables of the associated instance of the Hybrid problem and give the proof of the following lemma.

**Lemma 2.** If there is a tour in $G_S$ with cost $61m + k - 2$, then, there is an assignment to the variables of the corresponding instance of the Hybrid problem that leaves at most $k$ equations unsatisfied.
Again, let us give a useful definition. Consider a quasi-tour $E_T$ and a set $V' \subseteq V(G)$. Let $con_T(V')$ be the number of connected components induced by $E_T$ which are fully contained in $V'$. Then, the full local cost of the set $V'$ is defined as $c^F_T(V') = c_T(V') + 2con_T(V')$. By the definition, the full local cost of $V(G)$ is equal to the cost of the quasi-tour (plus 2).

Intuitively, $c^F_T(V')$ captures the cost of the quasi-tour restricted to $V'$: it includes the cost of edges and the cost of added connected components. Note that now for two disjoint sets $V_1, V_2$ we have $c^F_T(V_1 \cup V_2) \geq c^F_T(V_1) + c^F_T(V_2)$ since $V_1 \cup V_2$ could contain more connected components than $V_1, V_2$ together. If we know that the total cost of the quasi-tour is small, then $c^F_T(V)$ is small (less than $61m + k$). We can use this to infer that the sum of the local full costs of all gadgets is small.

The high-level idea of the proof is the following: we will use roughly the same partition of $V(G)$ into sets as in the proof of Lemma 1. For each set, we will give a lower bound on its full local cost for any quasi-tour, which will be equal to what the tour we constructed in Lemma 1 pays. If a given quasi-tour behaves differently its local cost will be higher. The difference between the actual local cost and the lower bound is called the credit of that part of the graph. We construct an assignment for $I_2$ and show that the total sum of credits is higher than the number of unsatisfied equations. But using the reasoning of the previous paragraph, the total sum of credits will be at most $k$.

Proof. We are going to prove a slightly stronger statement and show that if there exists a quasi-tour in $G_S$ with cost $61m + k - 2$, then, there exists an assignment leaving at most $k$ equations unsatisfied. Recall that the existence of a tour in $G_S$ with cost $C$ implies the existence of a quasi-tour in $G_S$ with cost at most $C$.

We may assume that simple edges are contained only once in $E_T$ due to the following preprocessing step: If $E_T$ contains two copies of the same simple edge, we remove them without increasing the cost, since the number of components can only increase by one.

In the following, given a quasi-tour $E_T$ in $G_S$, we are going to define an assignment $\phi_T$ and analyze the number of satisfied equations by $\phi_T$ compared to the cost of the quasi-tour.

The general idea is that each vertex of $G_S$ that corresponds to a variable of $I_2$ has exactly two forced and exactly two simple edges incident to it. If the forced edges are used once each, the variable is called honest. We set it to 1 if the simple edges are both used once and to 0 otherwise. It is not hard to see that, because simple cycle edges connect vertices that represent the variables, this procedure will satisfy all cycle equations involving honest variables. We then argue that if other equations are unsatisfied the tour is also paying extra, and the same happens if a variable is dishonest.

Let us give more details. First, we concentrate on the assignment for checker variables.

Assignment for Checker Variables

Let us consider the following equations with two variables $x^{u}_{i-1} \oplus x^{u}_i = 0$, $x^{u}_i \oplus x^{u}_{i+1} = 0$, $x^{n}_{j-1} \oplus x^{n}_j = 0$, $x^{n}_j \oplus x^{n}_{j+1} = 0$ and $x^{u}_i \oplus x^{n}_j = 1$. We are going to analyze the cost of a quasi-tour traversing the gadget displayed in Figure 2(a) and define an assignment according
to $E_T$. Let us first assume that our quasi-tour is honest, that is, the underlying quasi-tour traverses forced edges only once.

**Honest tours:** For $x \in \{x_u^i, x_e^i\}$, we set $x = 1$ if the quasi-tour traverses both simple edges incident on $x$ and $x = 0$, otherwise. Since we removed all copies of the same simple edge, we may assume that cycle equations are always satisfied. If the tour uses $x_{i-1}^u - x_i^u - F x_j^b - F x_i^u - x_{i+1}^u$, we get $x_{i-1}^u = x_{i+1}^u = 1$, $x_{j-1}^b = x_{j+1}^b = 0$ and 5 satisfied equations. Given $x_{j-1}^b - x_{j+1}^b - F x_j^e - F x_{j-1}^e - x_{j+1}^f$, we obtain 5 satisfied equations as well. Let us define $V_i^p := \{x_i^u, x_i^e\}$. Notice that in both cases, we have local cost $c_T^F(V_i^p) = 5$. We claim that $c_T^F(V_i^p) \geq 5$ for a valid quasi-tour. In order to obtain a valid quasi-tour, we need to traverse both forced edges in $G_F^i$ and use at least two simple edges, as otherwise, it implies $c_T^F(G_F^i) \geq 6$. Given a quasi-tour $E_T$, we introduce a local credit function defined by $c_T(V_i^p) = c_T^F(V_i^p) - 5$. If $x_i^u - F x_j^b - F x_i^u$ forms a connected component, we get 4 satisfied equations and $c_T(V_i^p) = 1$, which is sufficient to pay for the unsatisfied equation $x_i^u \oplus x_j^b = 1$. On the other hand, assuming $x_{i-1}^u = x_{i+1}^u = 1$ and $x_{j-1}^b = x_{j+1}^b = 1$, we get $c_T(V_i^p) = 1$ and 1 unsatisfied equation.

**Dishonest tours:** We are going to analyze quasi-tours, which are using one of the forced edges twice. By setting $x_i^u \neq x_i^e$, we are able to find an assignment that always satisfies $x_i^u \oplus x_j^b = 1$ and two other equations out of the five that involve these dishonest variables. The local cost in this case is at least 7. Hence, the credit $c_T(V_i^p) = 2$ is sufficient to pay for the two unsatisfied equations.

**Assignment for Contact Variables**

Again, we will distinguish between honest tours (which use forced edges exactly once) and dishonest tours. This time we are interested in seven equations: the size-three equation $x \oplus y \oplus z = 0$ and the six cycle equations containing the three contacts.

Observe that the local cost of $V_j^{3S} := \{x', x^j, x, y^j, y, z', z^j, z, e^j, e^j\}$ is at least 15.5. The local edge cost of any quasi-tour is 9.5 for the forced edges. For each component $\{\gamma, \gamma'^j, \gamma'^r\}$ with $\gamma \in \{x, y, z\}$, we need to pay at least 2 more because there are two vertices with odd degree ($\gamma'^j, \gamma'^r$) and we also need to connect the component to the rest of the graph (otherwise the component already costs 2 more). Let us define the credit of $V_j^{3S}$ with respect to $E_T$ by $c_T = c_T^F(V_j^{3S}) - 15.5$.

**Honest tours:** For each $\gamma \in \{x, y, z\}$, we set $\gamma = 1$ if the tour uses both simple edges incident on $\gamma$ and 0, otherwise. Notice that in the case $(x + y + z = b)$ with $b \in \{0, 2\}$, this satisfies all seven equations and the tour has local cost at least $c_T^F(V_j^{3S}) = 15.5$.

Case $(x = y = z = 1)$: The assignment now failed to satisfy the size-three equation, so we need to prove that the quasi-tour has local cost at least 16.5. Since all vertices are balanced with respect to $E_T$, the quasi-tour has to use at least one edge incident on $e^j$ and $e^j$ beside $\{s, e^j\}_F$ and $\{s, e^j\}_F$. If the quasi-tour takes $\{e^j, \gamma^\alpha\}$ for $\gamma \in \{x, y, z\}$ and all $\alpha \in \{r, l\}$, since all simple edges incident on $x, y, z$ are used, we get at total cost of at least 16.5, which gives a credit of 1.

Case $(x + y + z = 1)$: Without loss of generality, we assume that $x = y = 0 \neq z$
holds. Again, only the size-three equation is unsatisfied, so we must show that the local cost is at least 16.5. We will discuss two subcases. (i) There is a connected component 
\[ \delta - F, \delta^l - \delta^r - F, \delta \] for some \( \delta \in \{x, y\} \). We obtain that \( c^T_F(\{\delta, \delta^l, \delta^r\}) \geq 6 \) and therefore, a lower bound on the total cost of 16.5. (ii) Since we may assume that \( x^r, x^l, y^r \) and \( y^l \) are balanced with respect to \( E_T \), we have that \( \{e^\alpha_j, \gamma^\alpha\} \in E_T \) for all \( \alpha \in \{r, l\} \) and \( \gamma \in \{x, y\} \). Because \( e^\alpha_j \) are also balanced, we obtain \( \{e^\alpha_j, z^\alpha\} \in E_T \) for all \( \alpha \in \{r, l\} \), which implies a total cost of 16.5.

**Dishonest tours:** Let us assume that the quasi-tour uses both of the forced edges \( \{\gamma^r, \gamma\} \) and \( \{\gamma^l, \gamma\} \) for some \( \gamma \in \{x, y, z\} \) twice. We delete both copies and add \( \{\gamma^r, \gamma^l\} \) instead which reduces the cost of the quasi-tour. Hence, we may assume that only one of the two incident forced edges is used twice.

First, observe that if all forced edges were used once, then there would be eight vertices in the gadget with odd degree: \( x^r, x^l, y^r, y^l, z^r, z^l, e^r_j, e^l_j \). If exactly one forced edge is used twice, then seven of these vertices have odd degree. Thus, it is impossible for the tour to make the degrees of all seven even using only the simple edges that connect them. We can therefore assume that if a forced edge is used twice, there exists another forced edge used twice.

We will now take cases, depending on how many of the vertices \( x, y, z \) are incident on forced edges used twice. Note that if one of the forced edges incident on \( x \) is used twice, then exactly one of the simple edges incident on \( x \) is used once. So, first suppose all three of \( x, y, z \) have forced edges used twice. The local cost from forced edges is at least 14. Furthermore, there are three vertices of the form \( \gamma^\alpha \), for \( \gamma \in \{x, y, z\} \) and \( \alpha \in \{l, r\} \) with odd degree. These have no simple edges connecting them, thus the quasi-tour will use three simple edges to balance their degrees. Finally, the used simple edges incident on \( x, y, z \) each contribute 0.5 to the local cost. Thus, the total local cost is at least 18.5, giving us a credit of 3. It is not hard to see that there is always an assignment satisfying four out of the seven affected equations, so this case is done.

Second, suppose exactly two of \( x, y, z \) have incident forced edges used twice, say, \( x, y \). For \( z \), we select the honest assignment (1 if the incident simple edges are used, 0 otherwise) and this satisfies the cycle equations for this variable. We can select assignments for \( x, y \) that satisfy three of the remaining five equations, so we need to show that the cost in this case is at least 17.5. The cost of forced edges is at least 12.5, and the cost of simple edges incident on \( x, y \) adds 1 to the local cost. One of the vertices \( x^r, x^l \) and one of \( y^r, y^l \) have odd degree, therefore the cost uses two simple edges to balance them. Finally, the vertices \( z^l, z^r \) have odd degree. If two simple edges incident to them are used, we have a total local cost of 17.5. If the edge connecting them is used, then the two simple edges incident on \( z \) must be used, again pushing the local cost to 17.5.

Finally, suppose only \( x \) has an incident forced edge used twice. By the parity argument given above, this means that one of the forced edges incident on \( s \) is used twice. We can satisfy the cycle equations for \( y, z \) by giving them their honest assignment, and out of the three remaining equations some assignment to \( x \) satisfies two. Therefore, we need to
show that the cost is at least 16.5. The local cost from forced edges is 11.25 and the simple edge incident on \( x \) contributes 0.5. Also, at least one simple edge incident on \( x^l \) or \( x^r \) is used, since one of them has odd degree. For \( y^l, y^r \), either two simple edges are used, or if the edge connecting them is used the simple edges incident on \( y \) contribute 1 more. With similar reasoning for \( z^l, z^r \), we get that the total local cost is at least 16.75.

Let us now conclude our analysis. Consider the following partition of \( V \): we have a singleton set \{s\}, 9m sets of size 2 containing the matching edge gadgets and \( m \) sets of size 11 containing the gadgets for size-three equations (except \( s \)). The sum of their local costs is at most \( c_F^T(V) \leq 61m + k \). But the sum of their local costs is (using the preceding analysis) equal to \( 61m + \sum c_T(V_i) \). Thus, the sum of all credits is at most \( k \). Since we have already argued that the sum of all credits is enough to cover all equations unsatisfied by our assignment, this concludes the proof.

\[ \square \]

**Proof of Theorem 4**. We are given an instance \( I_1 \) of the MAX-E3LIN2 problem with \( \nu \) variables and \( m \) equations. For all \( \epsilon > 0 \), there exists a \( k \) such that if we repeat each equation \( k \) time we get an instance \( I_1^{(k)} \) with \( m' = km \) equations and \( \nu \) variables such that \( 2(\nu + 1)/m' \leq \delta \).

Then, from \( I_1^{(k)} \), we generate an instance \( I_2 \) of the Hybrid problem and the corresponding graph \( G_S \). Due to Lemmata 1, 2 and Theorem 3, we know that for all \( \epsilon > 0 \), it is \( \text{NP} \)-hard to tell whether there is a tour with cost at most \( 61m' + 2\nu + 2 + \epsilon \cdot m' \leq 61 \cdot m' + (\delta + \epsilon) m' \) or all tours have cost at least \( 61m' + (0.5 - \epsilon)m' - 2 \geq 61.5 \cdot m' - \epsilon \cdot m' - \delta \cdot m' \). The ratio between these two cases can get arbitrarily close to \( 123/122 \) by appropriate choices for \( \epsilon, \delta \).

\[ \square \]

6 ATSP

In this section, we prove the following theorem.

**Theorem 5.** It is \( \text{NP} \)-hard to approximate the ATSP to within any constant approximation ratio less than \( 75/74 \).

6.1 Construction

Let us describe the construction that encodes an instance \( I_2 \) of the Hybrid problem into an instance of the ATSP. Again, it will be useful to have the ability to force some edges to be used, that is, we would like to have bidirected forced edges. A bidirected forced edge of weight \( w \) between two vertices \( x \) and \( y \) will be created in a similar way as undirected forced edges in the previous section: construct \( L - 1 \) new vertices and connect \( x \) to \( y \) through these new vertices, making a bidirected path with all edges having weight \( w/L \).
It is not hard to see that without loss of generality we may assume that all edges of the path are used in at least one direction, though we should note that the direction is not prescribed. In the remainder, we denote a directed forced edge consisting of vertices \( x \) and \( y \) by \( (x, y)_{\text{F}} \), or \( x \rightarrow_{\text{F}} y \).

Let \( I_2 \) consist of the collection \( \{W_i\}_{i=1}^{n} \) of bi-wheels. Recall that the bi-wheel consists of two cycles and a perfect matching between their checkers. Let \( \{x_i^u, x_j^u\}_{i=1}^{n} \) be the associated set of variables of \( W_i \). We write \( u(i) \) to denote the function which, given the index of a checker variable \( x_i^u \) returns the index \( j \) of the checker variable \( x_j^u \) to which it is matched (that is, the function \( u \) is a permutation function encoding the matching). We write \( n(i) \) to denote the inverse function \( u^{-1}(i) \).

Now, for each bi-wheel \( W_i \), we are going to construct the corresponding directed graph \( C_{\text{n}}^p \) as follows. First, construct a vertex for each checker variable of the wheel. For each matching equation \( x_i^u \oplus x_j^n = 1 \), we create a bidirected forced edge \( \{x_i^u, x_j^n\}_F \) with \( w(\{x_i^u, x_j^n\}_F) = 2 \).

For each contact variable \( x_k \), we create two corresponding vertices \( x_k^r \) and \( x_k^l \), which are joined by the bidirected forced edge \( \{x_k^r, x_k^l\}_F \) with \( w(\{x_k^r, x_k^l\}_F) = 1 \).

Next, we will construct two directed cycles \( C_{\text{u}}^p \) and \( C_{\text{n}}^p \). Note that we are doing arithmetic on the cycle indices here, so the index \( z+1 \) should be read as equal to 1. For \( C_{\text{u}}^p \), for any two consecutive checker vertices \( x_i^u, x_{i+1}^u \) on the un-negated side of the bi-wheel, we add a simple directed edge \( x_{u(i)}^u \rightarrow x_{i+1}^u \). If the checker \( x_i^u \) is followed by a contact \( x_{i+1}^u \) in the cycle, then we add two simple directed edges \( x_{u(i)}^u \rightarrow x_{i+1}^u \) and \( x_{i+1}^u \rightarrow x_{i+2}^u \). Observe that by traversing the simple edges we have just added, the forced matching edges in the direction \( x_i^u \rightarrow_{F} x_{u(i)}^u \) and the forced contact edges for the un-negated part in the direction \( x_i^u \rightarrow_{F} x_{i+1}^u \) we obtain a cycle that covers all checkers and all the contacts of the un-negated part.

We now add simple edges to create a second cycle \( C_{\text{n}}^p \). This cycle will require using the forced matching edges in the opposite direction and, thus, truth assignments will be encoded by the direction of traversal of these edges. First, for any two consecutive checker vertices \( x_i^n, x_{i+1}^n \) on the un-negated side of the bi-wheel, we add the simple directed edge \( x_{n(i)}^n \rightarrow x_{i+1}^n \). Then, if the checker \( x_i^n \) is followed by a contact \( x_{i+1}^n \) in the cycle then we add the simple directed edges \( x_{n(i)}^n \rightarrow x_{i+1}^n \) and \( x_{i+1}^n \rightarrow x_{i+2}^n \). Now by traversing the edges we have just added, the forced matching edges in the direction \( x_i^n \rightarrow_{F} x_{n(i)}^n \) and the forced contact edges for the negated part in the direction \( x_i^n \rightarrow_{F} x_{i+1}^n \), we obtain a cycle that covers all checkers and all the contacts of the negated part, that is, a cycle of direction opposite to \( C_{\text{n}}^p \).

What is left is to encode the equations of size three. Again, we have a central vertex \( s \) that is connected to gadgets simulating equations with three variables. For every equation with three variables, we create the gadget displayed in Figure 11(b), which is a variant of the gadget used by Papadimitriou and Vempala [PV00]. Let us assume that the \( j \)-th equation with three variables in \( I_3 \) is of the form \( x \oplus y \oplus z = 1 \). This equation is simulated by \( G_{3}^{j} \). The vertices used are the contact vertices \( \gamma \alpha, \gamma \in \{x, y, z\}, \alpha \in \{r, l\} \), which we have already introduced, as well as the vertices \( \{s_j, t_j, e_i^j \mid i \in [3]\} \). For notational
simplicity, we define \( V_{3A}^j = \{ s_j, t_j, e_j^i, \gamma^\alpha | i \in [3], \gamma \in \{ x, y, z \}, \alpha \in \{ r, l \} \} \). All directed non-forced edges are simple. The vertices \( s_j \) and \( t_j \) are connected to \( s \) by forced edges with \( w((s, s_j)_F) = w((t_j, s)_F) = \lambda \), where \( \lambda > 0 \) is a small fixed constant. To simplify things, we also force them to be used in the displayed direction by deleting the edges that make up the path of the opposite direction. This is the whole description of the graph \( G_A \).

### 6.2 Assignment to Tour

We are going to construct a tour in \( G_A \) given an assignment to the variables of \( I_2 \) and prove the following lemma.

**Lemma 3.** Given an instance \( I_2 \) of the Hybrid problem with \( \nu \) bi-wheels and an assignment that leaves \( k \) equations in \( I_2 \) unsatisfied, then, there exists a tour in \( G_A \) with cost at most \( 37m + 5\nu + 2m\lambda + 2\nu\lambda + k \).

Before we proceed, let us again give a definition for a local edge cost function. Let \( G \) be an edge-weighted digraph and \( E_T \) a multi-set of edges of \( E(G) \) that defines a tour. Consider a set \( V' \subseteq V(G) \). The local edge cost of the set \( V' \) is then defined as

\[
c_T(V') = \sum_{u \in V'} \sum_{(u,v) \in E_T} w((u,v))
\]

In words, for each vertex in \( V' \) we count the total weight of its outgoing edges used in the quasi-tour (including multiplicities). Thus, that this sum contains the full weight for edges with their source in \( V' \), regardless of where their other endpoint is. Also note that again for two sets \( V_1, V_2 \) we have \( c_T(V_1 \cup V_2) \leq c_T(V_1) + c_T(V_2) \) (with equality for disjoint sets) and that \( c_T(V) = \sum_{e \in E_T} w(e) \).

**Proof of Lemma 3** Let \( W_p \) be a bi-wheel with variables \( \{ x_i^n, x_i^p \} \). Given an assignment to the variables of \( I_2 \), due to Theorem 3 we may assume that either \( x_i^n = 1 \neq x_j^n \) for all \( i, j \in [z] \) or \( x_i^n = 0 \neq x_j^n \) for all \( i, j \in [z] \). We traverse the cycle \( C_u^p \) if \( x_i^p = 1 \) and the cycle \( C_v^p \) otherwise. This creates \( \nu \) strongly connected components. Each contains all the checkers of a bi-wheel and the contacts from one side.

For each matching edge gadget, the local edge cost is 3. We pay two for the forced edge and 1 for the outgoing simple edge. We will account for the cost of edges incident on contacts when we analyze the size-three equation gadget below.

Let us describe the part of the tour traversing the graph \( G_{3A}^j \), which simulates \( x \oplus y \oplus z = 1 \). Recall that if \( x \) is set to true in the assignment we have traversed the bi-wheel gadgets in such a way that the forced edge \( x^r \rightarrow_F x^l \) is used, and the simple edge coming out of \( x^l \) is used.

According to the assignment to \( x, y, \) and \( z \), we traverse \( G_{3A}^j \) as follows:

**Case** \( (x + y + z = 1) \): Let us assume that \( z = y = 0 \neq x \) holds. Then, we use \( s \rightarrow_F s_j \rightarrow e_j^2 \rightarrow y^l \rightarrow_F y^r \rightarrow e_j^3 \rightarrow z^l \rightarrow_F z^r \rightarrow e_j^4 \rightarrow t_j \rightarrow_F s \). The cost is \( 3 + \lambda \) for the forced edges, 6 for the simple edges inside the gadget, plus 1 for the simple edge going out of \( x^l \). Total local edge cost cost: \( c_T(V_{3A}^j) = 10 + \lambda \).
Case \( (x + y + z = 3) \): Then, we use \( s \rightarrow_F s_j \rightarrow e^1_j \rightarrow e^2_j \rightarrow t_j \rightarrow_F s \). Again we pay \( 3 + \lambda \) for the forced edges, 4 for the simple edges inside the gadget and 3 for the outgoing edges incident on \( x^l, y^l, z^l \). Total local edge cost: \( c_T(V^{3A}_j) = \lambda + 10 \).

Case \( (x + y + z = 2) \): Let us assume that \( x = y = 1 \neq z \) holds. Then, we use \( s \rightarrow_F s_j \rightarrow e^3_j \rightarrow z^l \rightarrow_F z^r \rightarrow e^1_j \rightarrow e^2_j \rightarrow e^3_j \rightarrow t_j \rightarrow_F s \) with total local edge cost \( c_T(V^{3A}_j) = \lambda + 11 \).

Case \( (x + y + z = 0) \): We use \( s \rightarrow_F s_j \rightarrow e^2_j \rightarrow y^l \rightarrow_F y^r \rightarrow e^3_j \rightarrow z^l \rightarrow_F z^r \rightarrow e^1_j \rightarrow x^l \rightarrow_F x^r \rightarrow e^2_j \rightarrow t_j \rightarrow_F s \) with \( c_T(V^{3A}_j) = \lambda + 11 \).

The total edge cost of the quasi-tour we constructed is \( 3 \cdot 9m + (10 + 2\lambda)m + k = 37m + 2\lambda m + k \). We have at most \( \nu + 1 \) strongly connected components: one for each bi-wheel and one containing \( s \). A component representing a bi-wheel can be connected to \( s \) as follows: let \( x^l, x^r \) be two contact vertices in the component. Add one copy of each edge from the cycle \( s \rightarrow_F s_j \rightarrow e^1_j \rightarrow x^l \rightarrow_F x^r \rightarrow e^2_j \rightarrow t_j \rightarrow_F s \). This increases the cost by \( 5 + 2\lambda \) but decreases the number of components by one.

\[ \square \]

### 6.3 Tour to Assignment

In this section, we are going to prove the other direction of the reduction.

**Lemma 4.** If there is a tour with cost \( 37 \cdot m + k + 2\lambda \cdot m \), then, there is an assignment that leaves at most \( k \) equations unsatisfied.

**Proof.** Given a tour \( E_T \) in \( G_A \), we are going to define an assignment to checker and contact variables. As in Lemma 2, we will show that any tour must locally spend on each gadget at least the same amount as the tour we constructed in Lemma 3. If the tour spends more, we use that credit to satisfy possible unsatisfied equations.

#### Assignment for Checker Variables

Let us consider the following equations with two variables \( x^u_i \oplus x^u_{i+1} = 0, x^u_{i-1} \oplus x^u_i = 0, x^u_i \oplus x^u_{i+1} = 1, x^u_{j-1} \oplus x^u_j = 0 \) and the corresponding situation displayed in Figure 2(b). Since \( E_T \) is a valid tour in \( G_A \), we know that \( \{ x^u_i, x^u_j \} \) is traversed and due to the degree condition, for each \( x \in \{ x^u_i, x^u_j \} \), the tour uses another incident edge \( e \) on \( x \) with \( w(e) \geq 1 \). Therefore, we have that \( c_T(\{ x^u_i, x^u_j \}) \geq 3 \). The credit assigned to a gadget is defined as \( c_T(\{ x^u_i, x^u_j \}) = c_T(\{ x^u_i, x^u_j \}) - 3 \).

Let us define the assignment for \( x^u_i \) and \( x^u_j \). A variable \( x^u_i \) is honestly traversed if either both the simple edge going into \( x^u_i \) is used and the simple edge coming out of \( x^u_i \) is used, or neither of these two edges is used. In the first case, we set \( x^u_i \) to 1, otherwise to 0. Similarly, \( x^u_j \) is honest if both the edge going into \( x^u_j \) and the edge out of \( x^u_j \) are used, and we set it to 1 in the first case and 0 otherwise.

**Honest tours:** First, suppose that both \( x^u_i \) and \( x^u_j \) are honest. We need to show that the credit is at least as high as the number of unsatisfied equations out of the five equations.
that contain them. It is not hard to see that if we have set \( x_i^a \neq x_i^b \) all equations are satisfied. If we have set both to 1, then the forced edge must be used twice, making the local edge cost at least 6, giving a credit of 3, which is more than sufficient.

**Dishonest tours:** If both \( x_i^a \) and \( x_j^b \) are dishonest the tour must be using the forced edge in both directions. Thus, the local cost is 5 or more, giving a credit of 2. There is always an assignment that satisfies three out of the five equations, so this case is done. If one of them is dishonest, the other must be set to 1 to ensure strong connectivity. Thus, there are two simple edges used leaving the gadget, making the local cost 4 (perhaps the same edge is used twice). We can set the honest variable to 1 (satisfying its two cycle equations), and the other to 0, leaving at most one equation unsatisfied.

**Assignment for Contact Variables**

First, we note that for any valid tour, we have \( c_T(V_{jA}^{3A}) \geq 10 + \lambda \). This is because the two forced edges of weight \( \lambda \) must be used, and there exist 10 vertices in the gadget for which all outgoing edges have weight 1. Let us define the credit \( cr_T(V_{jA}^{3A}) = c_T(V_{jA}^{3A}) - (10 + \lambda) \).

**Honest Traversals:** We assume that the underlying tour is honest, that is, forced edges are traversed only in one direction. We set \( x \) to 1 if the forced edge is in the direction \( x^r \rightarrow F x^l \) and 0 otherwise. In the first case we know that the simple edges going into \( x^r \) and out of \( x^l \) are used. In the second, the edges \( e^1_j \rightarrow x^l \) and \( x^r \rightarrow e^2_j \) are used. We do similarly for \( y, z \).

We are interested in the equation \( x \oplus y \oplus z = 1 \) and the six cycle equations involving \( x, y, z \). The assignment we pick for honest variables satisfies the cycle equations, so if it also satisfies the size-three equation we are done. If not, we have to prove that the tour pays at least \( 11 + \lambda \).

**Case** \( x = y = z = 0 \): Due to our assumption, we know that \( e^2_j \rightarrow y^l \rightarrow F y^r \rightarrow e^3_j \rightarrow z^l \rightarrow_F z^r \rightarrow e^3_j \rightarrow x^l \rightarrow_F x^r \rightarrow e^2_j \) is a part of the tour. Since \( E_T \) is a tour, there exists a vertex in \( V_{jA}^{3A}\setminus\{s_j, t_j\} \) that is visited twice and we get \( c_T(V_{jA}^{3A}) \geq 11 + \lambda \). Thus, we can spend the credit \( cr_T(V_{jA}^{3A}) \geq 1 \) on the unsatisfied equation \( x \oplus y \oplus z = 1 \).

**Case** \( x + y + z = 2 \): Without loss of generality, let us assume that \( x = y = 1 \neq z \) holds. Then, we know that \( e^3_j \rightarrow z^l \rightarrow_F z^r \rightarrow e^1_j \) is a part of the tour. But, this implies that there is a vertex in \( V(G_{jA}^{3A}) \) that is visited twice. Hence, we have that \( cr_T(V_{jA}^{3A}) \geq 1 \).

**Dishonest Traversals:** Consider the situation, in which some forced edges \( \{\gamma^l, \gamma^r\}_F \) are traversed in both directions for some variables \( \gamma \in \{x, y, z\} \). For the honest variables, we set them to the appropriate value as before, and this satisfies their cycle equations. Observe now that if a forced edge \( \gamma^l \rightarrow_F \gamma^r \) is also used in the opposite direction, then there must be another edge used to leave the set \( \{\gamma^l, \gamma^r\} \). Thus the local edge cost of this set is at least 3. It follows that the credit we have for the gadget is at least as large as the number of dishonest variables. We can give appropriate values to them so each satisfies one cycle equation and the size-three equation is satisfied. Thus, the number of unsatisfied equations is not larger than our credit.

In summary, for every tour \( E_T \) in \( G_A \), we can find an assignment to the variables of \( I_2 \) such that all unsatisfied equations are paid by the credit induced by \( E_T \).
We are ready to give the proof of Theorem 5.

Proof of Theorem 5. We are again given an instance $I_1$ of the MAX-E3LIN2 problem with $\nu$ variables and $m$ equations. For all $\delta > 0$, there exists a $k$ such that if we repeat each equation $k$ time we get an instance $I_1^{(k)}$ with $m' = km$ equations and $\nu$ variables such that $\nu/m' \leq \delta$.

Then, from $I_1^{(k)}$, we generate an instance $I_2$ of the Hybrid problem and the corresponding directed graph $G_A$. Due to Lemmata 3, 4 and Theorem 3, we know that for all $\epsilon > 0$, it is NP-hard to tell whether there is a tour with cost at most $37m' + 5\nu + 2m(\nu + \lambda) + \epsilon \cdot m' \leq 37 \cdot m' + \epsilon' m'$ or all tours have cost at least $37m' + (0.5 - \epsilon)m' \geq 37.5 \cdot m' - \epsilon' \cdot m'$, for some $\epsilon'$ depending only on $\epsilon, \delta, \lambda$. The ratio between these two cases can get arbitrarily close to $75/74$ by appropriate choices for $\epsilon, \delta, \lambda$.

7 Concluding Remarks

In this paper, we proved that it is hard to approximate the ATSP and the TSP within any constant factor less than $75/74$ and $123/122$, respectively. The proof method required essentially new ideas and constructions from the ones used before in that context. Since the best known upper bound on the approximability is $O(\log n/\log \log n)$ for ATSP and $3/2$ for TSP, there is certainly room for improvements. Especially, in the asymmetric version of the TSP, there is a large gap between the approximation lower and upper bound, and it remains a major open problem on the existence of an efficient constant factor approximation algorithm for that problem. Furthermore, it would be nice to investigate if some of the ideas of this paper, and in particular the bi-wheel amplifiers, can be used to offer improved hardness results for other optimization problems, such as the Steiner Tree problem.

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