The Directed Minimum Latency Problem

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Abstract

We study the *directed minimum latency problem*: given an n-vertex asymmetric metric (V, d) with a root vertex $r \in V$, find a spanning path originating at r that minimizes the sum of latencies at all vertices (the latency of any vertex $v \in V$ is the distance from r to v along the path). This problem has been well-studied on symmetric metrics, and the best known approximation guarantee is 3.59 [3]. For any $\frac{1}{\log n} < \epsilon < 1$, we give an $n^{O(1/\epsilon)}$ time algorithm for directed latency that achieves an approximation ratio of $O(\rho \cdot \frac{n^\epsilon}{\epsilon^3})$, where ρ is the integrality gap of an LP relaxation for the *asymmetric traveling salesman path* problem [13, 5]. We prove an upper bound $\rho = O(\sqrt{n})$, which implies (for any fixed $\epsilon > 0$) a polynomial time $O(n^{1/2+\epsilon})$-approximation algorithm for directed latency.

In the special case of metrics induced by shortest-paths in an unweighted directed graph, we give an $O(\log^2 n)$ approximation algorithm. As a consequence, we also obtain an $O(\log^2 n)$ approximation algorithm for minimizing the weighted completion time in *no-wait permutation flowshop scheduling*. We note that even in unweighted directed graphs, the directed latency problem is at least as hard to approximate as the well-studied *asymmetric traveling salesman problem*, for which the best known approximation guarantee is $O(\log n)$.

1 Introduction

The minimum latency problem [17, 6, 14, 2] is a variant of the basic traveling salesman problem, where there is a metric with a specified root vertex r, and the goal is to find a spanning path starting from r that minimizes the sum of arrival times at all vertices (it is also known as the *deliveryman problem* or *traveling repairman problem*). This problem can model the traveling salesman problem, and hence is NP-complete. To the best of our knowledge, all previous work has focused on symmetric metrics– the first constant-factor approximation algorithm was in Blum et al. [2], and the currently best known approximation ratio is 3.59 due to Chaudhuri et al. [3]. In this paper, we consider the minimum latency problem on *asymmetric metrics*.

Network design problems on directed graphs are often much harder to approximate than their undirected counterparts– the traveling salesman and Steiner tree problems are well known examples. The currently best known approximation ratio for the asymmetric traveling salesman problem (ATSP) is $O(\log n)$ [9, 7], and improving this bound is an important open question. On the other hand, there is a 1.5-approximation algorithm for the symmetric TSP.

The *orienteering* problem is closely related to the minimum latency problem that we consider– given a metric with a length bound, the goal is to find a bounded-length path between two specified vertices that visits the maximum number of vertices. Blum et al. [1] gave the first constant factor approximation for the undirected version of this problem. Recently, Chekuri et al. [4] and the

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authors [15] independently gave $O(\log^2 n)$ approximation algorithms for the directed orienteering problem.

1.1 Problem Definition

We represent an asymmetric metric by (V, d), where V is the vertex set (with $|V| = n$) and $d : V \times V \to \mathbb{R}_+$ is a distance function satisfying the triangle inequality. For a directed path (or tour) π and vertices u, v, $d^\pi(u, v)$ denotes the distance from u to v along π; if v is not reachable from u along π, then $d^\pi(u, v) = \infty$. The directed minimum latency problem is defined as follows: given an asymmetric metric (V, d) and a root vertex $r \in V$, find a spanning *path* π originating at r that minimizes $\sum_{v \in V} d^\pi(r, v)$; the quantity $d^\pi(r, v)$ is the *latency* of vertex v in path π. Another possible definition of this problem would require a *tour* covering all vertices, where the latency of the root r is defined to be the distance required to return to r (i.e. the total tour length); note that in the previous definition of directed latency, the latency of r is zero. The approximability of both these versions of directed latency are related as below (the proof is deferred to Appendix A).

Theorem 1 *The approximability of the path-version and tour-version of directed latency are within a factor* 4 *of each other.*

In this paper, we work with the path version of directed latency.

For a directed graph $G = (V, E)$ and any $S \subseteq V$, we denote by $\delta^+(S) = \{(u, v) \in E \mid u \in S,\ v \notin S\}$ the arcs leaving set S, and $\delta^-(S) = \{(u, v) \in E \mid u \notin S,\ v \in S\}$ the arcs entering set S. When dealing with asymmetric metrics, the edge set E is assumed to be $V \times V$ unless mentioned otherwise. Given an asymmetric metric and a special vertex r, an r-path (resp. r-tour) is any directed path (resp. tour) originating at r.

Asymmetric Traveling Salesman Path (ATSP-path). The following problem is closely related to the directed latency problem. In ATSP-path, we are given a directed metric (V, d) and specified start and end vertices $s, t \in V$. The goal is to compute the minimum length $s - t$ path that visits all the vertices. It is easy to see that this problem is at least as hard to approximate as the ATSP (tour-version, where $s = t$). Lam and Newmann [13] were the first to consider this problem, and they gave an $O(\sqrt{n})$ approximation based on the Frieze et al. [9] algorithm for ATSP. This was improved to $O(\log n)$ in Chekuri and Pal [5], which extended the algorithm of Kleinberg and Williamson [12] for ATSP. Subsequently Feige and Singh [7] showed that the approximability of ATSP-tour and ATSP-path are within a constant factor of each other. We are concerned with the following LP relaxation of the ATSP-path problem.

$$
\begin{array}{rll}
 & \min \ \sum_e d_e \cdot x_e & \\
s.t. & & \\
 & x(\delta^+(u)) = x(\delta^-(u)) & \forall u \in V - \{s, t\} \\
 & x(\delta^+(s)) = x(\delta^-(t)) = 1 & \\
(ATSP - path) & x(\delta^-(s)) = x(\delta^+(t)) = 0 & \\
 & x(\delta^-(S)) \geq \frac{2}{3} & \forall \{u\} \subseteq S \subseteq V \setminus \{s\}, \quad \forall u \in V \\
 & x_e \geq 0 & \forall \text{ arcs e}
\end{array}
$$

The most natural LP relaxation for ATSP-path would have a 1 in the right-hand-side of the cut constraints, instead of $\frac{2}{3}$ as above. The above LP further relaxes the cut-constraints, and is still a valid relaxation of the problem. The precise value in the right-hand-side of the cut constraints is not important: we only require it to be some constant strictly between $\frac{1}{2}$ and 1.

1.2 Results and Paper Outline

Our main result is a reduction from the directed latency problem to the *asymmetric traveling salesman path* problem (ATSP-path) [13, 5], where the approximation ratio for directed latency depends on the integrality gap of an LP relaxation for ATSP-path. We give an $n^{O(1/\epsilon)}$ time algorithm for the directed latency problem that achieves an approximation ratio of $O(\rho \cdot \frac{n^\epsilon}{\epsilon^3})$ (for any $\frac{1}{\log n} < \epsilon < 1$), where ρ is the integrality gap of an LP relaxation for the ATSP-path problem. The best upper bound we obtain is $\rho = O(\sqrt{n})$ (Section 3); however we conjecture that $\rho = O(\log n)$. In particular, our result implies a polynomial time $O(n^{1/2+\epsilon})$-approximation algorithm (any fixed $\epsilon > 0$) for directed latency. We study the LP relaxation for ATSP-path in Section 3, and present the algorithm for latency in Section 2. Our algorithm for latency first guesses a sequence of break-points (based on distances along the optimal path) and uses a linear program to obtain an assignment of vertices to segments (the portions between consecutive break-points), then it obtains local paths servicing each segment, and finally stitches these paths across all segments.

We also consider the special case of metrics given by shortest paths in an underlying unweighted directed graph, and obtain an $O(\log^2 n)$ approximation for minimum latency in this case (Section 4). This algorithm is essentially based on using the directed orienteering algorithm [15, 4] within the framework for undirected latency [10]. On the hardness side, we observe that the directed latency problem (even in this 'unweighted' special case) is at least as hard to approximate as ATSP, for which the best known ratio is $O(\log n)$.

We note that ideas from the 'unweighted' case, also imply an $O(\log^2 n)$ approximation algorithm for minimizing weighted completion time in the no-wait permutation flowshop scheduling problem [20, 18]– this can be cast as the latency problem in a special directed metric. We are not aware of any previous results on this problem.

2 The Directed Latency Algorithm

For a given instance of directed latency, let π denote an optimal latency path, $L = d(\pi)$ its length, and Opt its total latency. For any two vertices $u, v \in V$, recall that $d^\pi(u, v)$ denotes the length along path π from u to v; note that $d^\pi(u, v)$ is finite only if u appears before v on path π. The algorithm first guesses the length L (within factor 2) and $l = \lceil \frac{1}{\epsilon} \rceil$ vertices as follows: for each $i = 1, \cdots, l$, v_i is the last vertex on π with $d^\pi(r, v_i) \leq n^{i\epsilon} \frac{L}{n}$. We set $v_0 = r$ and note that v_l is the last vertex visited by π. Let $F = \{v_0, v_1, \cdots, v_l\}$. Consider the linear program (MLP) in Figure 1

Basically this LP requires one unit of flow to be sent from v_i to v_{i+1} (for all $0 \leq i \leq l-1$) such that the total extent to which each vertex u is covered (over all these flows) is at least 1. In addition, the i-th flow is required to have total cost (under the length function d) at most $n^{(i+1)\epsilon} \cdot \frac{L}{n}$. It is easy to see that this LP can be solved in polynomial time for any guess $\{v_i\}_{i=1}^l$. Furthermore the number of possible guesses is $O(n^{1/\epsilon})$, hence we can obtain the optimal solution of (MLP) over all guesses, in $n^{O(1/\epsilon)}$ time.

Claim 1 *The minimum value of (MLP) over all possible guesses of $\{v_i\}_{i=0}^l$ is at most $2n^\epsilon \cdot \mathsf{Opt}$.*

Proof: This claim is straightforward, based on the guesses from an optimal path. Recall that π is the optimal latency path for the given instance. One of the guesses of the vertices $\{v_i\}_{i=0}^l$ satisfies the condition desired of them, namely each v_i (for $i = 1, \cdots, l$) is the last vertex on π with $d^\pi(s, v_i) \leq n^{i\epsilon} \frac{L}{n}$. For each $i = 0, \cdots, l-1$, define O_i to be the set of vertices that are visited between

$$
\begin{array}{lll}
\min & \sum_{i=0}^{l-1} n^{(i+1)\epsilon}\frac{L}{n}(\sum_{u\notin F} y_u^i) & \\
s.t. & & \\
& z^i(\delta^+(u)) = z^i(\delta^-(u)) & \forall u \in V \setminus \{v_i, v_{i+1}\}, \quad \forall i = 0, \cdots, l-1 \\
& z^i(\delta^+(v_i)) = z^i(\delta^-(v_{i+1})) = 1 & \forall i = 0, \cdots, l-1 \\
& z^i(\delta^-(v_i)) = z^i(\delta^+(v_{i+1})) = 0 & \forall i = 0, \cdots, l-1 \\
& z^i(\delta^-(S)) \geq y_u^i & \forall \{u\} \subseteq S \subseteq V \setminus \{v_i\}, \quad \forall u \in V \setminus F, \\
& & \forall i = 0, \cdots, l-1 \\
& \sum_e d_e \cdot z^i(e) \leq n^{(i+1)\epsilon} \cdot \frac{L}{n} & \forall i = 0, \cdots, l-1 \\
& \sum_{i=0}^{l-1} y_u^i \geq 1 & \forall u \in V \setminus F \\
& z^i(e) \geq 0 & \forall \text{ arcs e}, \quad \forall i = 0, \cdots, l-1 \\
& y_u^i \geq 0 & \forall u \in V \setminus F, \quad \forall i = 0, \cdots, l-1
\end{array}
$$

Figure 1: Linear program (MLP) for directed latency.

v_i and v_{i+1} in path π. Let z^i denote the (integral) edge values corresponding to path π restricted to the vertices $O_i \cup \{v_i, v_{i+1}\}$; note that the cost of this flow $d \cdot z^i \leq d^\pi(r, v_{i+1}) \leq n^{(i+1)\epsilon}\frac{L}{n}$. Also set $y_u^i = 1$ for $u \in O_i$ and 0 otherwise, for all $i = 0, \cdots, l-1$. Note that each vertex in $V \setminus \{v_i\}_{i=0}^l$ appears in some set O_i, and each z^i supports unit flow from v_i to all vertices in O_i; hence this (integral) solution $\{z^i, y^i\}_{i=0}^{l-1}$ is feasible for (MLP). The cost of this solution is $\sum_{i=0}^{l-1} n^{(i+1)\epsilon}\frac{L}{n} \cdot |O_i| \leq n^\epsilon L + n^\epsilon \sum_{i=1}^{l-1} n^{i\epsilon}\frac{L}{n} \cdot |O_i| \leq 2n^\epsilon \cdot \mathsf{Opt}$, since $|O_0| \leq n$, $L \leq \mathsf{Opt}$, and each vertex $u \in O_i$ (for $i = 1, \cdots, l-1$) has $d^\pi(r, u) > n^{i\epsilon}\frac{L}{n}$.

We now assume that we have an optimal fractional solution $\{z^i, y^i\}_{i=0}^{l-1}$ to (MLP) over all guesses (with objective value as in Claim 1), and show how to round it to obtain $v_i - v_{i+1}$ paths for each $i = 0, \cdots, l-1$, which when stitched give rise to *one* r-path having a small latency objective. We say that a vertex u is *well-covered* by flow z^i if $y_u^i \geq \frac{1}{4l}$. We partition the vertices $V \setminus F$ into two parts: V_1 consists of those vertices that are well-covered for *at least two* values of $i \in [0, l]$, and V_2 consists of all other vertices. Note that each vertex in V_2 is covered by some flow z^i to the extent at least $\frac{3}{4}$. We first show how to service each of V_1 and V_2 separately using local paths, and then stitch these into a single r-path.

Splitting off: A directed graph is called *Eulerian* if the in-degree equals the out-degree at each vertex. In our proofs, we make use of the following 'splitting-off' theorem for Eulerian digraphs.

Theorem 2 (Frank [8] (Theorem 4.3) and Jackson [11]) *Let $D = (U + r, A)$ be an Eulerian directed multi-graph. For each arc $f = (r, v) \in A$ there exists an arc $e = (u, r) \in A$ so that after replacing arcs e and f by arc (u, v), the directed connectivity between every pair of vertices in U is preserved.*

Note that any vector $\tilde{x}$ of rational edge-capacities that is Eulerian (namely $\tilde{x}(\delta^-(v)) = \tilde{x}(\delta^+(v))$ at all vertices v) corresponds to an Eulerian multi-graph by means of a (sufficiently large) uniform scaling of all arcs. Based on this correspondence, one can use the above splitting-off theorem directly on fractional edge-capacities that are Eulerian.

2.1 Servicing vertices V_1

We partition V_1 into l parts as follows: U_i (for $i = 0, \cdots, l-1$) consists of those vertices of V_1 that are well-covered by z^i but *not* well-covered by any flow z^j for $j > i$. Each set U_i is serviced separately by means of a suitable ATSP solution on $U_i \cup \{v_i\}$ (see Lemma 1): this step requires a bound on the length of back-arcs from U_i-vertices to v_i, which is ensured by the next claim.

Claim 2 *For each vertex $w \in U_i$, $d(w, v_i) \leq 8l \cdot n^{i\epsilon} \frac{L}{n}$.*

Proof: Let $j \leq i-1$ be such that $y^j_w \geq \frac{1}{4l}$; such an index exists by the definition of V_1 and U_i. In other words, arc-capacities z^j support at least $\frac{1}{4l}$ flow from w to v_{j+1}; so $4l \cdot z^j$ supports a unit flow from w to v_{j+1}. Thus $d(w, v_{j+1}) \leq 4l(d \cdot z^j) \leq 4l \cdot n^{(j+1)\epsilon} \frac{L}{n}$. Note that for any $0 \leq k \leq l$, z^k supports a unit flow from v_k to v_{k+1}; hence $d(v_k, v_{k+1}) \leq d \cdot z^k \leq n^{(k+1)\epsilon} \frac{L}{n}$. Now, $d(w, v_i) \leq d(w, v_{j+1}) + \sum_{k=j+1}^{i-1} d(v_k, v_{k+1}) \leq 4l \frac{L}{n} \sum_{k=j}^{i-1} n^{(k+1)\epsilon} \leq 8l \cdot n^{i\epsilon} \frac{L}{n}$.

We now show how all vertices in U_i can be covered by a v_i-tour.

Lemma 1 *For each $i = 0, \cdots, l-1$, there is a poly-time computable v_i-tour covering vertices U_i, of length $O(\frac{1}{\epsilon^2} n^{(i+1)\epsilon} \log n \cdot \frac{L}{n})$.*

Proof: Fix an $i \in \{0, \cdots, l-1\}$; note that the arc capacities z^i are Eulerian at all vertices except v_i and v_{i+1}. Although applying splitting-off (Theorem 2) requires an Eulerian graph, we can apply it to z^i after adding a dummy (v_{i+1}, v_i) arc of capacity 1, and observing that flows from v_i or flows into v_{i+1} do not use the dummy arc. So using Theorem 2 on vertices $V \setminus (U_i \cup \{v_i, v_{i+1}\})$ and triangle inequality, we obtain arc capacities α on the arcs induced by $U_i \cup \{v_i, v_{i+1}\}$ such that: $d \cdot \alpha \leq d \cdot z^i \leq n^{(i+1)\epsilon} \cdot \frac{L}{n}$ and α supports $y^i_u \geq \frac{1}{4l}$ flow from v_i to u and from u to v_{i+1}, for every $u \in U_i$. Below we use B to denote the quantity $n^{(i+1)\epsilon} \cdot \frac{L}{n}$. Consider adding a dummy arc from v_{i+1} to v_i of length B in the induced metric on $U_i \cup \{v_i, v_{i+1}\}$, and set the arc capacity $\alpha(v_{i+1}, v_i)$ on this arc to be 1. Note that α is Eulerian, has total cost at most $2B$, and every non-trivial cut has value at least $\min\{y^i_u : u \in U_i\} \geq \frac{1}{4l}$. So scaling α uniformly by $4l$, we obtain a fractional feasible solution to ATSP on the vertices $U_i \cup \{v_i, v_{i+1}\}$ (in the modified metric), having cost at most $8l \cdot B$. Since the Frieze et al. [9] algorithm computes an integral tour of length at most $O(\log n)$ times any fractional feasible solution (see Williamson [19]), we obtain a v_i-tour τ on the modified metric of length at most $O(l \log n) \cdot B$. Since the dummy (v_{i+1}, v_i) arc has length B, it may be used at most $O(l \log n)$ times in τ. So removing all occurrences of this dummy arc gives a set of $O(l \log n)$ $v_i - v_{i+1}$ paths in the original metric, that together cover U_i. Ignoring vertex v_{i+1} and inserting the direct arc to v_i from the last U_i vertex in each of these paths gives us the desired v_i-tour covering U_i. Finally note that each of the arcs to v_i inserted above has length $O(l \cdot n^{i\epsilon}) \frac{L}{n} = O(l) \cdot B$ (from Claim 2), and the number of arcs inserted is $O(l \log n)$. So the length of this v_i-tour is at most $O(l \log n) \cdot B + O(l^2 \log n) \cdot B = O(\frac{1}{\epsilon^2} n^{(i+1)\epsilon} \log n \cdot \frac{L}{n})$.

2.2 Servicing vertices V_2

We partition vertices in V_2 into $W_0, \cdots, W_{l-1}$, where each W_i contains the vertices in V_2 that are well-covered by z^i. As noted earlier, each vertex $u \in W_i$ in fact has $y^i_u \geq \frac{3}{4} > \frac{2}{3}$. We now consider any particular W_i and obtain a $v_i - v_{i+1}$ path covering the vertices of W_i. Vertices in W_i are covered by a fractional $v_i - v_{i+1}$ path as follows. Splitting off vertices $V \setminus (W_i \cup \{v_i, v_{i+1}\})$ in the fractional solution z^i gives us edge capacities β in the metric induced on $W_i \cup \{v_i, v_{i+1}\}$, such that:

β supports at least $\frac{2}{3}$ flow from v_i to u and from u to v_{i+1} for all $u \in W_i$, and $d \cdot \beta \leq d \cdot z^i$ (this is similar to how arc-capacities α were obtained in Lemma 2.1). Note that β is a fractional feasible solution to the LP relaxation $(ATSP - path)$ for the ATSP-path instance on the metric induced by $W_i \cup \{v_i, v_{i+1}\}$ with start-vertex v_i and end-vertex v_{i+1}. So if ρ denotes the (constructive) integrality gap of $(ATSP - LP)$, we can obtain an integral v_i-v_{i+1} path that spans W_i of length at most $\rho(d \cdot \beta) \leq \rho(d \cdot z^i) \leq \rho n^{(i+1)\epsilon} \frac{L}{n}$. This requires a polynomial time algorithm that computes an integral path of length at most ρ times the LP value; However even a non-constructive proof of integrality gap ρ' implies a constructive integrality gap $\rho = O(\rho' \log n)$, using the algorithm in Chekuri and Pal [5]. So we obtain:

Lemma 2 *For each $i = 0, \cdots, l-1$, there is a poly-time computable v_i-v_{i+1} path covering W_i of length at most $\rho \cdot n^{(i+1)\epsilon} \frac{L}{n}$.*

2.3 Stitching the local paths

We now stitch the v_i-tours that service V_1 (Lemma 1) and the $v_i - v_{i+1}$ paths that service V_2 (Lemma 2), to obtain a single r-path that covers all vertices. For each $i = 0, \cdots, l-1$, let π_i denote the v_i-tour servicing U_i, and let τ_i denote the $v_i - v_{i+1}$ path servicing W_i. The final r-path that the algorithm outputs is the concatenation $\tau^* = \pi_0 \cdot \tau_0 \cdot \pi_1 \cdots \pi_{l-1} \cdot \tau_{l-1}$. From Lemmas 1 and 2, it follows that for all $0 \leq i \leq l-1$, $d(\pi_i) \leq O(\frac{1}{\epsilon^2} \log n) \cdot n^{(i+1)\epsilon} \frac{L}{n}$ and $d(\tau_i) \leq O(\rho) \cdot n^{(i+1)\epsilon} \frac{L}{n}$. So the length of τ^* from r until all vertices of $U_i \cup W_i$ are covered is at most $O(\rho + \frac{1}{\epsilon^2} \log n) \cdot n^{(i+1)\epsilon} \frac{L}{n}$ (as $\epsilon \geq \Omega(\frac{1}{\log n})$). This implies that the total latency of vertices in $U_i \cup W_i$ along path τ^* is at most $O(\rho + \frac{1}{\epsilon^2} \log n) \cdot n^{(i+1)\epsilon} \frac{L}{n} \cdot (|W_i| + |U_i|)$.

Moreover, the contribution of each vertex in U_i (resp., W_i) to the LP objective is at least $\frac{1}{4l} \cdot n^{(i+1)\epsilon} \frac{L}{n}$ (resp., $\frac{3}{4} \cdot n^{(i+1)\epsilon} \frac{L}{n}$). Thus the contribution of $U_i \cup W_i$ to the LP objective is at least $\frac{1}{4l} \cdot n^{(i+1)\epsilon} \frac{L}{n} \cdot (|W_i| + |U_i|)$. Using the upper bound on the latency along τ^* for $U_i \cup W_i$, and summing over all i, we obtain that the total latency along τ^* is at most $O(\frac{1}{\epsilon}\rho + \frac{1}{\epsilon^3} \log n)$ times the optimal value of (MLP). From Claim 1, it now follows that the latency of τ^* is $O(\frac{1}{\epsilon}\rho + \frac{1}{\epsilon^3} \log n) n^{\epsilon} \cdot \mathsf{Opt}$.

Theorem 3 *For any $\Omega(\frac{1}{\log n}) < \epsilon < 1$, there is an $O(\frac{\rho + \log n}{\epsilon^3} \cdot n^{\epsilon})$-approximation algorithm for directed latency, that runs in time $n^{O(1/\epsilon)}$, where ρ is the integrality gap of the LP $(ATSP - path)$. Using $\rho = O(\sqrt{n})$, we have a polynomial time $O(n^{\frac{1}{2}+\epsilon})$ approximation algorithm for any fixed $\epsilon > 0$.*

We prove the bound $\rho = O(\sqrt{n})$ in the next section. A bound of $\rho = O(\log n)$ on the integrality gap of $(ATSP - path)$ would imply that this algorithm is a quasi-polynomial time $O(\log^4 n)$ approximation for directed latency.

Remark: The $(ATSP - path)$ rounding algorithm in Section 3 can be modified slightly to obtain (for any $0 < \delta < 1$), an $(O(n^{\delta} \log n), \lfloor \frac{1}{\delta} \rfloor)$ *bi-criteria approximation* for ATSP-path. This implies the following generalization of Theorem 3.

Corollary 1 *For any $\Omega(\frac{1}{\log n}) < \epsilon < 1$ and $0 < \delta < 1$, there is an $n^{O(1/\epsilon)}$ time algorithm for directed latency, that computes $\lfloor \frac{1}{\delta} \rfloor$ paths covering all vertices, having a total latency of $O(\frac{\log n}{\epsilon^3} \cdot n^{\epsilon+\delta}) \cdot \mathsf{Opt}$, where Opt is the minimum latency of a single path covering all the vertices.*

3 Bounding the integrality gap of ATSP-path

We prove an upper bound of $O(\sqrt{n})$ on the integrality gap ρ of the linear relaxation $(ATSP-path)$ (c.f. Section 1.1). Even for the seemingly stronger LP with 1 in the right-hand-side of the cut constraints, the best bound on the integrality gap we can obtain is $O(\sqrt{n})$: this follows from the cycle-cover based algorithm of Lam and Newmann [13]. As mentioned in Chekuri and Pal [5], it is unclear whether their $O(\log n)$-approximation can be used to bound the integrality gap of such a linear program. In this section, we present a rounding algorithm for the weaker LP $(ATSP-path)$, which shows $\rho = O(\sqrt{n})$. Our algorithm is similar to the ATSP-path algorithm of Lam and Newmann [13] and the ATSP algorithm of Frieze et al. [9]; but it needs some more work as we compare the algorithm's solution against a fractional solution to $(ATSP-path)$.

Let x be any feasible solution to $(ATSP-path)$. We now show how x can be rounded to obtain an integral path spanning all vertices, of total length $O(\sqrt{n})(d \cdot x)$. Let N denote the network corresponding to the directed metric with the *cost* of each arc equal to its metric length, and an extra (t, s) arc of cost 0. The *capacity* of this extra (t, s) arc is set to 3, and all other capacities are set to ∞. The rounding algorithm for x is as follows.

1. Initialize the set of representatives $R \leftarrow V \setminus \{s, t\}$, and the current integral solution $\sigma = \emptyset$.
2. While $R \neq \emptyset$, do:
   (a) Compute a minimum cost circulation $\mathcal{C}$ in $N[R \cup \{s, t\}]$ that sends at least 2 units of flow through each vertex in R (note: $\mathcal{C}$ can be expressed as a sum of cycles).
   (b) Repeatedly extract from $\mathcal{C}$ all cycles that do not use the extra arc (t, s), to obtain circulation $\mathcal{A} \subseteq \mathcal{C}$. Let $R' \subseteq R$ be the set of R-vertices that have degree at least 1 in $\mathcal{A}$.
   (c) Let $\mathcal{B} = \mathcal{C} \setminus \mathcal{A}$; note that $\mathcal{B}$ is Eulerian and each cycle in it uses arc (t, s).
   (d) If $|R'| \geq \sqrt{n}$, do:
      i. Set $\sigma \leftarrow \sigma \cup \mathcal{A}$.
      ii. *Modify* R by dropping all but one R'-vertex from each strong component of $\mathcal{A}$.
   (e) If $|R'| < \sqrt{n}$, do:
      i. Take an Euler tour on $\mathcal{B}$ and remove all (at most 3) occurrences of arc (t, s) to obtain s-t paths P_1, P_2, P_3.
      ii. Restrict each path P_1, P_2, P_3 to vertices in $R \setminus R'$ by short-cutting over R'-vertices, to obtain paths $\tilde{P}_1, \tilde{P}_2, \tilde{P}_3$.
      iii. Take a topological ordering $s = w_1, w_2, \cdots, w_h = t$ of vertices $(R \setminus R') \cup \{s, t\}$ relative to the arcs $\tilde{P}_1 \cup \tilde{P}_2 \cup \tilde{P}_3$.
      iv. Set $\sigma \leftarrow \sigma \cup \{(w_j, w_{j+1}) : 1 \leq j \leq h-1\}$.
      v. Repeat for each vertex $u \in R'$: find an arc $(w, w') \in \sigma$ such that x supports $\frac{1}{6}$ flow from w to u and from u to w', and modify $\sigma \leftarrow (\sigma \setminus (w, w')) \cup \{(w, u), (u, w')\}$.
      vi. Set $R \leftarrow \emptyset$.
3. Output any spanning s-t walk in σ.

We now show the correctness and performance guarantee of the rounding algorithm. We first bound the cost of the circulation obtained in Step 2a during any iteration.

Claim 3 *For any $R \subseteq V \setminus \{s,t\}$, the minimum cost circulation $\mathcal{C}$ computed in step 2a has cost at most $3(d \cdot x)$.*

Proof: The arc values x define a fractional $s-t$ path in network N. Extend x to be a (fractional) circulation by setting $x(t,s) = 1$. We can now apply splitting-off (Theorem 2) on each vertex in $V \setminus R$, to obtain capacities x' in network $N[R \cup \{s,t\}]$, such that every pairwise connectivity is preserved and (by triangle inequality) $d \cdot x' \leq d \cdot x$. Note that the extra (t,s) arc is not modified in the splitting-off steps. So x' supports $\frac{2}{3}$ flow from s to each vertex in R; this implies that $3x'$ is a feasible fractional solution to the circulation instance solved in step 2a (note that $x'(t,s)$ remains 1, so solution $3x'$ satisfies the capacity of arc (t,s)). Finally, note that the linear relaxation for circulation is integral (c.f. Nemhauser and Wolsey [16]). So the minimum cost (integral) circulation computed in step 2a has cost at most $3d \cdot x' \leq 3d \cdot x$.

Note that each time step 2d is executed, $|R|$ decreases by at least $\sqrt{n}/2$ (each strong component in $\mathcal{A}$ has at least 2 vertices); so there are at most $O(\sqrt{n})$ such iterations and the cost of σ due to additions in this step is $O(\sqrt{n})(d \cdot x)$ (using Claim 3). Step 2e is executed at most once (at the end); the next claim shows that this step is well defined and bounds the cost incurred.

Claim 4 *In step 2(e)iii, there exists a topological ordering $w_1, \cdots, w_h$ of $(R \setminus R') \cup \{s,t\}$ w.r.t. arcs $\tilde{P}_1 \cup \tilde{P}_2 \cup \tilde{P}_3$. Furthermore, $\{(w_j, w_{j+1}) : 1 \leq j \leq h-1\} \subseteq \tilde{P}_1 \cup \tilde{P}_2 \cup \tilde{P}_3$.*

Proof: Note that any cycle in $P_1 \cup P_2 \cup P_3$ is a cycle in $\mathcal{B}$ that does not use arc (t,s), which is not possible by the definition of $\mathcal{B}$ (every cycle in $\mathcal{B}$ uses arc (t,s)); so $P_1 \cup P_2 \cup P_3$ is acyclic. It is clear that if $\tilde{P}_1 \cup \tilde{P}_2 \cup \tilde{P}_3$ contains a cycle, so does $P_1 \cup P_2 \cup P_3$ (each path $\tilde{P}_i$ is obtained by short-cutting the corresponding path P_i). Hence $\tilde{P}_1 \cup \tilde{P}_2 \cup \tilde{P}_3$ is also acyclic, and there is a topological ordering of $(R \setminus R') \cup \{s,t\}$ relative to arcs $\tilde{P}_1 \cup \tilde{P}_2 \cup \tilde{P}_3$. We now prove the second part of the claim. In circulation $\mathcal{C}$, each vertex of R has at least 2 units of flow through it; but vertices $R \setminus R'$ are not covered (even to an extent 1) in the circulation $\mathcal{A}$. So each vertex of $R \setminus R'$ is covered to extent at least 2 in circulation $\mathcal{B}$, and hence in $P_1 \cup P_2 \cup P_3$. In other words, each vertex of $R \setminus R'$ appears on at least two of the three $s-t$ paths P_1, P_2, P_3. This also implies that (after the short-cutting) each $R \setminus R'$ vertex appears on at least two of the three $s-t$ paths $\tilde{P}_1, \tilde{P}_2, \tilde{P}_3$. Now observe that for each consecutive pair (w_j, w_{j+1}) $(1 \leq j \leq h-1)$ in the topological order, there is a common path $\tilde{P}_k$ (for some $k = 1,2,3$) that contains both w_j and w_{j+1}. Furthermore, in $\tilde{P}_k$, w_j and w_{j+1} are consecutive in that order (otherwise, the topological order would contain a back arc!). Thus each arc (w_j, w_{j+1}) (for $1 \leq j \leq h-1$) is present in $\tilde{P}_1 \cup \tilde{P}_2 \cup \tilde{P}_3$, and we obtain the claim.

We also need the following claim to bound the cost of insertions in step 2(e)v.

Claim 5 *For any two vertices $u', u'' \in V$, if $\lambda(u', u''; x)$ (resp. $\lambda(u'', u'; x)$) denotes the maximum flow supported by x from u' to u'' (resp. u'' to u'), then $\lambda(u', u''; x) + \lambda(u'', u'; x) \geq \frac{1}{3}$.*

Proof: If either u' or u'' is in $\{s,t\}$, the claim is obvious since for every vertex v, x supports $\frac{2}{3}$ flow from s to v and from v to t. Otherwise $\{s,t,u',u''\}$ are distinct, and define capacities $\hat{x}$ as:

$$\hat{x}(v_1,v_2) = \begin{cases} x(v_1,v_2) & \text{for arcs } (v_1,v_2) \neq (t,s) \\ 1 & \text{for arc } (v_1,v_2) = (t,s) \end{cases}$$

Observe that $\hat{x}$ is Eulerian; now apply Theorem 2 to $\hat{x}$ and split-off all vertices of V except $T = \{s,t,u',u''\}$, and obtain capacities y on the arcs induced on T. We have $\lambda(t_1,t_2;y) = \lambda(t_1,t_2;\hat{x})$ for all $t_1, t_2 \in T$. Note that since neither t nor s is split-off, their degrees in y are unchanged

from $\hat{x}$, and also $y(t,s) \geq \hat{x}(t,s) = 1$. Since the out-degree of t in $\hat{x}$ (hence in y) is 1 and $y_{t,s} \geq 1$, we have $y(t,u') = y(t,u'') = 0$ and $y(t,s) = 1$. The capacities y support at least $\frac{2}{3}$ flow from s to u'; so $y(s,u') + y(u'',u') \geq \frac{2}{3}$. Similarly for u'', we have $y(s,u'') + y(u',u'') \geq \frac{2}{3}$, and adding these two inequalities we get $y(u',u'') + y(u'',u') + (y(s,u') + y(s,u'')) \geq \frac{4}{3}$. Note that $y(s,u') + y(s,u'') \leq y(\delta^+(s)) = \hat{x}(\delta^+(s)) = 1$ (the degree of s is unchanged in the splitting-off). So $y(u',u'') + y(u'',u') \geq \frac{1}{3}$. Since y is obtained from $\hat{x}$ by a sequence of splitting-off operations, it follows that $\hat{x}$ supports flows corresponding to all edges in y *simultaneously*. In particular, the following flows are supported disjointly in $\hat{x}$: $\mathcal{F}_1$ that sends $y(u',u'')$ units from u' to u'', $\mathcal{F}_2$ that sends $y(u'',u')$ units from u'' to u', and $\mathcal{F}_3$ that sends $y(t,s) = 1$ unit from t to s. Hence the flows $\mathcal{F}_1$ and $\mathcal{F}_2$ are each supported by $\hat{x}$ and *do not* use the extra (t,s) arc (since $\hat{x}(\delta^+(t)) = \hat{x}(t,s) = 1$). This implies that the flows $\mathcal{F}_1$ and $\mathcal{F}_2$ are both supported by the original capacities x (where $x(t,s) = 0$). Hence $\lambda(u',u'';x) + \lambda(u'',u';x) \geq y(u',u'') + y(u'',u') \geq \frac{1}{3}$.

From Claim 4, we obtain that the cost addition in step 2e(iv) is at most $d(\tilde{P}_1) + d(\tilde{P}_2) + d(\tilde{P}_3) \leq d(P_1) + d(P_2) + d(P_3) \leq 3(d \cdot x)$ (from Claim 3). We now consider the cost addition to σ in step 2(e)v. Claim 5 implies that for any pair of vertices $u', u'' \in V$, x supports $\frac{1}{6}$ flow either from u' to u'' or from u'' to u'. Also for every vertex u, x supports $\frac{2}{3}$ flow from s to u and from u to t. Since σ always contains an $s-t$ path in step 2(e)v, there is always some position along this $s-t$ path to insert any vertex $u \in R'$ as required in step 2(e)v. Furthermore, the cost increase in any such insertion step is at most $12(d \cdot x)$. Hence the total cost for inserting all the vertices R' into σ is at most $12|R'|(d \cdot x) = O(\sqrt{n})(d \cdot x)$. Thus the total cost of σ at the end of the algorithm is $O(\sqrt{n})(d \cdot x)$. Finally note that σ is connected (in the undirected sense), Eulerian at all vertices in $V \setminus \{s,t\}$ and has outdegree 1 at s. This implies that σ corresponds to a spanning $s-t$ walk. This completes the proof of the following.

Theorem 4 *The integrality gap of $(ATSP - path)$ is at most $O(\sqrt{n})$.*

4 Unweighted Directed Metrics

In the special case where the metric is induced by shortest paths in an unweighted directed graph, we obtain an improved approximation guarantee for the minimum latency problem. This draws on ideas from the undirected latency problem, and the $O(\log^2 n)$ approximation ratio for directed orienteering ([15] and [4]). The directed orienteering problem is as follows: given a starting vertex r in an asymmetric metric and length bound L, find an r-path of length at most L covering the maximum number of vertices. We note that the reduction from ATSP to directed latency also holds in unweighted directed metrics, and the best known approximation ratio for ATSP even on this special class is $O(\log n)$. Here we show the following.

Theorem 5 *An α-approximation algorithm for directed orienteering implies an $O(\alpha + \gamma)$ approximation algorithm for the directed latency problem on unweighted digraphs, where γ is the best approximation ratio for ATSP. In particular there is an $O(\log^2 n)$ approximation.*

Let $G = (V, A)$ denote the underlying digraph that induces the given metric, and r the root vertex. We first argue (Section 4.1) that if G is strongly connected, then there is an $O(\alpha)$-approximation algorithm. Then we show (Section 4.2) how this can be extended to the case when G is not strongly connected.

4.1 G is strongly connected

In this case, the distance from any vertex to the root r is at most $n = |V|$. The algorithm and analysis for this case are identical to those for the undirected latency problem [2, 10, 3]. Details are given in Appendix B.

Remark: This 'greedy' approach does not work in the general directed case since it is unclear how to bound the length of back-arcs to the root r (which is required to stitch the paths that are computed greedily). In the undirected case, back-arcs can be easily bounded by the forward length, and this approach results in a constant approximation algorithm. In the unweighted strongly-connected case (considered above), the total length of back-arcs used by the algorithm could be bounded by roughly n^2 (which is also a lower bound for the latency problem). By an identical analysis, it also follows that there is an $O(\alpha)$-approximation for the directed latency problem on metrics (V, d) with the following property: for every vertex $v \in V$, the back-arc length to r is within a constant factor of the forward-arc length from r, i.e. $d(v, r) \leq O(1) \cdot d(r, v)$. As a consequence, we obtain an $O(\alpha) = O(\log^2 n)$ approximation for *no-wait flowshop scheduling* with the *weighted completion time* objective (n is the number of jobs in the given instance); this seems to be the first approximation ratio for the problem. The no-wait flowshop problem can be modeled as a minimum latency problem in an appropriate directed metric [20, 18], with the property that all back-arcs to the root r have length 0; hence the above greedy approach applies.

4.2 G is not strongly connected

In this case, we show an $O(\gamma+\beta)$-approximation algorithm, where γ is the approximation guarantee for ATSP and β is the approximation guarantee for the minimum latency problem on unweighted strongly-connected digraphs. From Section 4.1, $\beta = O(\alpha)$, where α is the approximation ratio for directed orienteering. Consider the strong components of G, which form a directed acyclic graph. If the instance is feasible, there is a Hamilton path in G from r; so we can order the strong components of G as $C_1, \cdots, C_l$ such that $r \in C_1$ and any spanning path from r visits the strong components in that order. For each $1 \leq i \leq l$, let $n_i = |C_i|$, and pick an arbitrary vertex $s_i \in C_i$ as root for each strong component (setting $s_1 = r$).

Lemma 3 *There exists a spanning r-path τ^* having latency objective at most $7 \cdot \mathsf{Opt}$ such that $\tau^* = \tau_1 \cdot (s_1, s_2) \cdot \tau_2 \cdot (s_2, s_3) \cdots (s_{l-1}, s_l) \cdot \tau_l$, where each τ_i (for $1 \leq i \leq l$) is an s_i-tour covering all vertices in C_i.*

Proof: Consider the optimal latency r-path P^*: this is a concatenation $P_1 \cdot P_2 \cdots P_l$ of paths in each strong component (P_i is a spanning path on C_i). For each $1 \leq i \leq l$, let $\mathsf{Lat}(P_i)$ denote the latency of vertices C_i just along path P_i, and $D_i = \sum_{j=1}^{i-1} d(P_j)$ be the distance traversed by P^* before P_i. Then the total latency along P^* is $\mathsf{Opt} = \sum_{i=1}^{l} (n_i \cdot D_i + \mathsf{Lat}(P_i))$.

For each $1 \leq i \leq l$, let τ_i denote a spanning tour on C_i, obtained by adding to P_i the direct arcs: from s_i to its first vertex and from its last vertex to s_i. Each of these extra arcs in τ_i has length at most $n_i - 1$ (since C_i is strongly connected), and $d(P_i) \geq n_i - 1$ (it is spanning on C_i); so $d(\tau_i) \leq 3d(P_i)$. Let $\mathsf{Lat}(\tau_i)$ denote the latency of vertices C_i along τ_i; from the above observation we have $\mathsf{Lat}(\tau_i) \leq n_i \cdot (n_i - 1) + \mathsf{Lat}(P_i)$. Now we obtain τ^* as the concatenation $\tau_1 \cdot (s_1, s_2) \cdot \tau_2 \cdots (s_{t-1}, s_l) \cdot \tau_l$. Note also that for any $1 \leq i \leq l-1$, $d(s_i, s_{i+1}) \leq n_i + n_{i+1}$. So the

latency in τ^* of vertices C_i is:

$$\begin{aligned}
& n_i \cdot \textstyle\sum_{j=1}^{i-1}(d(\tau_j) + d(s_j, s_{j+1})) + \mathsf{Lat}(\tau_i) \\
\leq\ & n_i \cdot \textstyle\sum_{j=1}^{i-1}(3d(P_j) + n_j + n_{j+1}) + n_i \cdot (n_i - 1) + \mathsf{Lat}(P_i) \\
\leq\ & n_i \cdot \textstyle\sum_{j=1}^{i-1}(3d(P_j) + 2n_j) + n_i^2 + n_i \cdot (n_i - 1) + \mathsf{Lat}(P_i) \\
\leq\ & n_i \cdot \textstyle\sum_{j=1}^{i-1} 7d(P_j) + n_i^2 + n_i \cdot (n_i - 1) + \mathsf{Lat}(P_i) \\
\leq\ & 7n_i \cdot D_i + 2n_i^2 + \mathsf{Lat}(P_i) \\
\leq\ & 7n_i \cdot D_i + 5 \cdot \mathsf{Lat}(P_i)
\end{aligned}$$

The last inequality follows from the fact that $\mathsf{Lat}(P_i) \geq n_i^2/2$ (P_i is a path on n_i vertices in an unweighted metric). So the total latency of τ^* is at most $7\sum_{i=1}^{l}(n_i \cdot D_i + \mathsf{Lat}(P_i)) = 7 \cdot \mathsf{Opt}$.

The algorithm for directed latency in this case computes an approximately minimum latency s_i-path for each C_i separately (using the algorithm in Section 4.1); by adding the direct arc from the last vertex back to s_i, we obtain C_i-spanning tours $\{\sigma_i\}_{i=1}^{l}$. We now use the following claim from [2] to bound the length of each tour σ_i.

Claim 6 ([2]) *Given C_i-spanning tours σ_i and π_i, there exists a poly-time computable tour σ_i' on C_i of length at most $3 \cdot d(\pi_i)$ and latency at most thrice that of σ_i.*

Proof: Tour σ_i' is constructed as follows: starting at s_i, traverse tour σ_i until a length of $d(\pi_i)$, then traverse tour π_i from the current vertex to visit all remaining vertices and then return to s_i. Note that tour π_i will have to be traversed at most twice, and so the length of σ_i' is at most $3d(\pi_i)$. Furthermore, the total latency along σ_i' for vertices visited in the σ_i part is at most $\mathsf{Lat}(\sigma_i)$ (the latency along σ_i). Also the latency along σ_i' of each vertex v visited in the π_i part is at most $3d(\pi_i)$, which is at most thrice its latency in σ_i. Hence the total latency along σ_i' is at most $3 \cdot \mathsf{Lat}(\sigma_i)$.

This implies that by truncating σ_i with a γ-approximate TSP on C_i, we obtain another spanning tour σ_i' of length $3\gamma \cdot L_i$ and latency $3 \cdot \mathsf{Lat}(\sigma_i)$ (where L_i is length of the minimum TSP on C_i). The final r-path is the concatenation of these local tours, $\pi = \sigma_1' \cdot (s_1, s_2) \cdot \sigma_2' \cdots (s_{l-1}, s_l) \cdot \sigma_l'$.

Claim 7 *The latency of r-path π is at most $O(\gamma + \beta) \cdot \mathsf{Opt}$.*

Proof: Consider the near-optimal r-path τ^* given by Lemma 3. For $1 \leq i \leq l$, let Opt_i denote the latency of the C_i-spanning tour τ_i, and $\tilde{D}_i = \sum_{j=1}^{i-1}(d(\tau_j) + d(s_j, s_{j+1}))$ denote the length of τ^* before C_i. Then the total latency of τ^* can be written as $\sum_{i=1}^{l}(n_i \cdot \tilde{D}_i + \mathsf{Opt}_i) \leq 7 \cdot \mathsf{Opt}$.

Now consider the r-path π output by the algorithm. The s_i-tour τ_i is a feasible solution to the minimum latency instance on C_i; so the latency of tour σ_i is at most $\beta \cdot \mathsf{Opt}_i$, since we use a β-approximation for each such instance. So for each $1 \leq i \leq l$, the truncated tour σ_i' has latency $\mathsf{Lat}(\sigma_i') \leq 3\beta \cdot \mathsf{Opt}_i$, and length $d(\sigma_i') \leq 3\gamma L_i$. Again, the latency of π can be written as $\sum_{i=1}^{l}(n_i \cdot D_i' + \mathsf{Lat}(\sigma_i'))$, where $D_i' = \sum_{j=1}^{i-1}(d(\sigma_j') + d(s_j, s_{j+1}))$ is the length of π before C_i. So the latency of vertices C_i in π is:

$$\begin{aligned}
& n_i \cdot \textstyle\sum_{j=1}^{i-1}(d(\sigma_j') + d(s_j, s_{j+1})) + \mathsf{Lat}(\sigma_i') \\
\leq\ & n_i \cdot \textstyle\sum_{j=1}^{i-1}(3\gamma \cdot L_j + d(s_j, s_{j+1})) + 3\beta \cdot \mathsf{Opt}_i \\
\leq\ & n_i \cdot \textstyle\sum_{j=1}^{i-1}(3\gamma \cdot d(\tau_j) + d(s_j, s_{j+1})) + 3\beta\mathsf{Opt}_i \\
\leq\ & 3\gamma n_i \cdot \textstyle\sum_{j=1}^{i-1}(d(\tau_j) + d(s_j, s_{j+1})) + 3\beta\mathsf{Opt}_i \\
=\ & 3\gamma n_i \cdot \tilde{D}_i + 3\beta\mathsf{Opt}_i \\
\leq\ & 3(\gamma + \beta)(n_i \cdot \tilde{D}_i + \mathsf{Opt}_i)
\end{aligned}$$

So the total latency of π is at most $3(\gamma+\beta)\sum_{i=1}^{l}(n_i \cdot \tilde{D}_i + \mathsf{Opt}_i) \leq O(\gamma+\beta)\cdot\mathsf{Opt}$. Theorem 5 now follows.

References

[1] A. Blum, S. Chawla, D. R. Karger, T. Lane, A. Meyerson, and M. Minkoff. Approximation Algorithms for Orienteering and Discounted-Reward TSP. In *FOCS*, pages 46–55, 2003.

[2] Avrim Blum, Prasad Chalasani, Don Coppersmith, William R. Pulleyblank, Prabhakar Raghavan, and Madhu Sudan. The minimum latency problem. In *STOC*, pages 163–171, 1994.

[3] Kamalika Chaudhuri, Brighten Godfrey, Satish Rao, and Kunal Talwar. Paths, Trees, and Minimum Latency Tours. In *FOCS*, pages 36–45, 2003.

[4] Chandra Chekuri, Nitish Korula, and Martin Pal. Improved Algorithms for Orienteering and Related Problems. In *SODA*, pages 661–670, 2008.

[5] Chandra Chekuri and Martin Pal. An $O(\log\ n)$ Approximation Ratio for the Asymmetric Traveling Salesman Path Problem. *Theory of Computing*, 3:197–209, 2007.

[6] C. H. Papadimitriou G. Papageorgiou N. Papakostantinou F. Afrati, S. Cosmadakis. The complexity of the traveling repairman problem. *Informatique Theorique et Applications*, 20:79–87, 1986.

[7] Uriel Feige and Mohit Singh. Improved Approximation Ratios for Traveling Salesperson Tours and Paths in Directed Graphs. In *APPROX-RANDOM*, pages 104–118, 2007.

[8] A. Frank. On Connectivity properties of Eulerian digraphs. *Annals of Discrete Mathematics*, 41:179–194, 1989.

[9] A. Frieze, G. Galbiati, and F. Maffioli. On the worst-case performance of some algorithms for the asymmetric travelling salesman problem. *Networks*, 12:23–39, 1982.

[10] Michel Goemans and Jon Kleinberg. An improved approximation ratio for the minimum latency problem. *Mathematical Programming*, 82:111–124, 1998.

[11] B. Jackson. Some remarks on arc-connectivity, vertex splitting, and orientation in digraphs. *Journal of Graph Theory*, 12(3):429–436, 1988.

[12] Jon Kleinberg and David Williamson. Unpublished note. 1998.

[13] Fumei Lam and Alantha Newman. Traveling salesman path problems. *Mathematical Programming*, online, 2006.

[14] Edward Minieka. The delivery man problem on a tree network. *Annals of Operations Research*, 18:261–266, 1989.

[15] Viswanath Nagarajan and R. Ravi. Poly-logarithmic approximation algorithms for directed vehicle routing problems. In *APPROX-RANDOM*, pages 257–270, 2007.

[16] George L. Nemhauser and Laurence A. Wolsey. Integer and Combinatorial Optimization. 1999.

[17] S. Sahni and T. Gonzalez. P-complete approximation problems. *Journal of the ACM*, 23:555–565, 1976.

[18] Maxim Sviridenko. Makespan Minimization in No-Wait Flow Shops: A Polynomial Time Approximation Scheme. *SIAM J. Discret. Math.*, 16(2):313–322, 2003.

[19] David Williamson. Analysis of the held-karp heuristic for the traveling salesman problem. *Master's thesis, MIT Computer Science*, 1990.

[20] D.A. Wismer. Solution of the flowshop sheduling problem with no intermediate queues. *Operations Research*, 20:689–697, 1972.

A Relation between ATSP and directed latency

Proposition 6 *An α-approximation for directed latency (path version) implies a 4α-approximation for ATSP. This reduction also holds in the special case of metrics induced by unweighted digraphs.*

Proof: We only present the reduction for unweighted metrics (the general case is identical). Let $\mathcal{A}$ be an α-approximation algorithm for the (path version) directed latency on unweighted metrics, and $G = (V, E)$ be any n-vertex directed graph. We show how $\mathcal{A}$ can be used to obtain a 4α approximation for ATSP on G. Pick some root $r \in V$ and modify G by adding to it a directed path of n new vertices U originating at r: let H be the graph so obtained. A candidate solution for latency on H first visits the vertices V along the optimal ATSP tour on G and then visits U along the new path: this implies that the optimal latency on H, $\mathsf{Lat}^*(H) \leq 2n(\mathsf{Tsp}^*(G) + n)$ where $\mathsf{Tsp}^*(G)$ is the optimal value of ATSP on G. Note that $\mathsf{Tsp}^* \geq n$ (G has n vertices), and so $\mathsf{Lat}^*(H) \leq 4n\mathsf{Tsp}^*(G)$. Note also that any spanning r-path in H must first visit all the vertices V in a tour (on graph G) and then the vertices U along the new path. Since there are n vertices (those in U) that appear after the spanning tour on G, we have $\mathsf{Lat}^*(H) \geq n \cdot \mathsf{Tsp}^*(G)$. Thus we have $\frac{1}{4n}\mathsf{Lat}^*(H) \leq \mathsf{Tsp}^*(G) \leq \frac{1}{n}\mathsf{Lat}^*(H)$, which implies the proposition.

We now prove Theorem 1, which shows the equivalence (up to a constant factor) of approximating the path and tour versions of directed latency.

Proof of Theorem 1. Recall the definitions of the two versions of directed latency on asymmetric metric (V, d) with root vertex $r \in V$. *Path latency*: find a spanning *path* π originating at r that minimizes $\sum_{v \in V} d^\pi(r, v)$ (here $d^\pi(r, r) = 0$). *Tour latency*: find a spanning *tour* τ originating at r that minimizes $\sum_{v \in V} d^\tau(r, v)$ (here $d^\tau(r, r) = d(\tau)$).

We first show that any approximation algorithm for path-latency implies the same guarantee for tour-latency. Modify the given metric (V, d) (for tour-latency) by adding a new vertex r' with distances to vertices V as follows: $d(v, r') = d(v, r)$ and $d(r', v) = \infty$ for all $v \in V$. It is clear that optimal value of tour-latency on (V, d) equals the optimal value of path-latency on the modified metric, which implies the first direction.

For the reverse direction, let $\mathcal{A}$ be any α-approximation algorithm for tour-latency (we assume $\alpha \leq n$). Set $\beta = \lceil 4\alpha \rceil + 1$ and define metric $(\tilde{V}, \tilde{d})$ which is obtained from (V, d) by making β copies of each vertex in $V \setminus r$. Let $\tilde{\mathsf{pLat}}$ (resp. pLat) denote the optimal value of path-latency on $\tilde{d}$ (resp. d); note that $\tilde{\mathsf{pLat}} = \beta \cdot \mathsf{pLat}$. We assume (by scaling) that the smallest non-zero distance in d is 1, and that there is a finite length spanning r-path in metric d (otherwise the latency problem

is trivial); so $\mathsf{pLat} \leq n^2 d_{max}$ where d_{max} is the maximum finite distance in d. We further modify metric $\tilde{d}$ by adding dummy arcs of length L (value to be set later) from each $v \in V \setminus r$ to r, and let l denote the shortest path metric on this modified graph. Note that the optimal value of tour-latency on metric l is $\mathsf{tLat} \leq 2 \cdot \tilde{\mathsf{pLat}} + L$. The algorithm for path-latency on d is as follows.

- For each $0 \leq i \leq \lg(\beta n^2 d_{max})$ do:
  1. Set $L \leftarrow 2^i$ and l to be the metric as defined above.
  2. Run algorithm $\mathcal{A}$ for tour-latency on l to get tour τ. Let π denote the portion of τ until the first usage of a dummy arc.
  3. If π visits at least one copy of each vertex in V, consider this as a feasible solution to path-latency on d; otherwise skip this iteration.
- Output the best path-latency solution encountered in Step 3 above.

Note that since $1 \leq \tilde{\mathsf{pLat}} \leq \beta n^2 d_{max}$, there is an iteration where $\tilde{\mathsf{pLat}} \leq L = 2^i \leq 2 \cdot \tilde{\mathsf{pLat}}$. We now argue that in this iteration i, the above algorithm gets a feasible solution in Step 3 with path-latency value (on metric d) at most $4\alpha \cdot \mathsf{pLat}$. The tour τ found by algorithm $\mathcal{A}$ has latency (in metric l) $\mathsf{Lat}(\tau) \leq \alpha \cdot \mathsf{tLat} \leq \alpha(2 \cdot \tilde{\mathsf{pLat}} + L) \leq 4\alpha \cdot \tilde{\mathsf{pLat}}$. If x denotes the number of unvisited vertices of $\tilde{V}$ in π (i.e. when a dummy arc is used for the first time in τ) then $\mathsf{Lat}(\tau) \geq L \cdot x$; hence $x \leq \mathsf{Lat}(\tau)/L \leq 4\alpha \cdot \tilde{\mathsf{pLat}}/L \leq 4\alpha < \beta$. Since $\tilde{V}$ has β copies of each vertex of V, π visits at least one copy of each V-vertex: so Step 3 records π as a potential solution. We modify τ so that it visits all copies of each V-vertex the first time some copy is visited: this only reduces the latency of τ. After this modification, note that dummy arcs are used in τ only after all vertices $\tilde{V}$ are visited: in other words, τ (and also π) induces a path-latency solution on metric $(\tilde{V}, \tilde{d})$ of latency value at most $\mathsf{Lat}(\tau) \leq 4\alpha \cdot \tilde{\mathsf{pLat}}$. Hence the corresponding solution in metric (V, d) has latency at most $\frac{1}{\beta} 4\alpha \cdot \tilde{\mathsf{pLat}} = 4\alpha \cdot \mathsf{pLat}$. Thus the best solution to path-latency on d obtained by the above algorithm has value at most $4\alpha \cdot \mathsf{pLat}$, and it is a 4α approximation algorithm.

B Directed latency on unweighted strongly connected graphs

The algorithm for the unweighted strongly connected case is as follows.

1. Set current length bound $L = n$, and current tour $\tau = \phi$.
2. While not all vertices covered, do:
   (a) Run the directed orienteering algorithm α times, each time obtaining a path of length $\leq L$ from r that covers (approximately) the maximum number of remaining vertices.
   (b) Stitch the α paths into an r-tour by adding arcs from the end of each path back to r, and append this tour to the end of τ.
   (c) Set length bound $L \leftarrow 2 \cdot L$.
3. Output tour τ.

In the optimal latency r-path, let N_i (for $i = 0, 1, \cdots$) be the number of vertices having latency at most $2^i n$. Then one can verify that the optimal latency $\mathsf{Opt} \geq \frac{1}{2} \sum_{i \geq 0} 2^{i-1} n \cdot (n - N_i)$. In the

r-path returned by the algorithm, let R_i (for $i \geq 0$) denote the number of vertices left uncovered at the end of iteration i (where length bound is $2^i n$). Note that each back arc to r has length at most n; so the increase in the length of the r-path in iteration i is at most $\alpha \cdot (2^i n + n) \leq 2^{i+1}\alpha n$.

Claim 8 *For all $i \geq 1$, either $R_i \leq n - N_i$ or $R_i \leq \frac{1}{e}R_{i-1} + (1 - \frac{1}{e})(n - N_i)$. Hence for all $i \geq 1$, $R_i \leq \max\{n - N_i, \frac{1}{e}R_{i-1} + (1 - \frac{1}{e})(n - N_i)\} \leq \frac{1}{e}R_{i-1} + (2 - \frac{1}{e})(n - N_i)$.*

Proof: If $R_{i-1} \leq n - N_i$, then $R_i \leq R_{i-1} \leq n - N_i$. Suppose $R_{i-1} > n - N_i$; then there is an r-path (of length $2^i n$) in iteration i containing at least $N_i - (n - R_{i-1})$ new vertices (the optimal path restricted to uncovered vertices & this length bound). Since we use an α-approximation for directed orienteering in step 2a (repeatedly α times), the resulting set of r-paths cover at least $(1 - 1/e)(R_{i-1} - n + N_i)$ new vertices- this follows from a standard maximum-coverage argument. In other words, $R_i \leq R_{i-1} - (1 - 1/e)(R_{i-1} - n + N_i) = \frac{1}{e}R_{i-1} + (1 - \frac{1}{e})(n - N_i)$.■

If m_0 is the number of vertices covered in the first iteration (when length bound is n), we can upper bound the latency Alg of the r-tour τ by $m_0 \cdot 2\alpha n + \sum_{i \geq 0} 2^{i+2}\alpha n \cdot R_i$; recall that the increase in length of τ in iteration i is at most $2^{i+1}\alpha n$. Since $m_0 \leq n$ and $\mathsf{Opt} \geq \binom{n}{2}$ in any n-vertex unweighted metric, it suffices to bound $T = \sum_{i \geq 0} 2^{i+2}\alpha n \cdot R_i$. From Claim 8, we have:

$$\begin{aligned}
T &\leq 4\alpha n R_0 + \alpha n \textstyle\sum_{i \geq 1} 2^{i+2}[\frac{1}{e}R_{i-1} + (2 - \frac{1}{e})(n - N_i)] \\
&= 4\alpha n R_0 + \tfrac{2}{e}\alpha n \textstyle\sum_{i \geq 0} 2^{i+2} R_i + 4(2 - \frac{1}{e})\alpha n \sum_{i \geq 1} 2^i (n - N_i) \\
&= 4\alpha n R_0 + \tfrac{2}{e}T + 8(2 - \tfrac{1}{e})\alpha n \textstyle\sum_{i \geq 1} 2^{i-1}(n - N_i) \\
&\leq 4\alpha n^2 + \tfrac{2}{e}T + 8(2 - \tfrac{1}{e})\alpha \mathsf{Opt} \\
&\leq \tfrac{2}{e}T + 8(3 - \tfrac{1}{e})\alpha \mathsf{Opt}
\end{aligned}$$

So $T \leq O(1)\alpha \cdot \mathsf{Opt}$, and $\mathsf{Alg} \leq O(\alpha) \cdot \mathsf{Opt}$.