A Contamination Aware Droplet Routing Algorithm for Digital Microfluidic Biochips

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ABSTRACT
In this paper, we propose a contamination aware droplet routing algorithm for digital microfluidic biochips (DMFBs). To reduce the routing complexities and the used cells, we first construct preferred routing tracks by analyzing the global moving vector of droplets to guide the droplet routing. To cope with contaminations within one subproblem, we first apply a k-shortest path routing technique to minimize the contaminated spots. Then, to take advantage of multiple wash droplets, we adopt a minimum cost circulation algorithm (MCC) for optimal wash-droplet routing to simultaneously minimize used cells and the cleaning time. Furthermore, a look-ahead prediction technique is used to determine the contaminations between successive subproblems. After that, we can simultaneously clean both contaminations within one subproblem and those between successive subproblems by using the MCC-based algorithm to reduce the execution time and the used cells. Based on four widely used bioassays, our algorithm reduces the used cells and the execution time significantly compared with the state-of-the-art algorithm.

1. INTRODUCTION
Digital microfluidic biochip (DMFB) is an emerging technology that aims to miniaturize and integrate droplet-handling on a chip. Recently, many on-chip laboratory procedures such as immunoassay, real-time DNA sequencing, and protein crystallization have all been successfully demonstrated on DMFBs. The dynamic reconfigurability inherent in DMFBs allows different droplet routes to share cells (electrodes) on the microfluidic array during different time intervals. However, contaminations caused by bead retention and liquid residue between successive droplet routes of different biomolecules may cause inevitable erroneous reaction. Moreover, these errors will possibly breakdown the electrodes and cause electrode short problems, which result in physical defects and produce incorrect behaviors in the electrical domain. Intuitively, contaminations can be avoided by routing in disjoint manner. This method avoids the overlap between different droplet routes thereby minimizing the likelihood of the contamination problem. However, as the increased design complexity enabled more and more biological operations to a DMFB, finding disjoint routes has become more and more difficult. Furthermore, disjoint routes also restrict the spare cells for replacing faulty primary cells to ensure the correctness of bioassay execution. Hence, the fault tolerance of bioassay is significantly reduced.

Although silicone oil with its low surface tension and spreading property has been advocated as a filler medium to prevent contaminations, it has been proved that it is not sufficient enough for many types of proteins and heterogeneous immunoassays. To cope with this problem, a wash droplet is introduced to clean the contaminated spots on the surface of the microfluidic array. Given an initial bioassay with two droplets and peripheral devices as shown in Figure 1 (a), if we adopt the disjoint routes to avoid the contamination problem as shown in Figure 1 (b), the execution time and the number of used cells for nets are 18 and 26, respectively. However, in Figure 1 (c), a contaminated spot (cross-section) occurs between two different routes by simply adopting shortest path routing. To clean this contaminated spot, a wash droplet is dispensed from the wash reservoir and transported via this contaminated spot. As shown in Figure 1 (d), to ensure the correctness of wash operation, the wash droplet must clean the contaminated spot in the time interval $[t_{2_s}, t_{2_e}]$, where $t_{2_s}$ and $t_{2_e}$ denote the arrival time at the contaminated spot of $d_1$ and $d_2$, respectively. If the wash droplet cannot arrive the contaminated spot before $t_{2_s}$, $t_{2_e}$ must be postponed until this contaminated spot has been cleaned. By this wash operation, the execution time and the used cells for nets are reduced to 12 and 19, thereby achieving a better solution quality. Thus, if the wash operation cannot be simultaneously considered with droplet routing, the droplet transportation time will increase significantly, thereby causing the time-to-result effects and reducing the reliability of bioassay.

Figure 1: Illustration of the contamination aware droplet routing. (a) Initial bioassay. (b) Disjoint routing for contamination avoidance. (c) Shortest path routing with a contaminated spot. (d) Wash-droplet routing.

Furthermore, contaminated spots occur not only within one subproblem (intra-contaminations) but also between successive subproblems (inter-contaminations). Contaminations in the previous subproblem are treated as blockages for the next subproblem. Additional wash droplets are needed to clean the inter-contaminations and may cause timing overhead for bioassays.

In this paper, we propose a contamination aware droplet routing algorithm on DMFBs. To fully utilize wash droplets, we simultaneously clean both intra- and inter-contaminations within one subproblem that can reduce the execution time significantly. Furthermore, we can effectively minimize the used cells to achieve better reliability and fault tolerance for bioassays.

1.1 Background and Related Prior Work
Droplet routing is a critical step in DMFB physical design automation. Unlike traditional VLSI routing, in addition to routing path selection, the droplet routing problem needs to address the issue of scheduling droplets under the practical constraints imposed by the fluidic property and the timing restriction of the synthesis result.
Droplet routing problem has attracted much attention in the literature recently [3, 4, 7, 9, 11]. Current droplet-routing algorithms have a detrimental effect on the unrestricted sharing of used cells by various droplet routes. In [12], a novel droplet routing algorithm is proposed for cross-contamination avoidance. It attempts to determine disjoint droplet routes by applying modified Lee algorithm. If it is infeasible to find disjoint routes, the wash droplets are then scheduled between successive droplet visits to clean up the contaminations. However, disjoint routes restrict the spare cells for replacing faulty primary cells to ensure the correctness of bioassay execution. Hence, the fault tolerance of bioassay is significantly reduced. Since contaminated spots may also occur between successive subproblems (inter-contaminations), a wash operation is also scheduled between successive subproblems to avoid the contamination problem. However, the extra execution time of wash operation interrupts the continuous biological operations between successive subproblems, which suffers from the time-to-result effects and sample degradation problem. Figure 2(a) illustrates the method proposed in [12].

1.2 Our Contribution

In this paper, we propose a contamination-aware droplet routing algorithm for DMFBs. Unlike the aforementioned routing algorithm of wash-droplet, which cleans intra- and inter-contaminations separately, our algorithm simultaneously clean them within one subproblem to reduce the execution time significantly (see Figure 2 for an illustration). To fully utilize the wash droplets, our algorithm consists of three major stages: preprocessing, intra-contamination aware routing, and inter-contamination aware routing between successive subproblems. Different from the aforementioned work, our algorithm has the following distinguished features:

- A global moving vector analysis for constructing preferred routing tracks to minimize the number of used unit cells.
- A $k$-shortest path routing technique to minimize the contaminated spots within one subproblem.
- A routing compaction technique by dynamic programming to determine the contaminated spots within one subproblem and minimize the execution time of bioassays.
- A look-ahead prediction technique to determine the contaminated spots between successive subproblems and minimize the total routing time of wash droplets.
- A minimum cost circulation technique is adopted to simultaneously clean intra- and inter-contaminations to minimize the used cells and the execution time.

Based on four widely used bioassays, our algorithm reduces the used cells and the execution time significantly compared with the state-of-the-art algorithm.

The rest of this paper is organized as follows. Section 2 formulates the contamination-aware droplet routing problem, while Section 3 describes the overview of our algorithm. Our algorithm consists of three major stages: preprocessing, intra-contamination aware routing, and inter-contamination aware routing which are given in Section 4, 5, and 6, respectively. Finally, the experimental results and the concluding remarks are provided in Section 7 and 8.

2. PROBLEM FORMULATION

There are three routing constraints in contamination-aware droplet routing: the fluidic constraint, the timing constraint, and the contamination constraint. The fluidic constraint is to avoid the unexpected mixtures between two droplets of different nets during their transportation and it can further be divided into the static and dynamic fluidic constraints [9]. Let $d_i$ at $(x_i, y_i)$ and $d_j$ at $(x'_j, y'_j)$ denote two independent droplets at time $t$. Then, the following constraints should be satisfied for any $t$ during routing:

- **Static constraint:** $|x_i - x'_j| > 1$ or $|y_i - y'_j| > 1$.
- **Dynamic constraint:** $|x_i + x'_j| > 1$ or $|y_i + y'_j| > 1$ or $|x_i - x'_j| + 1 > 1$ or $|y_i - y'_j| + 1 > 1$.

The static fluidic constraint states that the minimum spacing between two droplets is one cell for any $t$ during routing. The dynamic fluidic constraint states that the activated cell for $d_i$ cannot be adjacent to $d_j$. The reason is there can be more than one activated neighboring cell for $d_i$. Therefore, we may have an unexpected mixing between $d_i$ and $d_j$.

Besides the fluidic constraint, there exists the timing constraint. The timing constraint specifies the maximum arrival time of a droplet from source to target. For fast bioassay execution or better reliability, it is desirable to minimize the execution time among all droplets. Furthermore, it is desirable to minimize the number of unit cells that are used during routing. Since a unit cell of a DMFB can be defective due to manufacturing or environmental issues, using a smaller number of unit cells for routing can be beneficial for robustness.

The contamination constraint is to prevent the contamination problem from the erroneous bioassay outcome. This is because when two droplets successively pass through a same cell, the liquid residue left behind by the first droplet can contaminate the second droplet and may mislead the bioassay outcome. To characterize the contamination constraint, let the cell $c_{ij}$ denotes the contaminated spot caused by the two routes of $d_i$ and $d_j$, and $c_{ij}$ represents the corresponding arrival clock cycle at $c_{ij}$ for $d_i (d_j)$. To avoid the contamination problem, a wash droplet must clean (pass through) this contaminated spot in the time interval $[t^*, t^*)$ to perform the wash operation.

The contamination-aware droplet routing problem on a 2D plane can be formulated as follows:

**Input:** A netlist of droplets, a set of wash droplets, the locations of blockages, the locations of reservoirs, and the timing constraint.

**Objective:** Route all droplets from their source cells to their target cells while minimizing the contaminated spots, the execution time, and the used cells for better fault tolerance. Route wash droplets to clean contaminated spots between different droplet routes while minimizing the cleaning time.

**Constraint:** Both fluidic and timing constraints are satisfied.

3. ALGORITHM OVERVIEW

Our contamination-aware droplet routing algorithm which consists of three major stages: preprocessing, intra-contamination aware
TABLE I: NOTATIONS USED IN OUR ALGORITHM.

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>$U_t$</td>
<td>set of droplets</td>
</tr>
<tr>
<td>$W_t$</td>
<td>set of wash droplets</td>
</tr>
<tr>
<td>$Blk$</td>
<td>cell set of blockages</td>
</tr>
<tr>
<td>$SP_p$</td>
<td>subproblem $p$</td>
</tr>
<tr>
<td>$Bx$</td>
<td>bounding box from the source of $d$ to its target location</td>
</tr>
<tr>
<td>$CS$</td>
<td>cell set of contaminated spots</td>
</tr>
<tr>
<td>$Q_d$</td>
<td>priority queue for determination of routing order</td>
</tr>
<tr>
<td>$Q_c$</td>
<td>priority queue for number of contaminated spots of each routing path</td>
</tr>
<tr>
<td>$CS_{(p,q)}$</td>
<td>cell set of non-washed contaminated spots between subproblems $SP_p$ and $SP_q$. Note that $p = q$ represents the contaminated spots of one subproblem. When all droplets are routed sequentially to their targets, a dynamic-programming-based routing compaction technique is presented to minimize the completion time of bioassay. Since the contamination problem may occur during these routing stages, to prevent contamination problem from interfering the assay outcomes, wash droplets must be appropriately scheduled and routed with reasonable routing time and used cells. In inter-contamination aware routing stage, to avoid the timing overhead which is caused by scheduling wash operation between successive subproblems, we propose a look-ahead routing scheme that simultaneously considers the contaminated spots within one subproblem and successive subproblems. To handle these contaminated spots, a minimum cost circulation (MCC) algorithm is adopted to simultaneously clean these contaminated spots while minimizing the used cells and the execution time. When all the wash droplets are routed, a final solution compaction technique is applied to minimizing the completion time of assay.</td>
</tr>
<tr>
<td>$CS_t$</td>
<td>set of non-washed look-ahead contaminated spots in the bounding box of nodes $v_i$ and $v_j$.</td>
</tr>
</tbody>
</table>

4. PREPROCESSING STAGE

4.1 Preferred Routing Tracks Construction

The goal of droplet routing in a DMFBB is to find an efficient schedule for each droplet from its source to target while both fluidic and timing constraints are satisfied. Furthermore, it is desirable to minimize the execution time among all droplets and the number of unit cells used during routing for better reliability and fault tolerance. However, with the increased design complexities, there may exist substantial macros (ex. mixer) that cause fatal routability problems. Moreover, as multiple droplets are routed in a time-multiplex manner, violations of fluidic constraints occur frequently, involving deadlocks or detour overhead that increase the used unit cells and completion time of DMFBBs. To remedy these deficiencies, we first construct the preferred routing tracks by analyzing each droplet’s preferred moving direction and make droplets route on these specific tracks in order. Consider a droplet $d_i$ located at $(x_i, y_i)$ and its target located at $(x'_i, y'_i)$ in a two-dimensional microfluidic array (if Cartesian coordinate system is used); we define the preferred moving vector $e^d_i$ to be the connection from the point $(x_i, y_i)$ to the point $(x'_i, y'_i)$. Since droplets can only move horizontally or vertically in a microfluidic array, to obtain precise routing tracks information, we decompose $e^d_i$ into two linearly dependent vectors $e^x_i$ and $e^y_i$ along the $x$ and $y$ axis. Then, we define the preferred routing tracks on rows and columns as follows:

- **Track on row $i$:** $k_{x} = \sum_{Bx_{x_i}} e^x_i$ for all $Bx_{x_i}$
- **Track on column $j$:** $k_{y} = \sum_{Bx_{y_j}} e^y_j$ for all $Bx_{y_j}$

By analyzing the preferred moving vector of each droplet in the bounding box, we can construct the preferred routing tracks. Since the fluidic constraints require a minimum spacing between two droplets, the preferred routing tracks are constructed on non-adjacent rows and non-adjacent columns.

The intuition behind our preferred routing tracks construction is similar to traffic control, as each droplet can be regarded as a car. If most of the cars have common driving direction on the global routing track, we will assign this routing track to the preferred driving direction, which is beneficial to the traffic control (avoid the conflicts of fluidic constraints). If these cars did not drive on the preferred routing track, they can still drive the alternative shoulders adjacent to droplet $d_i$ even on the opposite direction of it, but they will be charged for additional costs (minimize the used unit cells).

We model the routing resource as a routing graph $G = (V, E)$. A node in the routing graph represents a cell in the microfluidic array, whereas an edge denotes the connection between two adjacent cells. We define the routing result as a directed graph $G = (V, E)$, where $V$ is the set of all controllable cells, $E$ is the set of directed edges, $c_\text{eq}$ is the edge cost, and $c_\text{illegal}$ is the edge penalty. Note that $c_\text{eq}$ is the least congested path, and $c_\text{illegal}$ is the most congested path.

4.2 Routing Priority Calculation

A key issue in the droplet routing problem is the determination of the droplet routing order. If droplets route in disorder, it will cause fatal routability problem, which increases the routing complexity. To solve this problem, we propose a routing-resource-based equation which considers the congestion of routing region globally and interference with other nets to determine the routing order of droplets for better routability. We first define the available routing resource $Res^+$ and unavailable routing resource $Res^-$ for droplet $d_i$ as follows:

$$Res^+ = |Bx_t| + |E_x(x_t, y_t)|, \forall x_t \in D/d_i$$

By analyzing the preferred moving vector of each droplet in the bounding box, we can construct the preferred routing tracks. Since the fluidic constraints require a minimum spacing between two droplets, the preferred routing tracks are constructed on non-adjacent rows and non-adjacent columns.

The intuition behind the definition of the routing resource can be described as follows. The available cells inside the bounding box $Bx_t$ are possibly used frequently when routing $d_i$. We attempt to route droplet $d_i$ to the first that have many target cells $t_j$ inside the bounding box $Bx_t$ due to these target cells will become blockages if their droplets $d_j$ are routed. On the contrary, the cells of blockages and unrouted droplets inside the bounding box $Bx_t$ of droplet $d_i$ will have detrimental effects on routability due to the module blockage and fluidic constraints.

To determine the routing order, we calculate the $Res^-$ for each un-routed droplet $d_i$, then the maximum one represents that it is less congested inside its routing region and has the highest routing priority. In the detailed implementation, we use a priority queue $Q_{eq}$ to find the routing order of un-routed droplets efficiently, where the priority of $Q_{eq}$ is the $Res^-$ of each un-routed droplets.

We adaptively pop a un-routed droplet with the highest $Res^+$ to route, and update the $Res^-$ of un-routed droplets $d_i$ left in $Q_{eq}$.
5. INTRA-CONTAMINATION AWARE ROUTING STAGE

5.1 Routing Path Modification by K-Shortest Path

Although the above proposed scheme can schedule a high throughput routing paths in the 2D plane. However, it may cause a large amount of contaminated spots. This is because the droplets are ordered routed on these preferred routing tracks, multiple droplets may be routed along the same tracks to minimize the used cells. Therefore, contamination occurs and causes erroneous outcome for bioassay. To overcome these drawbacks, we proposed a k-shortest-path-based algorithm [5] to reduce the number of contaminated spots. As the aforementioned algorithm, we iteratively select an un-routed droplet \( d_i \) with the highest \( \text{Red}_i \) from \( Q_{cs} \). Then we model the routing path from \( d_i \) to its target \( t_i \) as \( R(d_i,t_i) \) and search a minimum cost path. After \( d_i \) is routed, we push \( d_i \) into a priority queue \( Q_{cs} \), where the priority is the number of contaminated spots within \( R(d_i,t_i) \). And we update the number of contaminated spots for the droplet \( d_j \) \((j \neq i)\) in \( Q_{cs} \), where \( R(d_i,t_i) \cap R(d_j,t_j) \neq \emptyset \). Iteration terminates until all droplets are routed.

When all droplets are routed, we iteratively select a routed-droplet \( d_i \) with the highest number of contaminated spots from \( Q_{cs} \). Then we remove the cost of the edge \((v_e,v_d) \in R(d_i,t_i)\) with a higher value for penalty, where node \( v_e \) and \( v_d \) are contaminated spots. Based on the remodeled routing graph, we adopt the k-shortest path algorithm to reduce the number of contaminated spots. For example, Figure 3 (a) shows the original routing paths and the routing path of \( d_2 \) has the highest number of contaminated spots. We then select \( d_2 \) by adopting the k-shortest path algorithm to the remodeled routing graph. As shown in Figure 3 (b) and (c), when \( k \) is set to be one, the number of contaminated spots is reduced to five; when \( k \) is set to two, the number of contaminated spots is reduced to two. Two are two essential ideas behind our k-shortest path algorithm. Since we still want to make the droplet route on tracks to minimize the used cells, the k-shortest path algorithm can be adopted to find the k-minimum cost path. However, if we only modify the original path once \((k=1)\), it may still overlap with other routed path as shown in Figure 3 (b), which causes only slightly reduction of contaminated spots. Thus, it is necessary to find another minimum cost path to avoid this situation. In this paper, the value of \( k \) is set to be three.

![Figure 3: Adopt the k-shortest path algorithm to reduce the cross section. (a) The original routing solution of \( d_2 \). (b) The first shortest path of \( d_2 \) in the remodeled routing graph. (c) The second shortest path of \( d_2 \) in the remodeled routing graph.](image)

5.2 Routing Compaction by Dynamic Programming

One of the major challenges is to convert the routing paths from sequential to concurrent manner. Since all droplets are routed to A-cell sequentially in the order determined by the routing-resource based equation, however, the completion time may violate the timing constraint. Thus, it is necessary to perform routing compaction that minimize the execution time for fast bioassay execution and better reliability. Another difficult challenge is the determination of the occurring clock cycle of contaminated spots. Since contamination is likely to occur when multiple droplet routes cross or overlap with each other. At the contaminated spot, a droplet that arrives at a later clock cycle can be contaminated by the residue left behind by another droplet that passed through at an earlier clock cycle. Therefore, to perform simultaneous wash operations within the droplet routing, it is desirable to obtain the occurring clock cycle of each contaminated spot. To achieve these goals, we propose a dynamic programming based approach to compact the 2D routing and determine the occurring clock cycle of each contaminated spot. There are at least two advantages by using this approach. First, the original routing paths can be preserved on the preferred routing tracks to minimize the routing detour and the used cells for better bioassay reliability. Second, an efficient and robust routing scheduling for concurrent routing among droplets can be derived due to the elegant property of dynamic programming. Since it is hard to directly apply the 2D routing paths to routing compaction, we encode each routing path into a corresponding 1D moving string by using four direction characters \( u, d, l, r \), where represent up, down, left, and right, respectively. Generally, consider two moving strings \( MS_1 \) and \( MS_2 \), the routing compaction problem is transformed to minimize the length of the compacted string without violating any fluidic constraints. To characterize the optimal substructure of the compacted string, we define \( C[[i]] \) to be the compacted string length using \( j \) prefixes of \( MS_1 \) and \( j \) prefixes of \( MS_2 \). The optimal substructure of the routing compaction gives the following recursive formula:

\[
C[[i]] = \min_{1 \leq j \leq n(i)} \{ C[i-1] + C[j-1], C[i-j-1] + 1 \}
\]

Based on the routing order, we first encode the first routing path \( P_1 \) into the corresponding moving string. Then we iteratively encode the next routing path, and compact it with the previous compacted moving string. During the compaction, we should check the legality of the compacted string, that is, it should satisfy the fluidic constraints for each droplet. If it is legal to compact \( MS_1 \) and \( MS_2 \), the optimal string length \( C[[i]] \) will be the minimum length among \( C[i-1] \), \( C[j-1] \), and \( C[i-j-1] \), which are stored in the table previously, plus the unit length. Otherwise, it will set to be infinite. Finally, iteration terminate when all moving string are compacted.

5.3 Minimum Cost Circulation Flow Technique

Wash droplets can remove the liquid residue and non-target molecules that have been adsorbed onto the surface to prevent the contamination problem. However, if we only adopt one wash droplet to perform the wash operation, the overloading of contaminated particles may reduce the purity of wash droplet thereby decreasing the washing rate. Furthermore, wash droplets are exposed to the environmental temperature and may suffer from the evaporation problem due to the long execution time. Therefore, it is desirable to design a wash-droplet routing algorithm that utilizes multiple wash droplets to shorten the execution time for assay thereby increasing the reliability of DMFBs. Therefore, a new minimum cost circulation-based algorithm (MCC) for optimal wash-droplet routing to simultaneously minimize used cells and the cleaning time.

5.3.1 Introduction to Minimum Cost Circulation Problem

We first briefly introduce the minimum cost circulation problem (MCC). The circulation problem and its variants is a generalization of network flow problems, with the added constraint of a lower bound on edge flows, and with flow conservation also being required for the source and sink. Each arc has an associated cost and the objective function is to find a feasible flow through this graph with the minimum cost, such that the sum over all arcs of the multiplication of the flow in each arc and the corresponding cost is minimum. Given a flow graph \( G = (V, E) \), the minimum cost circulation problem can be represented as follows:

Minimize:

\[
 z = \sum_{(i,j) \in E} C_{ij} \cdot x_{ij}
\]

Subject to:

Bounded constraint:

\[
l_i \leq x_{ij} \leq u_{ij}, \forall (i, j) \in E
\]

Conservation constraint:

\[
\sum_{(i,j) \in E} x_{ij} = \sum_{(j,i) \in E} x_{ij}
\]

where \( l_{ij} \) and \( u_{ij} \) represent the lower (upper) bound on flow from node \( i \) to \( j \), \( C_{ij} \) represent the cost per unit flow from node \( i \) to \( j \), and \( x_{ij} \) represent the flow volume from node \( i \) to \( j \). The MCC problem can be solved in strongly polynomial algorithm by using linear programming or more commonly by using faster and more efficient network algorithms. In this paper, we solve the MCC problem by using negative cycles canceling techniques [6].

5.3.2 Circulation Flow Formulation

The key concept behind our minimum cost circulation formulation is to schedule a routing method for high throughput wash operations. The most difficult challenge is to model the contaminated spots into the flow constraints and remove the cost of the wash operations. To ensure the real-time response and reduce the time-to-result effects, we should minimize the cleaning time used for wash operations. Furthermore, to improve the fault-tolerance of DMFBs, we
should minimize the used cells for wash-droplet routing. Consider the multi-objective optimizations, we propose the MCC-based algorithm to solve the wash-droplet routing problem. We first construct the flow graph \( G_f = (V_f, E_f) \), where \( V_f \) is the set of nodes, and \( E_f \) is the set of edges. Different from the general network flow problem, each edge \((i, j)\) in the flow graph \( G_f \) can be represented as a 3-tuple \((l_{ij}, u_{ij}, C_{ij})\), where \( l_{ij} \) and \( u_{ij} \) denote the lower (upper) bound on the flow of this edge, and \( C_{ij} \) represent the associated cost per unit flow on this edge. As shown in Figure 4, there are four major phases in our MCC formulation, which are a dummy source, set of wash droplets, set of contaminated spots, and a sink (reservoir), respectively. There are two basic assignments and two formulation rules in our MCC formulation, and we assume that the subproblem \( SP_0 \) is under formulation.

**Assignment 1: Node Capacity Assignment**

The most important issue is that the contaminated spot should be cleaned by the wash droplet to prevent the contamination problem. To perform the wash operation correctly (e.g. to schedule the wash droplet to pass through the node), we use the node split technique [1] to solve the contamination problem. We decompose each node \( v_i \in CS \) into two intermediate nodes \( v_{i1} \) and \( v_{i2} \), and an edge is connected from \( v_{i1} \) to \( v_{i2} \). Since we formulate the wash operation as the flow problem, to ensure at least one wash droplet to pass through the node \( v_i \), the 3-tuple of the edge \((v_{i1}, v_{i2})\) is set to be \((l, \infty, 0)\). Note that \( u_{i1} \) is set to \( \infty \) due to multiple wash droplets can pass through the same contaminated spot.

**Assignment 2: Edge Cost Assignment**

Another important issue is to determine the routing path of wash droplets. We also make the wash droplets route on the preferred routing tracks to minimize the used cells. Therefore, we use the same model \( R_{v_i, v_j} \) which is discussed in section 4.1 to represent the routing path from node \( v_i \) to \( v_j \) and the associated cost of this edge \((v_i, v_j)\) is \( Cost(R_{v_i, v_j}) \). After the two basic assignments, the circulation flow can be constructed by the following two flow formulation rules.

### Rule 1: Timing-Based Transitive Topology

Within one subproblem, the contamination happens in a cell when a droplet arrives at a later clock cycle that is contaminated by the residue left behind by another droplet that passed through at an earlier clock cycle. Before a droplet arrive at the cell that is contaminated by the previous droplet, it is desirable to wash the surface to avoid incorrect diagnosis outcomes. By adopting the routing compact technique, the occurring time and location of contaminated spots can be determined. We use a 2-tuple \((v_i, t_i)\), to denotes the occurring time of contaminated spot \( v_i \). Given a set of contaminated spots \( CS_p \), \((v_i, t_i)\), \((v_2, t_2)\), \((v_3, t_3)\), \((v_k, t_k)\), we construct the timing-based transitive topology in the following descriptions.

**(a) Timing-based topology:** For each pair \((v_i, v_j)\) where \((v_i, v_j) \in CS_p \), there exists a edge connection from \( v_{i1} \) to \( v_{j1} \) if and only if \( t_i \leq t_j \). We assign the associated 3-tuple of the edge by \((0, \infty, Cost((R_{v_i, v_j})) \).

**(b) Transitive closure:** For each triple \((v_i, v_k, v_j) \in CS_p \), if there exists edge connections from \( v_{i1} \) to \( v_{k1} \) and from \( v_{k1} \) to \( v_{j1} \), there also exists an edge connection from \( v_{i1} \) to \( v_{j1} \). We assign the associated 3-tuple of the edge by \((0, \infty, Cost(R_{v_i, v_j})) \).

Contamination problem occurs when two droplet pass through the same spot in different clock cycle, the timing-based topology shows that the wash operation should be performed before the second droplet arrives at this contaminated spot. The transitive closure property allows the multiple wash droplets to perform the wash operations, while the timing-based topology should be also followed.

### Rule 2: Connection Strategy Between Phases

There are four major phases in our MCC formulation. We first construct edge connections from the dummy source to wash droplets and the associated 3-tuple of each edge is set to \((0, 1, 0)\). From the second phase to the third phase, for each wash droplet \( v_j \in W \) and \( v_j \in CS_p \), we connect \( v_j \) with \( v_{j1} \) and assign the associated 3-tuple of this edge by \((0, 1, Cost(R_{v_j, v_{j1}})) \), where \( v_{j1} \) is the location of wash droplet \( v_j \). Then, for each node \( v_i \in CS_p \), there exists an edge connection from \( v_{i1} \) to the sink \( v_{sink} \) (waste reservoir), with the associated 3-tuple \((0, \infty, Cost(R_{v_i, v_{sink}})) \). Finally, since we should use at least one wash droplet to perform the wash operation, we connect the sink back to the dummy source, with the associated 3-tuple \((1, 4, 0)\). Note that in this paper, the maximum number of available wash droplets is four.

Figure 4 shows the flow construction. Based on the aforementioned basic assignments (node capacity and edge cost) and the two flow formulation rules, we have the following two theorems.

**Theorem 1.** There exists a feasible solution under the two basic assignments and two flow formulation rules.

**Theorem 2.** Under the proposed flow construction, we can adopt the minimum cost circulation algorithm to solve optimal and correct wash operations.

### 6. INTER-CONTAMINATION AWARE ROUTING STAGE

**6.1 Look-Ahead Routing Scheme**

Although wash droplets clean the intra-contaminations within one subproblem optimally by using the MCC-based algorithm, some contaminations that treated as blockages for the next subproblem (i.e. inter-contaminations) may incur timing overhead since extra execution time for wash operations should be performed between successive subproblems. To remedy these deficiencies, we propose a look-ahead routing scheme that predicts the inter-contaminations for the next subproblem then simultaneously clean both intra- and inter-contaminations by using the MCC-based algorithm to reduce the execution time significantly. For example, before cleaning the intra-contamination of subproblem \( SP_j \), using wash droplets, we also derive the routing solution of subproblem \( SP_{j+1} \). Based on both routing solutions, the inter-contaminations between subproblem \( SP_j \) and \( SP_{j+1} \) can be derived and are handled with intra-contaminations in subproblem \( SP_p \).

**6.1.1 Travelling Salesman Problem Formulation**

Consider each edge \((v_i, v_j)\) in the flow graph, we perform the wash operation for those inter-contamination nodes \( v_k \in L(v_i, v_j) \). Since each \( v_k \) should be cleaned by a wash droplet, which means a wash droplet should pass through all the nodes \( v_k \). Furthermore, we also encourage the wash droplet to route on the preferred routing tracks to minimize the total used cells. To tackle these problems, we construct a TSP graph \( G_{tsp} = (V_{tsp}, \{v_i, v_j\}, E_{tsp}) \), where the vertex set of \( V_{tsp} \) is the node \( v_k \in L(tsp, v_i, v_j) \), the edge set, and the \( v_i(v_j) \) is the starting (ending) point of the TSP graph. We connect each pair \((v_k, v_k)\) where \( v_k \in V_{tsp} \) with the associated cost \( Cost(R_{v_k, v_k}) \). Then, we connect the starting point \( v_i \) with each node \( v_k \) and connect each node \( v_k \) with the ending point \( v_j \). The goal is to schedule a minimum cost routing path of the wash droplet that starts from \( v_i \) to \( v_j \) while passing through all the nodes \( v_k \), which is a variant of travelling salesman problem (TSP). Assume the optimal routing path is \( R_{tsp}^{opt} = (v_k_1, v_k_2, \ldots, v_k_{r_{tsp}}) \) and the corresponding minimum cost is \( Cost(R_{tsp}^{opt}) \). We re-formulate the cost of the edge \((v_i, v_j)\) in \( G_f \) by \( Cost(R_{tsp}^{opt}) \), and the corresponding routing path of \((v_i, v_j)\) is \( R_{tsp}^{opt} \). The dynamic-programming-based approach for solving the TSP problem is used in this paper [2].

After the wash droplets are scheduled, we compact the routing paths of wash droplets with the previous compacted paths by using the same dynamic programming technique. In addition to checking the fluidic constraints, we add the contamination constraint to ensure that the wash droplet should arrive before the contaminated spots.
7. EXPERIMENTAL RESULTS

We have implemented our contamination-aware droplet router in the C++ language on a 2-GHz 64-bit Linux machine with 8GB memory. We perform experiments to verify the efficiency and effectiveness of our algorithm on four widely used bioassays from [9, 11]. Table II shows the statistics of each benchmark. In the table, “Size” gives the size of microfluidic array, “#Sub” denotes the number of subproblems, “#Net” represents the number of nets, “Dmax” gives the maximum number of contaminations among subproblems, and “#W” denotes the number of wash droplets.

In the first experiment, we evaluated the reduction of intra-contaminations by using the k-shortest path technique. Table III shows the comparison results of our algorithm without and with the k-shortest path technique. By applying the k-shortest path technique, our algorithm not only reduced the intra-contaminations by 48%, but also reduced the respective used cells and execution time by 8% and 13% with small increase in the CPU time.

In the second experiment, we evaluated the reduction of the execution time by using the look-ahead routing scheme. Table IV shows the comparison results of our algorithm without and with the look-ahead routing scheme. The look-ahead routing scheme predicts the inter-contaminations for the next subproblem then simultaneously cleans both intra- and inter-contaminations by using the MCC-based algorithm to reduce the execution time significantly. By applying the look-ahead routing scheme, our algorithm not only reduced the execution time by 9%, but also reduced the used cells by 11% with small increase in the CPU time.

In the third experiment, we compared the effectiveness of our contamination-aware droplet router algorithm with the state-of-the-art algorithm [12] in Table V. For comparison purpose, we implement the state-of-the-art algorithm [12]. Overall, our algorithm reduced the respective used cells and execution time by 28% and 12% with small increase in the CPU time. The increase in the intra-contaminations is as expected, because our droplet router has to route on the pre- ing algorithms to minimize the contaminated used cells and the execution time for better reliability and fast bioassay execution.

8. CONCLUSION

In this paper, we proposed a contamination-aware droplet routing algorithm for DMFBs. To reduce the routing complexities and used cells, we first construct preferred routing tracks by analyzing the global moving vector of droplets to guide the droplet routing. To cope with contaminations within one subproblem, we first apply a k-shortest path routing technique to minimize the contaminated spots. Then, to take advantage of multiple wash droplets, we adopt a minimum cost circulation algorithm (MCC) for optimal wash-droplet routing to simultaneously minimize used cells and the cleaning time. Furthermore, a look-ahead prediction technique is used to determine the contaminations between successive subproblems. After that, we can simultaneously clean both contaminations within one subproblem and those between successive subproblems by using the MCC-based algorithm to reduce the execution time significantly. Based on four widely used bioassays, our algorithm reduced the used cells and the execution time significantly compared with the state-of-the-art algorithm.

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10. REFERENCES