A Fast Multilayer General Area Router for MCM Designs

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Abstract-In order to reduce interconnection delay and increase packaging density, the multichip module (MCM) technology is used in the design of high-performance VLSI systems. A commonly used method for multilayer MCM designs is the three-dimensional (3-D) maze routing, which suffers from a number of problems: it is very sensitive to the net ordering, it requires long computation time, and it often results in a large number of vias in the routing solutions. The objective of this research is to develop an efficient multilayer general area router as an alternative to the 3-D maze router for solving the multilayer MCM routing problem. Our router, named SLICE, is independent of net ordering, requires much shorter computation time, and uses fewer vias. A key step in our router is to compute a maximum noncrossing bipartite matching, which is solved optimally in $O(n \log n)$ time where n is the number of possible connections. We tested our router on a number of examples, including two MCM designs from MCC. The total wirelength used by SLICE is only a few percent away from the optimal on average. Compared with a 3-D maze router, SLICE is six times faster and uses 29% fewer vias. Another feature of SLICE is that it works on only a "thin slice" of a two-layer routing grid at a time, while a 3-D maze router works on the entire 3-D routing grid. Therefore, SLICE can successfully produce solutions for large MCM routing examples where 3-D maze routers fail due to insufficient memory.

I. INTRODUCTION

S VLSI fabrication technology advances, interconnections and packaging (P/I) technologies have become a bottleneck in system performance [1]-[3]. In the traditional approach, each chip is first packed into single chip packaging (SCP) and then mounted on a printed circuit board (PCB). The area of each SCP is usually several times larger than the corresponding bare chip. As a result, the packing density is severely limited. Moreover, there exist two levels of inter-chip interconnections: the connections on SCP's and the connections on the PCB. The wasted space and the addition level of interconnections limit the packing density and degrade the system performance.

The multichip module (MCM) technology reduces the wasted space on a board and eliminates a level of interconnection by assembling and connecting bare chips on a common substrate. The substrate consists of multiple routing layers used for inter-chip interconnections. Without individual packaging for the chips, the bare chips can

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be placed much closer on the MCM substrate, which leads to a significant increase in packing density and decrease in interconnection delay.

Due to the high packing density in MCM designs, the MCM routing problem is more difficult than the conventional IC or PCB routing problems. First, MCM's may have far more interconnection layers than IC's. For example, the MCM developed for the IBM 3081 mainframe has 33 layers of molybdenum conductors (including one bonding layer, five distribution layers, 16 interconnection layers, eight voltage reference layers, and three power distribution layers [4], [5]). Fujitsu's latest supercomputer, the VP-2000, uses a ceramic substrate with more than 50 interconnection layers [6]. Moreover, unlike routing in IC's, where the entire routing region can be naturally decomposed into channels and switchboxes, there is no natural routing hierarchy in MCM routing. The MCM routing problem is an immense three-dimensional (3-D) general area routing problem where connection can be carried out almost everywhere in the entire multilayer substrate. Finally, the pitch spacing is much smaller and the routing result is much denser in MCM routing as compared to those in conventional PCB routing. Thus, traditional PCB routing tools are often inadequate in dealing with MCM designs.1

Few methods are available for MCM routing. A commonly used method for multilayer MCM designs is the 3-D maze routing [6], [7]. Although this method is conceptually simple to implement, it suffers from several problems. First, the quality of the maze routing solution is very sensitive to the ordering of the nets being routed, yet there is no effective algorithm for determining a good net ordering in general. Moreover, since each net is routed independently, global optimization is difficult and the final routing solution often uses a large number of vias despite the fact that there are many interconnection layers. Finally, 3-D maze routing requires long computational time and large memory space. For example, one industrial example that we obtained from MCC has a 75-micron routing pitch and a routing area of 174 × 174 mm²; this results in a routing grid of 2032 × 2032 for a single layer! It is certainly not a trivial task to store such a grid for each layer and search in it efficiently.

¹Beside the problem of efficient utilization of routing resource, there are also several performance issues involved in MCM routing. For example, for high-performance designs, the wires need to be modeled as lossy transmission lines, where signal reflection and cross-talk need to be taken into consideration.

Another method for multilayer MCM routing is to divide the routing layers into a number of x-y layer pairs. Nets are first assigned to x-y layer pairs and then two-layer routing is carried out for each x-y layer pair (the x-layer runs horizontal wires and the y-layer runs vertical wires) [8]. Although this approach is efficient, it faces a few problems. First we have to predetermine the number of the routing layers before we can carry out layer assignment. Moreover, the approach does not take advantage of the existence of the large number of routing layers. Thus some nets may use many vias since they are forced to be routed within two layers. For high-performance MCM designs, vias not only increase the manufacture cost but also degrade the system performance since they form inductive and capacitive discontinuities and cause reflections when the interconnections have to be modeled as transmission lines [2].

Several efficient routers have been proposed for siliconon-silicon based MCM technology [1], [9]–[11]. Since the number of signal routing layers is usually small (2 to 4 layers) in this technology, some techniques for IC routing, such as hierarchical routing and rubber-band routing, can be applied to yield good solutions.

The objective of our research is to develop an efficient multilayer general area router as an alternative to the commonly used 3-D maze router for solving the multilayer MCM routing problem. Our router, named SLICE, has a number of advantages. First, it processes many nets simultaneously so that the routing solution is independent of net ordering. Moreover, it requires much shorter computation time and much smaller memory storage. Finally, it emphasizes planar routing so that most of the nets use very few vias. A key step in our method is to compute a maximum noncrossing bipartite matching, which is solved optimally in $O(n \log n)$ time (where n is the number of possible connections). We tested our router on a number of examples, including two MCM designs from MCC, and compared the results with those by a 3-D maze router. On average, both routers use about the same total wirelength, but the 3-D maze router is six times slower and uses 29% more vias. Another important feature is that SLICE works on only a "thin slice" of a two-layer routing grid at a time, while a 3-D maze router works on the entire 3-D routing grid. Therefore, SLICE can successfully produce solutions for large MCM routing examples where 3-D maze routers fail due to the memory requirement.

The remainder of this paper is organized as follows. Section II formulates the multilayer MCM routing problem. In Section III, we give an overview of our algorithm and we describe each step of the algorithm in detail. Experimental results and a comparative study are presented in Section IV. Finally, we discuss the extension of our work in Section V.

II. PROBLEM FORMULATION

The MCM routing problem consists of a set of modules, a set of nets, and a multilayer routing substrate. Modules (dies) are mounted on the top of the substrate by wire

bonding, tape-automated bonding (TAB), or flip-chip bonding with solder bump connections. The substrate consists of multiple signal routing layers, with (possible) obstacles in some routing layers, such as power/ground connections and thermal conducting vias. The I/O terminals (pads) of the modules are connected to the substrate either directly or through routing to the external pads that surround the individual modules for engineering changes [2]. The pads are brought to the first signal routing layer either directly through distribution vias or through one or more redistribution layers. The redistribution layers are required when the pads are too dense to be connected directly to the signal routing layers. A pin redistribution algorithm was presented in [12]. The goal of our MCM router is to complete the connections for the I/O terminals in each net using the signal routing layers in the substrate.

The signal routing layers in the substrate are numbered from top to bottom. We assume that there is a routing grid superimposed on each routing layer where the spacing between grid lines is determined by the routing pitch for the given P/I technology. We assume that the routing grid is a Manhattan grid. However, our algorithm can handle 45° routing as well. Two wires in adjacent signal routing layers can be connected by a via. Vias may be stacked on top of each other to connect wires in nonadjacent layers. Stacked vias can be formed in several ways, e.g., by filling the etched via with nickel in the AT&T AVP process or by plating copper posts as in the MCC process [13]. Fig. 1 shows a cross-section of a sample four-layer routing region.

The output of the routing problem is a set of routing segments and vias that connect each net. The quality of the routing can be measured by the total wirelength, the number of vias, the number of wire bends (jogs), and the number of layers required to complete the routing. Long wire paths increase propagation time and should be avoided. Vias and wire bends degrade the signal's fidelity by introducing impedance discontinuity in signal paths thus should also be minimized. Vias usually cause more serious problems than jogs, so that our router gives via minimization a higher priority. Each additional routing layer increases the manufacturing cost, and thus the number of layers should also be minimized.

III. DESCRIPTION OF THE ALGORITHM

In this section, we present our fast multilayer general area router, called SLICE, for MCM and single-sided PCB designs. We first give an overview of the entire algorithm, and then we describe each step in detail.

3.1. Overview of the Algorithm

The basic idea behind our algorithm is to perform planar routing on a layer by layer basis. After routing on one layer, we propagate the terminals of the uncompleted nets to the next layer. Then we continue routing on the next layer and perform the single layer routing again. We repeat the process until all the nets are routed.

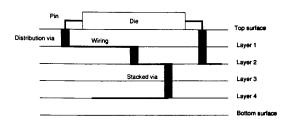


Fig. 1. Cross-section of a multilayer (4-layer) routing region.

A crucial part of our algorithm is to compute a planar routing solution for each layer and try to connect as many nets as possible in that layer. For nets that cannot be completed in the layer, we try to perform a partial routing so that these nets can be completed in the next layer with shorter wires. We scan the routing region from left to right and process each pair of columns that has terminals at a time. For each adjacent column-pair, we compute a maximum weighted noncrossing matching (MWNCM) which consists of a set of noncrossing edges that extend from the left column to the right column. This gives us a topological planar routing solution between the columnpair. Next, we generate the physical routing between the current column-pair based on the selected edges in the noncrossing matching. Then, we move on to the next column-pair and compute the noncrossing matching and physical routing again for that column-pair. The planar routing process is completed when the right end of the current layer is reached. Then, we distribute the terminals of the uncompleted nets so that they can be propagated to the next layer without causing local congestions. Since the left-to-right scanping operation in the planar routing results in mainly (horizontal) wires in the planar routing solution, in order to complete the routing in the vertical direction, we use a restricted two-layer maze router which is much faster than a general maze router to route within a thin vertical slice of the substrate one at a time from left to right.² We clean up the routing solution by removing unnecessary jogs and wires in the current layer. For nets that are not completed, their terminals are propagated to the next layer. Finally, we rotate the routing region by 90° so that the scanning direction in the next layer is orthogonal to the one used in the current layer. The entire process is iterated until all the nets are routed. This top level algorithm is summarized in Fig. 2. We shall describe each step in detail in the remaining subsections.

3.2. Planar Routing

The terminals that lie on the same vertical grid line form a *column*. In our planar routing algorithm, we scan the routing region across from left to right and perform routing between each column-pain. Let x_l and x_r be the x-coordinates of the left and right column of the current column-pair, respectively. Conceptually, a column-pair

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Algorithm SLICE N = \{\text{list of nets}\}; l = 1; l^* Start at layer 1 = l while (N \neq \emptyset) do Compute planar routing on layer l; Redistribute the uncompleted terminals on layer l; Perform restricted maze routing on layer l; Remove unnocessary jogs.

Propagate the terminals of the uncompleted nets to layer l + 1; Rotate the routing area by 90 degrees: l = l + 1; end end
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Fig. 2. Overview of the SLICE router.

forms a channel and we define the channel capacity to be $C_{\text{cap}} = x_r - x_l$. During planar routing in the current layer, the terminals of the uncompleted nets are put in the list P_{prop} . These terminals will be propagated to the next routing layer.

For each column-pair, the occupied grid points on the left column are called <u>start-points</u>. Clearly, each start-point is either a terminal propagated from the previous layer, or the endpoint of a partial routing solution computed in the previous channel. We denote a start-point n_i in the current layer by a triple $n_i = (x_i, y_i, net_i)$, where (x_i, y_i) is the coordinates of the point, and net_i is the net number of the point. For a start-point n_i , the terminal that it is to be connected to is called the <u>target</u> of n_i , denoted by $target(n_i)$.

We shall concentrate our discussion on routing between a single column-pair. We begin with a list P_l that contains all the start-points on the left column and go through the following three steps to complete the planar routing. 1) For all start-points on the left column of the current column-pair, we generate a set S of weighted edges that connect these start-points to the right column. The weight for each edge represents the gain if we include this connection in the planar routing solution. 2) We compute the maximum weighted noncrossing matching S_{MWNCM} of S, which corresponds to the best topological planar routing solution between the current column-pair. We shall show that this step can be carried out optimally in O(n) $\log n$) time, where n is the number of edges in S. 3) Finally, we compute the physical routing solution based on the edges in S_{MWNCM} . The steps in the planar routing algorithm are illustrated in Fig. 3, where the net number for each terminal is given besides the terminal, and there are three column-pairs. Routing in the first column-pair has been completed, and we are processing the second column-pair. Fig. 3(a) shows the weighted edges extending from the start-points on the left column to the right column, Fig. 3(b) shows the edges selected in the maxi-

²Note that the scanning direction in the next layer will be orthogonal to the scanning direction in the current layer, which will also help to complete nets that require mainly vertical wires.

³As a preprocessing step of SLICE, we decompose each multiterminal net into a set of two-terminal subnets based on the Prim's minimum spanning tree algorithm. Therefore, each start-point always has a well-defined target. The prior decomposition does not affect the routing quality very much since 1) most nets are two-terminal nets in MCM routing; 2) we allow the routes of two subnets of the same net to meet and form a Steiner point in our planar routing procedure.

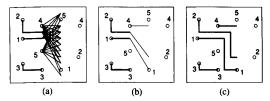


Fig. 3. Steps in planar routing for a column-pair. (a) Generate possible (topological) connections. (b) Compute a maximum weighted noncrossing matching. (c) Generate physical routing.

mum weighted noncrossing matching, and Fig. 3(c) shows the physical routing based on the selected edges. We now describe these steps in detail.

3.2.1. Generating the Weighted Edges: Given a list of start-points P_i on the left column, we want to generate the set S of weighted edges that connects the start-points in P_l to the grid points on the right column. Conceptually, for a start-point $n_i = (x_i, y_i, net_i)$ in P_l , we can generate an edge from (x_i, y_i) to every grid point on the right column which is free or occupied by a terminal of net n_i . However, this may result in too many edges, and most of them will have very little chance of being selected in the maximum noncrossing matching in the next step. To conserve both memory and time, we use a simple heuristic, called range reduction, to reduce the number of edges that are generated. Clearly, the channel capacity C_{cap} bounds the density of the vertical segments that can be routed between a column-pair. Let y_{i+n} and y_{i-n} be the y-coordinates of the nth start-point (not grid point) above and below n_i on the left column, respectively. Now if we assume that all the start-points on the left column will be routed, then the connection for each start-point above or below n_i will increase the channel density by one. Therefore, it is sufficient to generate edges whose right endpoints are within the y-interval $[y_{i-C_{ap}}, y_{i+C_{cap}}]$. This is the reduced range where the edges for n_i can end on the

The weight of each edge represents the gain if we include the edge in our planar routing solution. For each start-point n_i , let $n_i = target(n_i)$. We define the preferred region to be the y-interval on the right column defined by the y-coordinates of n_i and n_j . Clearly, if an edge from n_i ends within the preferred region, it does not increase the wirelength of net_i. So we assign a high weight, weight_preferred, to the edges ending in the preferred region. Moreover, if an edge from n_i ends exactly on n_i (when n_i is on the right column), we assign an even higher weight, weight_completed, to the edge in favor of completion of the net. For the edges that end outside the preferred region, we assign them a small weight weight_outside. In general, weight_completed > weight_preferred > weight_outside. We experimented with several weighting functions and the best choice is the following. We set weight_completed and weight_preferred to be two constants, and let weight_outside decrease linearly as the right end of the edge moves away from the preferred region.

3.2.2. Computing the Maximum Weighted Noncrossing Matching: The most important part of our planar routing algorithm is computing a topological planar routing solution between each column-pair. We begin with a set of weighted edges $S = \{s_1, s_2, \dots, s_n\}$. Each edge in S represents a possible topological route that extends from a start-point to the right column. The weight for each edge represents the gain if we include this route in the planar routing solution. Each edge s_i is a four-tuple (l, r, w, net)where l is the y-coordinate of the left end of the edge, ris the y-coordinate of the right end of the edge, w is the weight of the edge, and net is the net number of the edge. Since we want to choose a set of best edges that can be routed on the current layer, we need to select a set of edges from S that are non-crossing and have the maximum total weight. This is the maximum weighted noncrossing bipartite matching (MWNCM) problem. However, in our formulation, we permit two edges in a noncrossing matching to share a common endpoint at the right column if they belong to the same net.

In order to compute a MWNCM, we map each edge (l, r, w, net) to a point in the x-y plane using the one-toone mapping (x, y) = (l, r), where (x, y) is the position of the point in the x-y plane. Given two points $p_i = (x_i, y_i)$ and $p_i = (x_i, y_i)$ in the x-y plane, p_i dominates p_j if i) $x_i \ge x_i$ and $y_i > y_i$, or ii) $x_i \ge x_j$ and $y_i = y_j$ and $net_i = net_j$ (note that net_i is the net that the corresponding edge of p_i belongs to). If condition i) is satisfied, we say that p_i strictly dominates p_i ; otherwise, when condition ii) is satisfied, we say that p_i laterally dominates p_i . The dominance relations are illustrated in Fig. 4, where edge c strictly dominates edges a and b, and edge b laterally dominates edge a (assuming that a and b are of the same net). Clearly, if p_i dominates p_j , the two edges that are mapped to p_i and p_i are either strictly noncrossing or sharing the same endpoint on the right column when they are of the same net. $\overline{}^{4}$ We define a *chain* among a set of points P in the x-y plane, to be an ordered list of points C = $\{p_1, \overline{p_2, \dots, p_m}\}$ where each $p_k \in P$, and p_{k+1} dominates p_k for $k = 1, 2, \dots, m - 1$. We call p_m the head-node of the chain. We define the weight of a chain C, denoted by weight(C), to be the sum of the weights of the points in C. We define the maximum-chain C_{max} to be the chain that has the maximum weight among a given set of points. We then have the following results.

Lemma 1: The dominance relation is transitive.

Proof: Suppose that p_i dominates p_j and that p_j dominates p_k . There are four possibilities: i) p_i strictly dominates p_j and p_j strictly dominates p_k ; ii) p_i strictly dominates p_j and p_j laterally dominates p_k ; iii) p_i laterally dominates p_j and p_j strictly dominates p_k ; and iv) p_i laterally dominates p_j and p_j laterally dominates p_k . It is straightforward to verify that in cases i)-iii), p_i strictly



⁴Note that when two edges of the same net share a right endpoint, a Steiner point is formed. So our method can generate Steiner routing trees automatically.

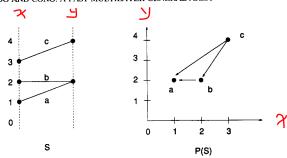


Fig. 4. Mapping the edges to points in the plane. Edges in P(S) indicate the dominance relations (assuming the start-points 1 and 2 are of the same net).

dominates p_k , and in case iv) p_i laterally dominates p_k . Therefore, the dominance relation is transitive.

Theorem 1: Let S be the set of edges between a column pair and P(S) be the set of corresponding points on the x-y plane. Then, P(S) forms a partially ordered set. Moreover, the set of edges M in S is a maximum weighted noncrossing matching if and only if the corresponding points P(M) form a maximum-chain in P(S).

Proof: It is easy to see that the dominance relation is reflexive (i.e., p_i dominates p_i itself) and antisymmetric (i.e., if p_i dominates p_j and p_j dominates p_i , then $p_i = p_j$). Also, according to Lemma 1, the dominance relation is transitive. Therefore, the point set P(S) with the dominance relation forms a partially ordered set [14].

According to the definition of the dominance relation, it is easy to verify that two edges in M are noncrossing if and only if the two corresponding points in P(M) are related by the dominance relation (i.e., one dominates the other). Therefore, the set of edges $M \subseteq S$ forms a noncrossing matching if and only if any two points in P(M) are related by the dominance relation. Since P(S) is a partially ordered set, any two points in P(M) are related by the dominance relation is equivalent to that P(M) is a chain in P(S). Moreover, since the weight of the chain P(M) is the same as the weight of the edge set M, we conclude that M is a maximum weighted noncrossing matching if and only if P(M) is a maximum-chain in P(S).

A maximum-chain of a point set P can be computed as follows: We construct a directed graph G_p , called the dominance graph, in which each node represents a point in P, and add an edge (p_i, p_j) to G_P if and only if point p_i dominates point p_i . Fig. 5 shows a set of edges and the dominance graph on the corresponding point set (the edges implied by the transitive relation are omitted for clarity). It is not difficult to show that G_P thus constructed is a directed acyclic graph and a maximum-chain in P corresponds to a maximum weighted path in G_p . Since the maximum weighted path in a directed acyclic graph can be computed in $O(n^2)$ time, where n is the number of nodes in the graph [15], we can compute a MWNCM in $O(n^2)$ time, where n is the number of edges we generated between a column-pair. The maximum weighted edges in S and the maximum weighted path in P(S) are shown in

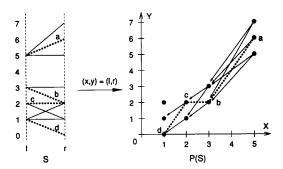


Fig. 5. A set of edges S and its corresponding point set P(S). Note that start-points 2 and 3 are of the same net. The dominance relations are shown in P(S) as thin arrows. The MWNCM in S, and the corresponding maximum-chain in P(S) are shown as dotted lines.

Fig. 5 as dotted edges. However, since the procedure for computing a maximum weighted noncrossing matching will be used for every column-pair, we seek for a more efficient implementation. We have developed an $O(n \log n)$ time algorithm for computing the MWNCM based on a data structure called the priority search tree [16]. Before we describe the algorithm in detail, we first state a lemma.

Lemma 2: Suppose that each point in P has a positive weight. Then, if point p laterally dominates point q in a maximum-chain in P, there does not exist a point r such that p laterally dominates r and r laterally dominates q.

Proof: If such a point r exists, we can add it into the maximum-chain to get a chain of even larger weight, which leads to a contradiction.

According to Lemma 2, if point p laterally dominates point q in a maximum-chain, then q is the first point of the same net left of p in the same row. This fact is used in the construction of a maximum-chain by our algorithm.

Let P be the corresponding point set of the given set of edges, we shall compute a maximum chain in P under the dominance relation. The points in P having the same x-coordinate form a column, and the points in P with the same y-coordinate form a row. Our algorithm processes the points on a row by row basis, and we process the points in the same row from left to right. This guarantees that when we are processing a point, all the points which are dominated by the current point have been processed already. During the execution of the algorithm, we maintain a binary priority search tree, called PTREE. Each leaf L of PTREE corresponds to a column occupied by a point in P, and it has three fields, L.x, L.weight, and L.head. The field L.x stores the x-coordinate of the column. During the execution of our algorithm, assume that p is the highest point that we have processed so far at column L.x, then L.weight is the weight of the maximum chain among the points that are strictly dominated by p or in the same column below p, and L.head is the head node of that maximum chain (note that p may not be the head node). We shall show how to maintain L.weight and L.head later on in the algorithm. Each internal node I of PTREE has a field I.weight which records the largest weight of the leaves that are in the subtree subtree(I) rooted at I. (Clearly, if X and Y are the two children of I, then $I.weight = \max(X.weight, Y.weight)$.) Fig. 6 shows an instance of the PTREE and the point set in the x-y plane. The weights for the leaf-nodes in PTREE and the points in P are also given in the Fig. 6. For example, when we start processing point q, we have H.weight = 27 and H.head = a since the maximum chain among the points strictly dominated by q or in the same column below q is (a, b, c).

Our algorithm processes the points in P one row at a time from bottom to top, and processes the points in the same row from left to right, so we sort all the points according to their y-coordinates first and then x-coordinates. We maintain four fields for each point p: p.weight, p.net, $p.total_weight$, and p.next. The fields p.weight and p.net store the weight and the net number of the edge corresponding to the point, respectively. The field $p.total_weight$ stores the weight of the maximum-chain C_p with p as its head node, and p.next points to the next point after p in C_p . Initially $p.total_weight = 0$ and p.next = nil for all points.

For each point p, $p.total_weight$ can be determined as follows: let leaf L in PTREE correspond to the column where p is located. Let $PATH_L$ be the path from L to the root in PTREE. Let l_1, l_2, \cdots, l_k be the roots of the left subtrees hanging from the path $PATH_L$ (i.e., l_i is the left child of some node in $PATH_L$). The algorithm searches for leaf L' such that

$$L'.weight = \max_{i=1}^{k} l_i.weight. \tag{1}$$

Then, we have $p.total_weight = p.weight + L'.weight$ and p.next = L'.head. This covers the case where p strictly dominates p.next. To cover the case that p laterally dominates p.next, we look for the point q in the same row as p with q.net = p.net. According to Lemma 2, we need only to consider such a q which is closest to p. If $p.total_weight < p.weight + q.total_weight$, then we set $p.total_weight = p.weight + q.total_weight$ and p.next = q. Fig. 6 illustrates the computation of $p.total_weight$ and p.next for the point p. The leaf node L in PTREE corresponds to the column where p is located. The path $PATH_L$ from L to the root is shown by the dashed line. The nodes F and H are the roots of the left subtrees hanging from $PATH_L$. According to (1),

$$L'.weight = \max(F.weight, H.weight)$$

= $\max(17, 27) = 27$.

If point q is not in the same net as p, then

$$p.total_weight = p.weight + L'.weight = 15 + 27 = 42.$$

If q and p are in the same net, then

$$p.total_weight = max(q.total_weight, L'.weight)$$

$$+ p.weight = max(47,27) + 15 = 62.$$

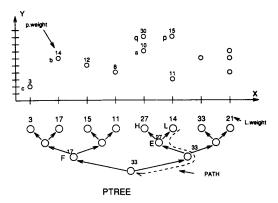


Fig. 6. Priority search tree (PTREE) used to compute the MWNCM.

(Note that $q.total_weight = q.weight + F.weight = 30 + 17 = 47$, which was computed when q was processed in the previous step.)

After we have processed a row of points, we update the entries in PTREE. (Note that we do not update the node leaf L immediately after we have processed the current point p, because we use L.weight to record the weight of the maximum-chain among the points that are strictly dominated by p or are in the same column below p.) For each point q in the current row, let leaf L_q correspond to the column that q is located; if $q.total_weight > L_q.weight$, then $L_q.weight = q.total_weight$ and $L_q.head = q$. Furthermore, we update the weights of the internal nodes in $PATH_L$.

After all the points are processed, the point p_m with the largest $total_weight$ is the head of the maximum chain in P and we can follow the $p_m.next$ field to get the rest of the maximum chain. Our algorithm is summarized in Fig. 7.

Theorem 2: Given the set S of n edges between a column-pair, the maximum weighted noncrossing matching can be computed in $O(n \log n)$ time.

Proof: We shall show that our algorithm spends $O(\log n)$ time for processing each point p in P. Since PTREE is a binary priority search tree, it has depth $O(\log n)$. Thus, there are $O(\log n)$ left subtrees hanging from PATH₁. We can find out the subtree whose root, say l_i , has the maximum weight in $O(\log n)$. Moreover, we can find the leaf node L' in the subtree $subtree(l_i)$ with L'.weight = l_i .weight in $O(\log n)$ time. (In order to find such a leaf node L', we start with l_i as the current node and always move to the child who has the same weight as the current node. When we eventually hit a leaf node, we return it as L'.) Furthermore, to locate the point q, which is immediately laterally dominated by p, takes only constant time. (We preprocess each net and record for each point p the point which is immediately laterally dominated by p. This preprocessing can be done in linear time.) Therefore, p.total_weight can be computed in- $O(\log n)$ time for each point p in P.

When p is the last point in a row, we update the leaf node L_q in *PTREE* for each point q in that row. More-

```
Algorithm Compute MWNCM
                      ..., p_n; /* The set of (x, y) points mapped from S */
  PTREE = A priority search tree with leaves associated with the x-coor
                                                                                               ates of the points in
  Sort P according to p.y followed by p.x:
  foreach n in P
      let leaf L in PTREE correspond to the column where p is located:
     let PATH_L be the path from L to the root in PTREE; let l_1, l_2, ..., l_k be the roots of the left subtrees hanging from PATH_L;
      locate the leaf node L' such that L'. weight = max l_i. weight;
      let q be the rightmost point in the same row as p with q.net = p.net;
      if (q exists) and (q.total_weight > L'.weight) then

p.total_weight = q.total_weight + p.weight and p.next = q;
         p.total\_weight = L'.\_weight + p.weight and p.next = L'.head
      if (p is the last point in a row) there
          for each q in the row do
            let leaf L_q correspond the column where q is located; if (q.total\_weight > L_q.weight) then
                 L_a.weight = q.total weight;
               update the weights of the internal nodes in PATH_{L_n};
            endif
      endif
    let p. be the point with the largest p.total weight;
             of the maximum chain C_{\text{max}} by following the p_m next field;

p_m = 1 the edges corresponding to the points in C_{\text{max}}; end
    S_{MWNCM} = the edges corresp
```

Fig. 7. Algorithm for computing the MWNCM.

over, we update the weights of the internal nodes in $PATH_{L_q}$ (i.e., we start with L_q as the current node X. Let parent(X) be the parent node of X in PTREE, then we update parent(X).weight by max(parent(X).weight, X.weight). We assign X = parent(X) and repeat the updating operation until parent(X) becomes the root of PTREE). Since the length of each $PATH_{L_q}$ is $O(\log n)$, our algorithm spends $O(\log n)$ time updating PTREE for each point q in P.

Therefore, our algorithm spends at most $O(\log n)$ time processing each point p in P. Hence, the time complexity of our algorithm is bounded by $O(n \log n)$.

We noticed a significant speed-up when the $O(n^2)$ time algorithm was replaced by the $O(n \log n)$ time algorithm for computing a maximum weighted noncrossing matching between each column-pair.

3.2.3. Physical Routing: The solution we obtained from computing the maximum weighted noncrossing matching in the previous section gives us a set of edges, S_{MWNCM} = $\{s_1, s_2, \dots, s_n\}$, where $s_i = (l_i, r_i, net_i)$, l_i and r_i being the y-coordinates of the left and right endpoints of the edge, and net; is the net number of the edge. Each edge represents a connection between the two points, (x_l, l_i) and (x_r, r_i) in the current routing plane, where x_i and x_r are the x-coordinates of the left and right columns, respectively. As the result of the physical routing, we end up with a list of endpoints of the routings, P_r , on the right column, which together with the terminals that are on the right column, will form the new list of start-points on the new left column in the next iteration of column-pair routing. Because the edges in S_{MWNCM} are topological routing solutions, not all edges can be routed due to the channel capacity constraint. We add to P_{prop} the start-points whose edges failed to be routed.

We perform the routing separately on two classes of edges from S_{MWNCM} as defined below. Given an edge $s_i = (l_i, r_i, net_i)$, we say that s_i is a rising edge if $l_i < r_i$. Otherwise, we say that s_i is a falling edge (i.e., $l_i \ge r_i$). We group all the rising edges in S_{rise} and all the falling edges in S_{fall} . We also order the edges in S_{fall} in increasing y-coordinates, and order the edges in S_{rise} in decreasing y-coordinates. That is, if S_{rise} or $S_{fall} = \{s_1, s_2, \dots, s_n\}$, then for S_{fall} , $l_i < l_{i+1}$ for $i = 0, \dots, n-1$, whereas for S_{rise} , $l_i \ge l_{i+1}$ for $i = 0, \dots, n-1$. It is not difficult to show that the edges in S_{rise} and S_{fall} can be routed separately.

We perform the physical routing one edge at a time. We now describe the routing for S_{rise} . For each edge s_i in S_{rise} , we start routing from (x_l, l_i) in the routing plane, and route towards (x_r, r_i) in the following manner. We advance the routing along the y-axis upward until the routing is blocked, or if we have reached the y-coordinate r_i . Then we shall route one grid unit along the x-axis rightward if possible and repeat the routing along the y-axis. This process is repeated until one of the following three cases is encountered. 1) The connection is completed, 2) the routing has ended on the right column but did not reach (x_r, r_i) , and 3) the routing has failed to reach the right column. For the case 1), we simply add the start-point (x_r, r_i, net_i) to P_r . For case 2), we add the start-point with the new y-coordinate, (x_r, new_r_i, net_i) to P_r , where r_new_i is the y-coordinate of the end-point of the physical routing on the right column. For case 3) we remove any partial routing that we might have added between the column-pair, and add the start-points n_i = (x_l, l_i, net_i) , and $target(n_i)$ to the list of terminals P_{prop} to be propagated to the next layer. Fig. 8 illustrates these cases. The left side of Fig. 8 shows four rising edges in the MWNCM named a, b, c, d. Terminals t1 and t2 are of other nets. Routing for edges a and b are completed. Edge c is routed to the right column but at a different v-coordinate. That is, we have altered the topological solution since the end-point of the routing does not correspond to the end-point of the edge in MWNCM. However, we fail to route edge d because of the blocking terminal on the right column. In this case, both the start-point of the edge d and its target will be added to P_{prop} . For edges in S_{fall} , the procedure is similar except that routing along the y-axis is always downward.

3.3. Pin Redistribution

Another feature of the SLICE router is that it redistributes terminals during the routing process to avoid local congestion. Unlike the pin redistribution algorithm presented in [12], our pin redistribution process is interleaved with the routing process. At the end of planar routing of each layer, a pin redistribution step is performed. Since the planar routing for a net is blocked only when it encounters some obstacles or other routings, the terminals in P_{prop} tend to be clustered. This will make the routing on subsequent layers difficult because the routings

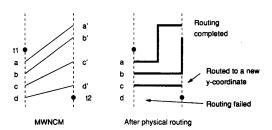


Fig. 8. Example of physical routing.

will be congested around the clustered terminals. To reduce the routing congestion, we want to ensure that the terminals in P_{prop} are evenly distributed. We define the terminal density of a given column to be the number of terminals in that column. Then to reduce the routing congestion, we try to move the terminals in P_{prop} such that the terminal densities are roughly equal among all the columns.⁵ Furthermore, we should try to move the terminals in such a way so that the increase in wirelength is minimized. Our pin redistribution algorithm processes the terminals one at a time moving the terminals horizontally to a column with the lowest terminal density. For a given terminal of net n, the possible columns that it can be moved to is restricted to be in the range $[x(n)]_t$ slack, $x(n)_r + slack$, where $x(n)_r$ and $x(n)_r$ are the x-coordinates of the leftmost and rightmost terminals of net n, and slack is a small constant. Experimental results show that the pin redistribution algorithm consistently improves the utilization of the routing resources.

3.4. Restricted Maze Routing

Our planar algorithm will produce routing segments extending predominantly in the scanning direction. Therefore, many start-points may not be routed because they are lined up almost vertically. To complement the planar routing, we use a restricted maze router to complete as many left over nets as possible.

To conserve memory, we restrict the maze-routing to within two routing layers. Moreover, we restrict the range of the maze router to a thin "vertical slice" of the routing region since we are primarily interested in vertical connections. Typically, the maze range is 10% of the width of the routing region.

3.5. Jog Removal

Since the planar routing algorithm does not penalize the formation of wiring jogs, the completed routings may contain many unnecessary jogs. Therefore, a clean up phase is necessary to remove these jogs to improve the quality of the planar routing solution.

We identify two kinds of jogs. We call simple jogs to be those that can be eliminated by moving a single wire

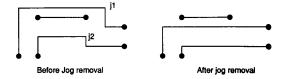


Fig. 9. Removal of simple jogs by moving horizontal segments downward.

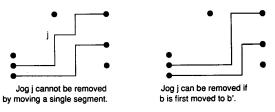


Fig. 10. Example of a complex jog.

TABLE I
CHARACTERISTICS OF EXAMPLES

Example	number of chips	number of nets	number of pins	size of substrate (mm ²)	pitch (um)	grid size
test1	4	500	1000	22.5×22.5	75	300×300
test2	9	957	1914	30×30	75	400×400
test3	9	1254	2508	37.5×37.5	75	500×500
mcc1	6	802	2495	45×45	75	599×599
mcc2	37	7118	14659	152.4 × 152.4	75	2032×2032

segment as shown in Fig. 9. Otherwise, the remaining jogs are called the *complex jogs*, where more than one wire segments need to be moved to eliminate a jog as shown in Fig. 10. SLICE tries to remove the simple jogs first, then it tries to remove the remaining complex jogs. Both algorithms are based on the efficient plane sweeping technique used extensively in computational geometry [17]. Experimental results show that on the average, more than 47% of the jogs can be removed by our algorithms.

IV. EXPERIMENTAL RESULTS

We implemented SLICE on the Sun workstations using the C language. The following experimental results were recorded on a Sun SPARC station II with 32 MB of memory. We tested the program on five examples shown in Table I. The examples, test1, test2, and test3, are generated with random netlists. Examples mcc1 and mcc2 were industrial MCM routing examples provided by MCC. Example mcc2 is a supercomputer with 37 VHSIC gate arrays.

The routing results obtained by SLICE on these examples are given in Table II. The lower bound on wire length for each net is computed by the half perimeter of the bounding rectangle that encloses all the terminals in the net. This is a conservative lower bound for multiterminal

⁵Since the routing direction for the next layer is vertical, the terminals that are on the same column often block each other. Therefore, reduction in terminal density at each column leads to better routing results.

TABLE II CHARACTERISTICS OF SOLUTIONS

Example	number of	number of	number of	wi	run time		
	layers	vias	jogs	lower bound	SLICE	ratio	(hr:min)
test1	5	2013	3453	102238	109092	1.067	0:02
test2	6	5271	9656	265000	286723	1.082	0:06
test3	6	6892	13552	426308	459046	1.077	0:12
mcc1	5	6386	11215	339226	402258	1.186	0:12
mcc2	7	47751	107888	5362181	5691700	1.061	7:25

TABLE III
DISTRIBUTION OF COMPLETED NETS

Example	% of nets complexed in x layers									
	1	2	3	4	5	6	7			
test1	4.8	52.8	80.2	97.6	100.0					
test2	2.5	27.6	55.8	82.7	96.3	100.0				
test3	2.3	30.1	56.9	84.9	97.3	100.0				
mcc1	12.2	53.1	82.8	98.0	100.0					
mcc2	1.6	33.9	60.9	81.2	94.2	99.3	100			

nets. It can be seen that SLICE uses at most 9% more than this lower bound for all examples except mccl.⁶

Table III shows that a large percentage of the nets are completed within the first few routing layers. For all cases, more than 80% of the nets are completed within the first 4 routing layers.

Table IV shows the effect of the two jog removal algorithms. Note that each of the individual jog removal algorithms may remove both kinds of jogs, thus the total jogs removed by applying both algorithms are less than the sum of the jogs removed by applying the two individual algorithms independently.

Table V shows the effect of the pin redistribution algorithm. The algorithm consistently reduces the number of layers needed to complete the routing at the expense of a slight increase (2.6%) in the total wirelength. The impact of the pin redistribution algorithm on the number of vias and jogs is usually small. On average, the number of jogs is increased by 2.5% and the number of vias is increased by 0.3%.

We also compare our results with a general 3-D maze router in Table VI. The 3-D maze router uses a reserved layer model,^{7,8} in which the horizontal wires and vertical

⁶There are many multiterminal nets in mcc1. Since the lower bound given by the half perimeter of the bounding box is not tight for multiterminal nets of size 4, our lower bound for mcc1 is considerably smaller than the optimal wirelength. In fact, it is commonly believed that the wirelength of a minimum Steiner tree for a multiterminal net is at most 88% of the wirelength of a minimum spanning tree on average [18], which leads to a new wirelength lower bound of 362497 for mcc1. The result by the SLICE router is only 11.0% more than the new lower bound.

⁷The 3-D maze router using the reserved layer model performed much better than the one using the unreserved layer model. For example, in our experiment, the number of layers required for example mccl is 17 with the unreserved layer model versus 5 layer with the reserved layer model.

⁸We tried both the reserved layer and the unreserved layer model two-layer maze router in SLICE but the difference in the results is insignificant. For all the test cases, the results by SLICE reported in Tables II–VI are based on the unreserved layer model two-layer maze router.

TABLE IV EFFECTS OF JOG REMOVAL ALGORITHMS

Total number of jogs									
Example	Without jog removal	Algo. to remove Simple Jogs	Algo. to remove Complex Jogs	Both Algorithms					
test1	6732	3785	4015	3453					
test2	18519	10991	11328	9656					
test3	25725	15944	16079	13552					
mcc1	20399	12260	12999	11215					
mcc2	241844	129127	122566	107888					

wires are routed in different layers. The 3-D maze router was not able to run on mcc2 on our system due to the large size of the example. On the average, SLICE is more than six times faster than the 3-D maze router, and uses 29% fewer vias than the 3-D maze router. However, the number of layers used by SLICE is generally more than the 3-D maze router. But as shown in Table III, the last few layers in the SLICE solutions are very sparse.

Another advantage that SLICE has over the 3-D maze router is its low memory requirement. For the example mcc2 (a supercomputer with 37 gate arrays), in order to store the entire grid of size $7 \times 2032 \times 2032$, the 3-D maze router needs 110 MB of memory (assuming that we use four bytes for each grid point to store the net number, routing cost, etc.) That is why the 3-D maze router failed to route the example on our system. However, using a maze routing range of 10%, at any time, the working space of SLICE is only $2 \times 10\% \times 2032 \times 2032 = 3.3$ MB of memory. So SLICE successfully produced a satisfactory solution. Furthermore, if the routing pitch for the same example is reduced by a factor of two, the 3-D maze router will require 441 MB of memory whereas SLICE will require only 13.2 MB of working memory. Clearly, for the next generation of dense MCM design, the 3-D maze router will face more severe memory limitation, and the advantage of SLICE will become much more significant.

V. CONCLUSIONS AND FUTURE EXTENSIONS

In this paper, we presented a fast multilayer general area router named SLICE for MCM Designs. The routing result of the SLICE router is independent of net ordering and uses fewer vias. The total wirelength produced by SLICE is only a few percents away from the optimal. Compared with a general 3-D maze router, with a small increase in the number of routing layers, SLICE runs more than six times faster, uses 29% fewer vias, and requires far less memory.

9656

13552

11215

107888

9230

13326

11183

104665

	EFFECTS OF PIN REDISTRIBUTION ALGORITHM									
no. of layers		no.	of vias	no. c	of jogs	total wirelength				
1	without	with	without	with	without	with	witho			
	6	2013	2025	3453	3366	109092	1072			

5268

6821

6120

49475

5271

6892

6386

47751

TABLE V

TABLE VI	
COMPARISON WITH MAZE	ROUTER

	number of layers		number of vias		number of jogs		total wire length		run time (hr:min)	
Example	SLICE	maze	SLICE	maze	SLICE	maze	SLICE	maze	SLICE	maze
test1	5	4	2013	2975	3453	421	109092	107908	0:02	0:08
test2	6	4	5271	7127	9656	892	286723	273642	0:06	0:48
test3	6	4	6892	9347	13552	1094	459046	441552	0:12	1:40
mcc1	5	5	6386	8794	11215	1244	402258	397221	0:12	0:59
mcc2	7	_	47751	_	107888	_	5691700	_	7:25	

The SLICE router can also handle 45° routing. After we obtain a maximum weighted noncrossing matching, we can use a more sophisticated procedure to map the topological routing solution into a physical routing solution which allows 45° routing. The SLICE router can handle arbitrary obstacle in the routing region as well since it can avoid generating edges whose end-points are on the obstacles.

n

with

5

6

Example

test1

test2

test3

mcc1

mcc2

Although the SLICE router reduces the total number of vias significantly, it might be possible that some individual nets have high via counts. We are in the process of developing a MCM router which can bound the number of vias used for every net in order to achieve predictable performance. We also hope to take some performance issues (such as coupling and reflection) into consideration in our design of the next generation of MCM

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286723

459046

402258

5691700

without

107247

279542

445208

388078

5588512

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