



Supplementary materials for

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Algorithm S1 Simulated annealing algorithm based on the logical layout

Input: Initial temperature T_0 , termination temperature T_f , temperature decline rate r , number of iterations at each temperature l , initial row sequence state P_0 , component set U .

Output: Logical layout result L .

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1:  $T = T_0, P = P_0$                                 ▷ Initialize parameters.
2:  $Count_{NC} = \text{countCrossings}(P_0)$                 ▷ Initialize the optimal number of wire crossings.
3:  $P_{\text{best}} = P$                                     ▷ Initialize the optimal state.
4: while  $T > T_f$  do
5:   for  $i = 1$  to  $l$  do
6:      $P_{\text{new}} = \text{randomSwapRow}(U)$ 
7:      $\text{NewCount}_{NC} = \text{countCrossings}(P_{\text{new}})$         ▷ Calculate the number of crossings in new state.
8:      $\Delta = \text{NewCount}_{NC} - \text{Count}_{NC}$ 
9:     if  $\Delta < 0$  or  $\exp(-\frac{\Delta}{T})$  then
10:       $P = P_{\text{new}}$ 
11:       $P_{\text{best}} = P$ 
12:       $\text{Count}_{NC} = \text{NewCount}_{NC}$ 
13:     end if
14:   end for
15:    $T = T \times r$ 
16: end while
17:  $L = \text{layout}(P)$                                 ▷ Form a layout based on State.
18: return  $L$ 

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Algorithm S2 Layout algorithm based on reinforcement learning

Input: Maximum training episodes Ep , learning rate l_r , netlist H .

Output: Logical layout result L .

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1: Initialize neural network Agent.
2: Initialize the layout environment Env.
3: Define the reward function  $R$ .
4: repeat
5:   Reset(Env) ▷ Reset layout environment.
6:    $S = \text{getInitialState}(\text{Env})$  ▷ Get initial state.
7:   while not done do
8:      $S = \text{embedFeatures}(S, \text{Env})$  ▷ Embedded building block features.
9:      $A = \text{selectAction}(\text{Agent})$ 
10:    done,  $S, r = \text{envExecute}(A)$  ▷ Execute layout action, update state.
11:    if done then
12:      Update Agent.
13:    end if
14:  end while
15: until  $e > Ep$ 
16:  $L = \text{layout}(S)$  ▷ Form a layout based on State.
17: return  $L$ 

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Algorithm S3 A* algorithm

Input: Route starting point N_0 , route ending point N_f .

Output: The routing path p from N_0 to N_f .

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1: Initialize Routing Grid Map  $G$ .
2: Initialize the set of nodes to be traversed  $O$  and the set of nodes already traversed  $C$ .
3:  $\text{aCost}[N_0] = 0$  ▷ Initialize the actual cost from  $N_0$  to the current point  $N_c$ .
4:  $\text{eCost}[N_0] = \text{getManhattanCost}(N_0) \times 0.8$  ▷ Initialize the estimated cost from  $N_c$  to  $N_f$ .
5:  $\text{Parent}[N_0] = \text{null}$ 
6:  $O.\text{add}(N_0, \text{aCost}[N_0] + \text{eCost}[N_0])$ 
7: while  $O \neq \emptyset$  do
8:    $N_c = \text{getMinCostNode}(O)$ 
9:   if  $N_c == N_f$  then
10:    return reconstructPath( $N_c$ )
11:   end if
12:   for  $N_i$  in getNeighbors( $N_c, G$ ) do
13:     if  $N_i \in C$  then
14:       continue
15:     end if
16:      $\text{tCost}[N_i] = \text{aCost}[N_c] + \text{getaCost}(N_c, N_i)$ 
17:     if  $N_i \notin O$  or  $\text{tCost}[N_i] < \text{aCost}[N_i]$  then
18:        $\text{aCost}[N_i] = \text{tCost}[N_i]$ 
19:        $\text{eCost}[N_i] = \text{getManhattanCost}(N_i) \times 0.8$ 
20:        $\text{Parent}[N_i] = N_c$ 
21:        $O.\text{add}(N_i, \text{aCost}[N_i] + \text{eCost}[N_i])$ 
22:     end if
23:   end for
24: end while
25: return null ▷ Return null if no path is found at the end of the loop.

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