An Efficient Branch-and-Bound Routing Algorithm for Optical NoCs

Yihao Liu, Yaoyao Ye

Shanghai Jiao Tong University, China yeyaoyao@sjtu.edu.cn

Outline

- Research background
- Branch-and-bound routing for ONoCs
 - Branch-and-bound routing pruned at turning nodes
 - Variant-1: Branch-and-bound thermal-aware routing
 - Variant-2: Branch-and-bound bi-objective routing
- Simulation results
- Conclusion and future works

Optical NoC Architecture

Advantages of ONoCs

- High bandwidth
- High energy efficiency

Challenges of ONoCs

- Temperature sensitivity
- Optical link contention



Mesh based photo-electric hybrid NoC [1]



Wavelength division multiplexing [2]

[1] M. Li, W. Liu, L. H. Duong, P. Chen, L. Yang, and C. Xiao, "Contention-aware routing for thermalreliable optical networks-on-chip," IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems, vol. 40, no. 2, pp. 260–273, 2020.

[2] A. Narayan, Y. Thonnart, P. Vivet and A. K. Coskun, "PROWAVES: Proactive Runtime Wavelength Selection for Energy-Efficient Photonic NoCs," IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems, vol. 40, no. 10, pp. 2156-2169, Oct. 2021

Challenges of ONoCs

Temperature sensitivity

- Silicon microresonator (MR) as optical switching element
- Chip temperature fluctuates in time and space
- Resonant wavelength of MR does not match laser wavelength



Challenges of ONoCs

Thermal-aware adaptive routing [1, 2]

- Circumvent high-temperature regions for routing
- Congestion exacerbates within the low-temperature regions on the chip



[1] Z. Zhang and Y. Ye, "A learning-based thermal-sensitive power optimization approach for optical NoCs," ACM Journal on Emerging Technologies in Computing Systems (JETC), 2018, 14(2): 21

[2] W. Zhang and Y. Ye, "A table-free approximate Q-learning-based thermal-aware adaptive routing for optical NoCs," IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems, vol. 40, no. 1, pp. 199–203, 2021.

Challenges of ONoCs

Heuristic contention-aware centralized routing [1]

- The scope of available paths is too small
- The optical power loss optimization is less efficient
- Centralized routing matches circuit switching mechanisms of ONoCs



[1] M. Li, W. Liu, L. H. Duong, P. Chen, L. Yang, and C. Xiao, "Contention-aware routing for thermalreliable optical networks-on-chip," IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems, vol. 40, no. 2, pp. 260–273, 2020.

Characteristics of optical power loss

- Thermal-induced optical power loss primarily originates from active switches
- Paths meeting thermal reliability contain fewer turning nodes



Branch-and-Bound (B&B)

Using a tree search strategy to implicitly enumerate all possible solutions

S

Components of B&B

- Branching strategy
- Search strategy
- Pruning rules



- Turning node based branching strategy
- Turning node based bounding function





Child Node List

Algorithm 1: Branch-and-Bound Routing (BBR) **input** : global information, source S, destination D. initial upper bound V_{init} , initial optimal path Pinit **output:** the optimal path P_{best} 1 $\overline{PT} \leftarrow \{S\}, [Lower, Upper] \leftarrow [0, V_{init}],$ $\frac{P_{best} \leftarrow P_{init} / / \text{ initialization}}{2 \text{ while } PT \neq \emptyset \text{ do}}$ $S_i \leftarrow \text{pop node } i \text{ with minimum value from } PT$ 3 for $x \in child node list of S_i$ do 4 $V_x \leftarrow \text{GetBoundFuncValue}(V_i, x)$ 5 $Flag_{x} \leftarrow CheckPrunFlag()$ 6 if $Flag_x$ then 7 if x is branch node then 8 add node x to PT9 else if x is leaf node then 10 $P_{best} \leftarrow P_i + \{x, D\}, Upper \leftarrow V_x$ 11 delete nodes in PT where V > Upper12

	Pending node	Best path	Pending table
Iteration ₁		P _{init}	S
Iteration ₂			



Child Node List

Algorithm 1: Branch-and-Bound Routing (BBR) **input** : global information, source S, destination D. initial upper bound V_{init} , initial optimal path Pinit **output:** the optimal path P_{hest} 1 $PT \leftarrow \{S\}, [Lower, Upper] \leftarrow [0, V_{init}],$ $\frac{P_{best} \leftarrow P_{init} / / \text{ initialization}}{2 \text{ while } PT \neq \emptyset \text{ do}}$ $S_i \leftarrow \text{pop node } i \text{ with minimum value from } PT$ 3 for $x \in child$ node list of S_i do 4 $V_x \leftarrow \text{GetBoundFuncValue}(V_i, x)$ 5 $Flag_{x} \leftarrow CheckPrunFlag()$ 6 if $Flag_x$ then 7 if x is branch node then 8 add node x to PT9 else if x is leaf node then 10 $P_{best} \leftarrow P_i + \{x, D\}, Upper \leftarrow V_x$ 11 delete nodes in PT where V > Upper12

	Pending node	Best path	Pending table
Iteration ₁	S	P _{init}	
Iteration ₂			



Child Node List

Algorithm 1: Branch-and-Bound Routing (BBR)	
input : global information, source S, destination D,	_
initial upper bound V_{init} , initial optimal path	
P_{init}	
output: the optimal path P_{best}	
1 $PT \leftarrow \{S\}, [Lower, Upper] \leftarrow [0, V_{init}],$	
$P_{best} \leftarrow P_{init} / /$ initialization	
2 while $PT \neq \emptyset$ do	
$S_i \leftarrow pop \text{ node } i \text{ with minimum value from } PT$	
4 for $x \in child node list of S_i$ do	
5 $V_x \leftarrow \text{GetBoundFuncValue}(V_i, x)$	
6 $Flag_r \leftarrow CheckPrunFlag()$	
7 if $Flaa_x$ then	
8 if x is branch node then	
9 add node x to PT	
10 else if x is leaf node then	-
11 $P_{best} \leftarrow P_i + \{x, D\}, Upper \leftarrow V_x$	
12 delete nodes in PT where $V > Upper$	
-+++	

	Pending node	Best path	Pending table
Iteration ₁	S	P ₁	
Iteration ₂			



Algorithm 1: Branch-and-Bound Routing (BBR) **input** : global information, source S, destination D. initial upper bound V_{init} , initial optimal path Pinit **output:** the optimal path P_{hest} 1 $PT \leftarrow \{S\}, [Lower, Upper] \leftarrow [0, V_{init}],$ $P_{best} \leftarrow P_{init}$ // initialization 2 while $PT \neq \emptyset$ do $S_i \leftarrow \text{pop node } i \text{ with minimum value from } PT$ 3 for $x \in child$ node list of S_i do 4 $V_x \leftarrow \text{GetBoundFuncValue}(V_i, x)$ 5 $Flag_{x} \leftarrow CheckPrunFlag()$ 6 if $Flag_x$ then 7 if x is branch node then 8 add node x to PT9 else if x is leaf node then 10 $P_{best} \leftarrow P_i + \{x, D\}, Upper \leftarrow V_x$ 11 delete nodes in PT where V > Upper12

	Pending node	Best path	Pending table
Iteration ₁	S	P ₁	S ₁ , S ₂ , S ₃
Iteration ₂			



Algorithm 1: Branch-and-Bound Routing (BBR) **input** : global information, source S, destination D. initial upper bound V_{init} , initial optimal path Pinit **output:** the optimal path P_{hest} 1 $PT \leftarrow \{S\}, [Lower, Upper] \leftarrow [0, V_{init}],$ $\frac{P_{best} \leftarrow P_{init} / / \text{ initialization}}{2 \text{ while } PT \neq \emptyset \text{ do}}$ $S_i \leftarrow \text{pop node } i \text{ with minimum value from } PT$ 3 for $x \in child$ node list of S_i do 4 $V_x \leftarrow \text{GetBoundFuncValue}(V_i, x)$ 5 $Flag_{x} \leftarrow CheckPrunFlag()$ 6 if $Flag_x$ then 7 if x is branch node then 8 add node x to PT9 else if x is leaf node then 10 $P_{best} \leftarrow P_i + \{x, D\}, Upper \leftarrow V_x$ 11 delete nodes in PT where V > Upper12

	Pending node	Best path	Pending table	
Iteration ₁	S	P ₁	S ₁ , S ₂ , S ₃	
Iteration ₂	S ₂	P ₁	S ₁ , S ₃	



Algorithm 1: Branch-and-Bound Routing (BBR) **input** : global information, source S, destination D. initial upper bound V_{init} , initial optimal path Pinit **output:** the optimal path P_{hest} 1 $PT \leftarrow \{S\}, [Lower, Upper] \leftarrow [0, V_{init}],$ $P_{best} \leftarrow P_{init}$ // initialization 2 while $PT \neq \emptyset$ do $S_i \leftarrow \text{pop node } i \text{ with minimum value from } PT$ for $x \in$ child node list of S_i do $V_x \leftarrow \text{GetBoundFuncValue}(V_i, x)$ 5 $Flag_{x} \leftarrow CheckPrunFlag()$ 6 if $Flag_x$ then 7 if x is branch node then 8 add node x to PT9 else if x is leaf node then 10 $P_{best} \leftarrow P_i + \{x, D\}, Upper \leftarrow V_x$ 11 delete nodes in PT where V > Upper12

	Pending node	Best path	Pending table	
Iteration ₁	S	P ₁	S ₁ , S ₂ , S ₃	
Iteration ₂	S ₂	P ₅		



input : global information, source S, destination D. initial upper bound V_{init} , initial optimal path Pinit **output:** the optimal path P_{hest} 1 $PT \leftarrow \{S\}, [Lower, Upper] \leftarrow [0, V_{init}],$ $P_{best} \leftarrow P_{init}$ // initialization 2 while $PT \neq \emptyset$ do $S_i \leftarrow \text{pop node } i \text{ with minimum value from } PT$ 3 for $x \in child$ node list of S_i do 4 $V_x \leftarrow \text{GetBoundFuncValue}(V_i, x)$ 5 $Flag_{x} \leftarrow CheckPrunFlag()$ 6 if $Flag_x$ then 7 if x is branch node then 8 add node x to PT9 else if x is leaf node then 10 $P_{best} \leftarrow P_i + \{x, D\}, Upper \leftarrow V_x$ 11 delete nodes in PT where V > Upper12

Algorithm 1: Branch-and-Bound Routing (BBR)

	Pending node		Pending table	
Iteration ₁	S	P ₁	S ₁ , S ₂ , S ₃	
Iteration ₂	S ₂	P ₅		

Variant-1: Branch-and-bound thermalaware routing (BBTR)

Additional optical power loss of paths

- L_{active_i} is optical power loss of turning node S_i
- $\Delta L = \sum_{i=1}^{k} \left(L_{active_i} + \Delta h_i \cdot L_{wire} \right)$

Bounding function

• $\Delta L_x = \Delta L_{actual_i} + \Delta L_{actual_{i \to x}} + \Delta L_{min_{x \to D}}$

Pruning rules

• $\Delta L_x < Upper_{\Delta L}$



Variant-1: Branch-and-bound thermalaware routing (BBTR)

Additional optical power loss of paths

- L_{active_i} is optical power loss of turning node S_i
- $\Delta L = \sum_{i=1}^{k} (L_{active_i} + \Delta h_i \cdot L_{wire})$

Bounding function

• $\Delta L_x = \Delta L_{actual_i} + \Delta L_{actual_{i \to x}} + \Delta L_{min_{x \to D}}$

Pruning rules

• $\Delta L_x < Upper_{\Delta L}$



Link occupancy model

- $t_{max} = max(t_i), i \in P_{link}$
- $\Delta t = \sum_{i=1}^{h} (t_{max} t_i) + \alpha \cdot \Delta h \cdot t_{payload}$
- Bi-objective model
 - $BIO = \Delta L + \beta \cdot t_{max} + \gamma \cdot \Delta t$

Cost of routing paths

- $\Delta L_3 < \Delta L_2 < \Delta L_1 < \Delta L_4$
- $\bullet BIO_2 < BIO_3 < BIO_1 < BIO_4$

	k	Δh	h	t _{max}	Δt
P ₁	1	0	3	4	5
P ₂	2	0	3	2	1
P ₃	2	2	5	5	8
P ₄	3	6	9	5	31
Assume $t_{payload} = 5$ cycles, $\alpha = 0.5$					



- Two-stage pipeline process
- Bounding function
 - $BIO_x = BIO_{actual_i} + BIO_{actual_{i \to x}} + BIO_{min_{x \to D}}$

Pruning rules

- $BIO_x < Upper_{BIO}$
- $\Delta L_x < \Delta L_{min} + max_{\Delta L}$
- $\Delta t_x < h_s \cdot max_{\Delta t}$

Adjustable parameters

- Weight factor: α , β , γ
- Pruning rules: $max_{\Delta L}$, $max_{\Delta t}$

Algorithm 2: Branch-and-Bound Bi-Objective Routing Optimization (3BOR)

input : global information TT and LT, routing request buffer RB and RB'
output: the bi-objective optimal path buffer PB

1 for $Req' \in RB'$ do 2 $P_{Bio} \leftarrow Func 3BOR(TT, LT, Req')$

3 $PB.add(P_{Bio})$ 4 for $Req \in RB$ do

$$\begin{array}{c|c} \mathbf{5} & P_{\Delta L} \leftarrow \texttt{FuncBBTR}(TT, Req) \\ \mathbf{6} & Req' \leftarrow (Req, P_{\Delta L}) \end{array}$$

$$7 \mid RB'.add(Req')$$

- Two-stage pipeline process
- Bounding function
 - $BIO_x = BIO_{actual_i} + BIO_{actual_{i \to x}} + BIO_{min_{x \to D}}$

Pruning rules

- $BIO_x < Upper_{BIO}$
- $\Delta L_x < \Delta L_{min} + max_{\Delta L}$
- $\Delta t_x < h_s \cdot max_{\Delta t}$

Adjustable parameters

- Weight factor: α , β , γ
- Pruning rules: $max_{\Delta L}$, $max_{\Delta t}$

Algorithm 2: Branch-and-Bound Bi-Objective Routing Optimization (3BOR) **input** : global information TT and LT, routing request buffer RB and RB'**output:** the bi-objective optimal path buffer PB1 for $Req' \in RB'$ do $P_{Bio} \leftarrow \text{Func3BOR}(TT, LT, Req')$ 2 $PB.add(P_{Bio})$ 3 4 for $Reg \in RB$ do $P_{\Delta L} \leftarrow \texttt{FuncBBTR}(TT, Req)$ 5 $Req' \leftarrow (Req, P_{\Delta L})$ 6 RB'.add(Req')7

- Two-stage pipeline process
- Bounding function
 - $BIO_x = BIO_{actual_i} + BIO_{actual_{i \to x}} + BIO_{min_{x \to D}}$

Pruning rules

- $BIO_x < Upper_{BIO}$
- $\Delta L_x < \Delta L_{min} + max_{\Delta L}$
- $\Delta t_x < h_s \cdot \max_{\Delta t}$

Adjustable parameters

- Weight factor: α , β , γ
- Pruning rules: $max_{\Delta L}$, $max_{\Delta t}$

Algorithm 2: Branch-and-Bound Bi-Objective Routing Optimization (3BOR)

input : global information TT and LT, routing request buffer RB and RB'

output: the bi-objective optimal path buffer PB

1 for
$$Req' \in RB'$$
 do

2 |
$$P_{Bio} \leftarrow \text{Func3BOR}(TT, LT, Req')$$

 $\mathbf{3} \quad PB.add(P_{Bio})$

4 for $Req \in RB$ do

5
$$P_{\Delta L} \leftarrow \texttt{FuncBBTR}(TT, Req)$$

6
$$Req' \leftarrow (Req, P_{\Delta L})$$

7 $\ \ RB'.add(Req')$

Outline

- Research background
- Branch and bound routing for ONoCs
 - Branch-and-bound routing pruned at turning nodes
 - Branch-and-bound thermal-aware routing
 - Branch-and-bound bi-objective routing
- Simulation results
- Conclusion and future works

- A cycle-accurate simulation platform based on SystemC
- Comparison baseline
 - Contention-aware routing (CAR) [1]
- Two variants of 3BOR through adjusting parameters
 - 3BOR-LOSS
 - 3BOR-BIO

Temperature distribution

Center-Block



[1] M. Li, W. Liu, L. H. Duong, P. Chen, L. Yang, and C. Xiao, "Contention-aware routing for thermalreliable optical networks-on-chip," IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems, vol. 40, no. 2, pp. 260–273, 2020.

Average optical loss

BBTR < 3BOR-LOSS < 3BOR-BIO < CAR



Average optical loss with varying injection rate

Average packet latency

3BOR-BIO < CAR < 3BOR-LOSS < BBTR</p>



Average packet latency with varying injection rate

The proposed BBTR and 3BOR demonstrate universality in optimizing average optical loss



Average optical loss with varying 3-dB bandwidth

The proposed BBTR and 3BOR exhibits scalability



Comparisons for algorithm running time on various ONoC sizes

Conclusions and Future works

Conclusion

- The proposed BBTR has the best optimization effect on average optical power loss
- The proposed 3BOR can comprehensively optimize communication performance and energy efficiency
- By adjusting parameters, 3BOR can balance the two optimization objectives, adapting to different network traffic patterns and chip temperature distributions.

Future works

- 3BOR with automatic parameter adjustment capability
- New variants of BBR for optimizing different goals