

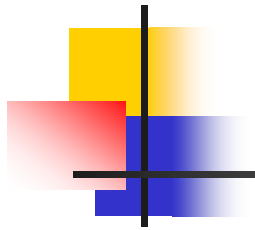


The Era of Many-Module SoC: Revisiting the NoC Mapping Problem

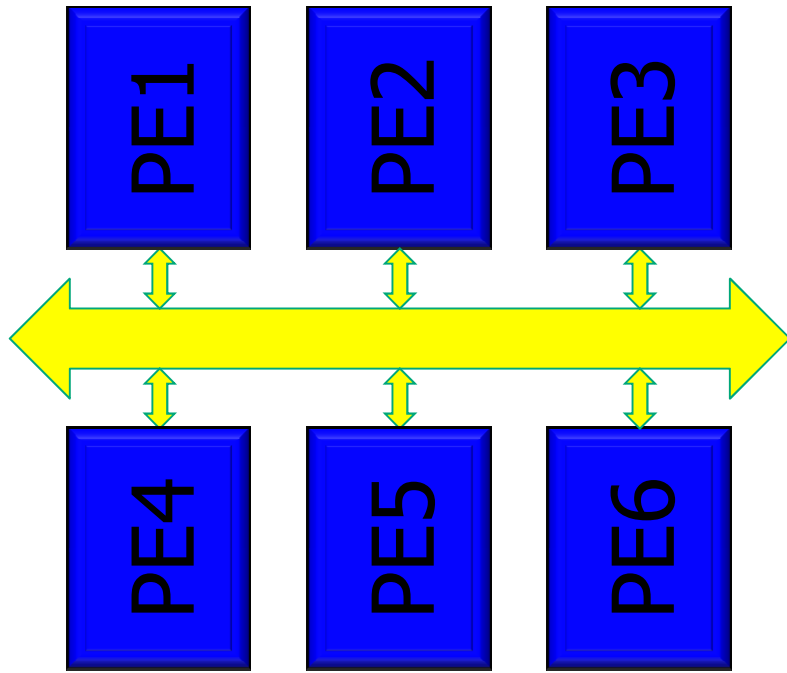
Isask'har (Zigi) Walter, Israel Cidon, Avinoam Kolodny, Daniel Sigalov

Technion – Israel Institute of Technology

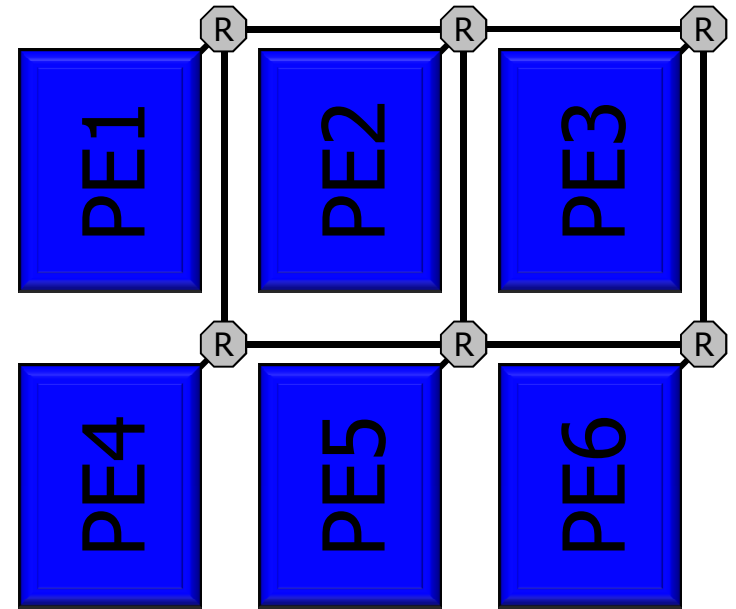
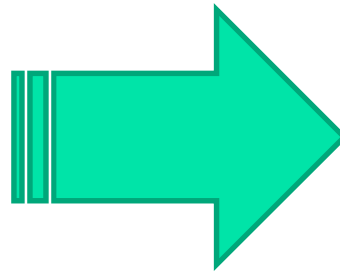
December, 2009



SoC Revolution

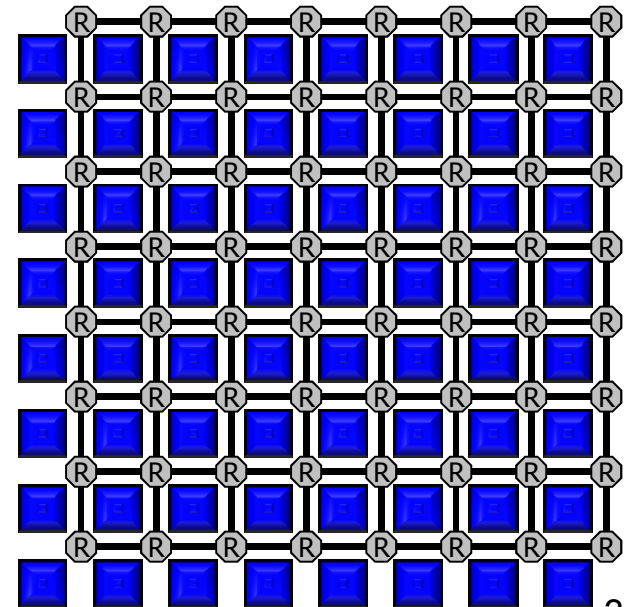
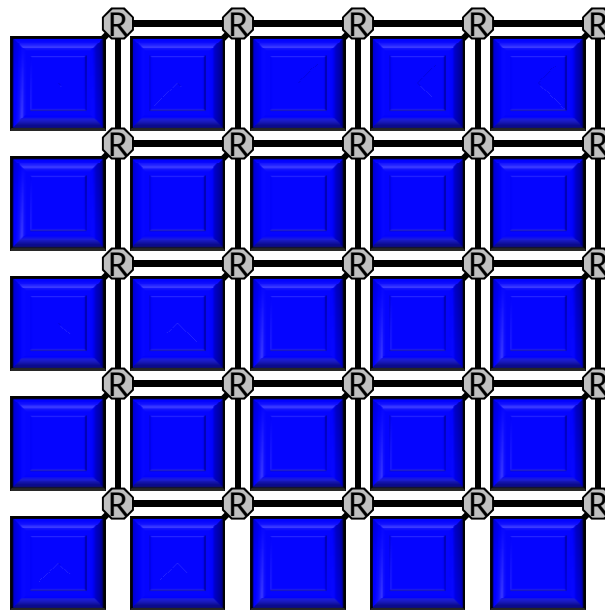
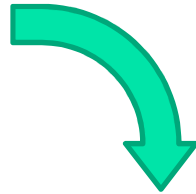
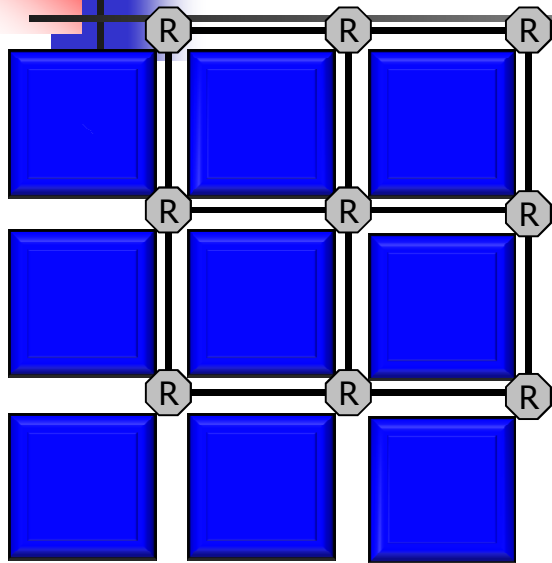


Bus-based system



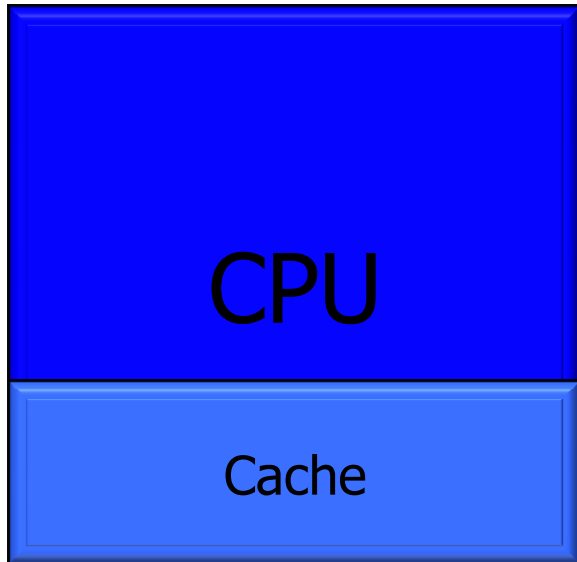
NoC-based system

SoC Evolution

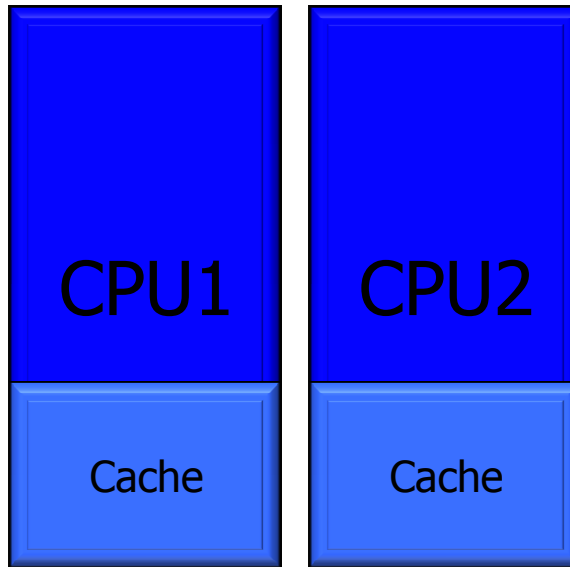


Processor Evolution

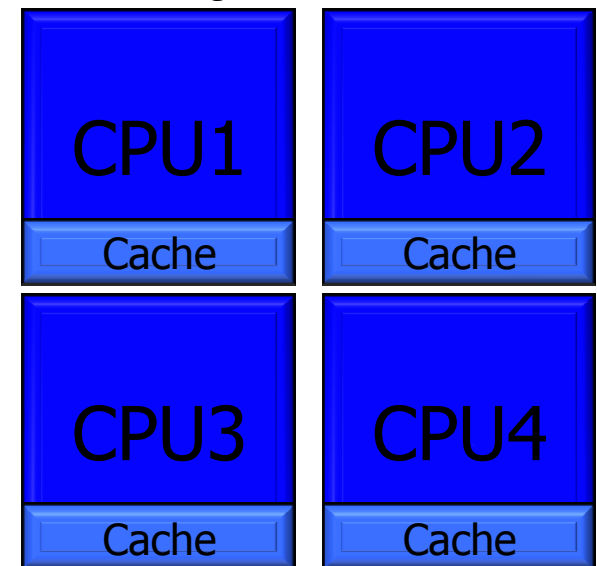
Single Core



Dual Core

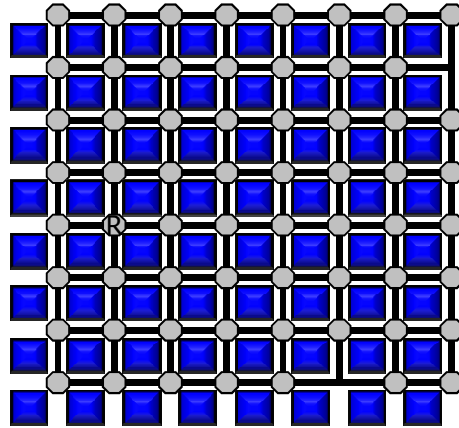


Quad Core

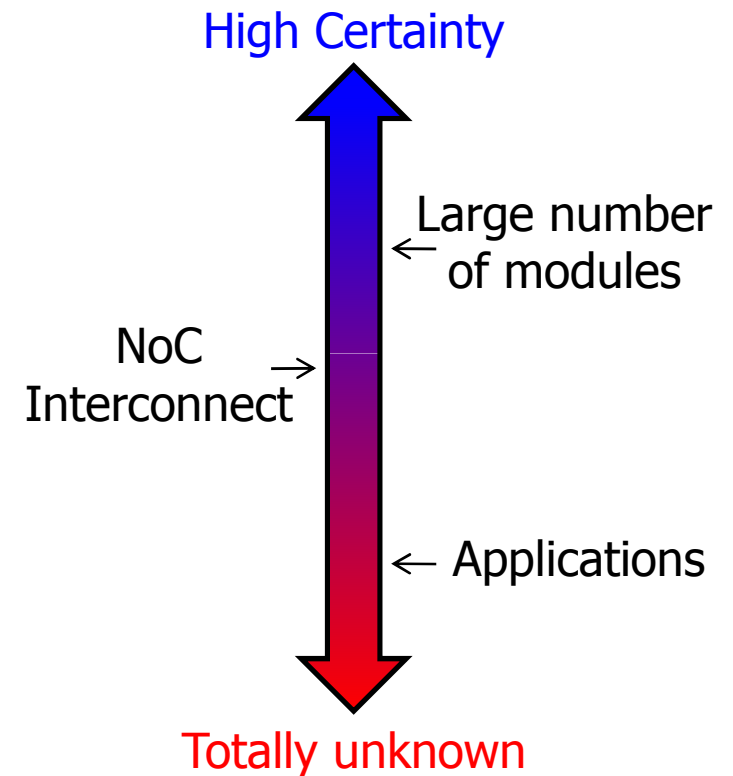


The Era of Many-Module SoC

- How would such chips be like?

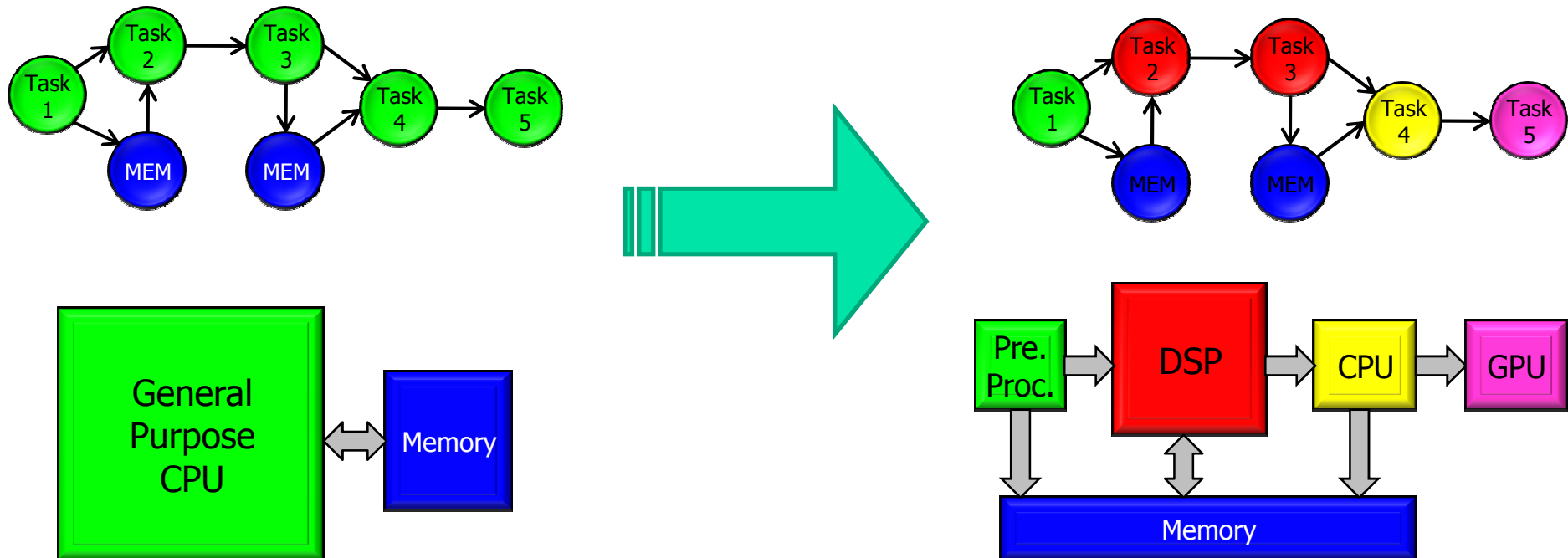


- Most likely
 - Power still important
 - Highly parallel
 - IP reuse
 - Ease of design and verification



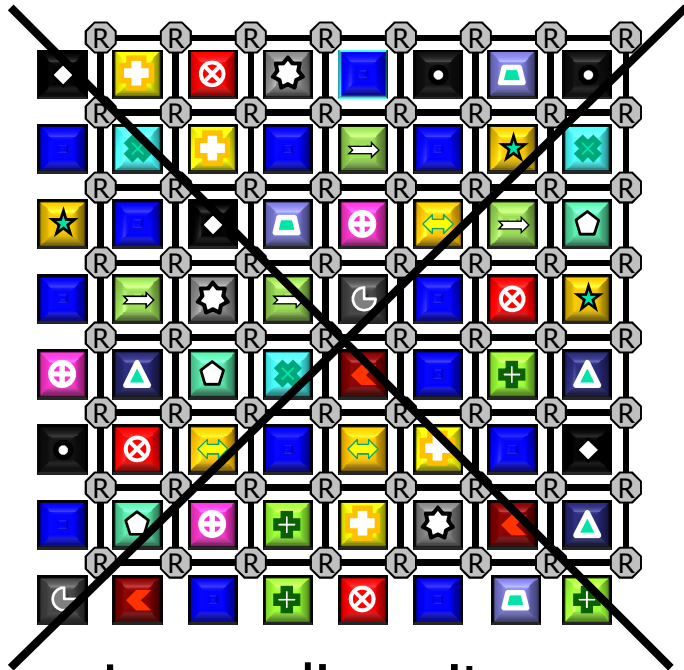
Future SoCs - Observation#1

- Special purpose cores replace general purpose processors
 - Power considerations

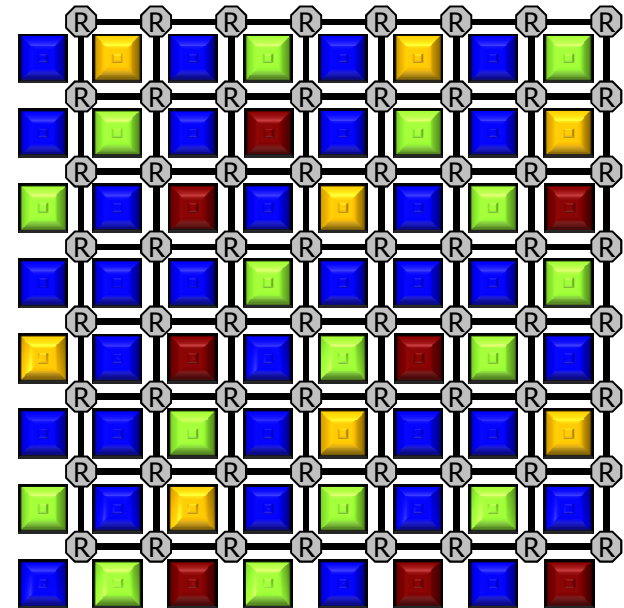
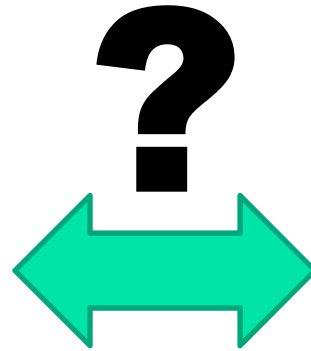


- Processing pipes are getting longer

Future SoCs - Observation#2



- Large diversity
- All modules are unique



- Highly regular
- Classes of Replicated cores
 - standard modules (DSP, HW accelerators, Cache banks, etc.)



The Era of Many-Module SoC

- Increased use of specialized cores
 - Pipes are getting longer
- Replication of processing elements
- How is the design flow affected?
 - This work – mapping of the NoC

Observation#1

Observation#2



Outline

- The Era of Many Module SoC
- **Revisiting the Mapping Problem**
- Cross-Entropy Optimization
- Evaluation

NoC Mapping

- Given
 - Traffic pattern(s)
 - a set (or sets) of pair-wise bandwidth requirements and timing constraints
 - Routing
 - Topology
- Goal
 - Find efficient mapping of cores to tiles





Mapping Optimization

- An **important** design step
 - Mapping affects **power** and **performance!**
- A **difficult** problem!
 - Often heuristic algorithms are used
- Common optimization goals
 - Minimize (dynamic) power
 - Minimize power + maximize performance
 - **Minimize power subject to *performance constraints***



Modeling

- Typical modeling
 - Power and latency proportional to distance
 - Cost function:

$$Cost(\pi \in P) = \sum_{l \in L} BW_l = \sum_{1 \leq i, j \leq N} [b_{i \rightarrow j} \cdot Dist(i, j)]$$

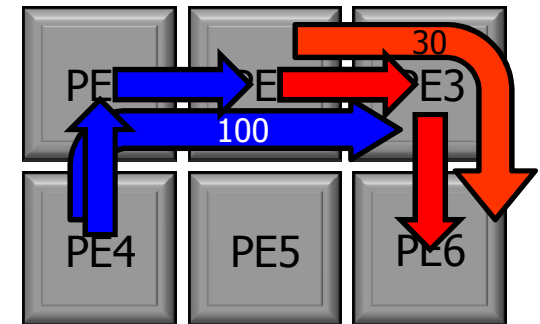
Calculating Mapping Cost

$$Cost(\pi \in P) = \sum_{l \in L} BW_l = \sum_{1 \leq i, j \leq N} [b_{i \rightarrow j} \cdot Dist(i, j)]$$

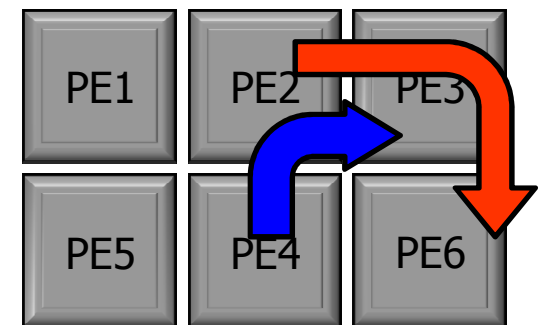
$$Cost(\pi_1) = [bw(PE_2 \rightarrow PE_6) \cdot Dist(PE_2 \rightarrow PE_6)] + [bw(PE_4 \rightarrow PE_3) \cdot bw(PE_4 \rightarrow PE_3)]$$

$$Cost(\pi_1) = 30 \cdot Dist(PE_2 \rightarrow PE_6) + 100 \cdot Dist(PE_4 \rightarrow PE_3)$$

$$Cost(\pi_1) = 30 \cdot 2 + 100 \cdot 3 = 360$$



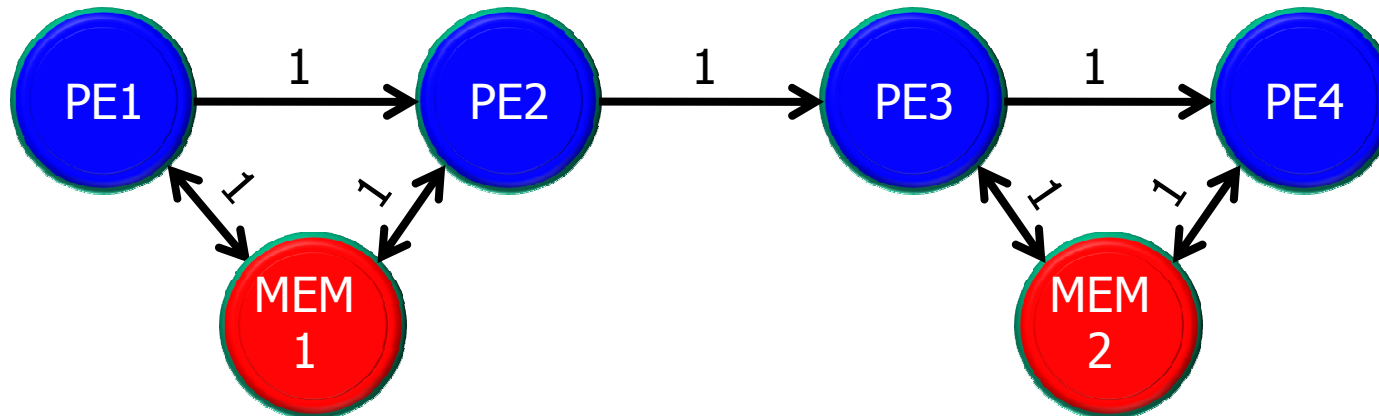
Mapping π_1



Mapping π_2

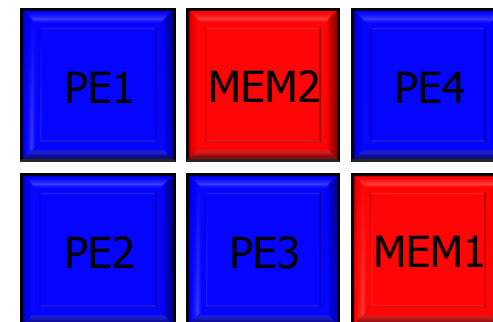
$$Cost(\pi_2) = 30 \cdot 2 + 100 \cdot 2 = 260$$

Motivation - Example #1



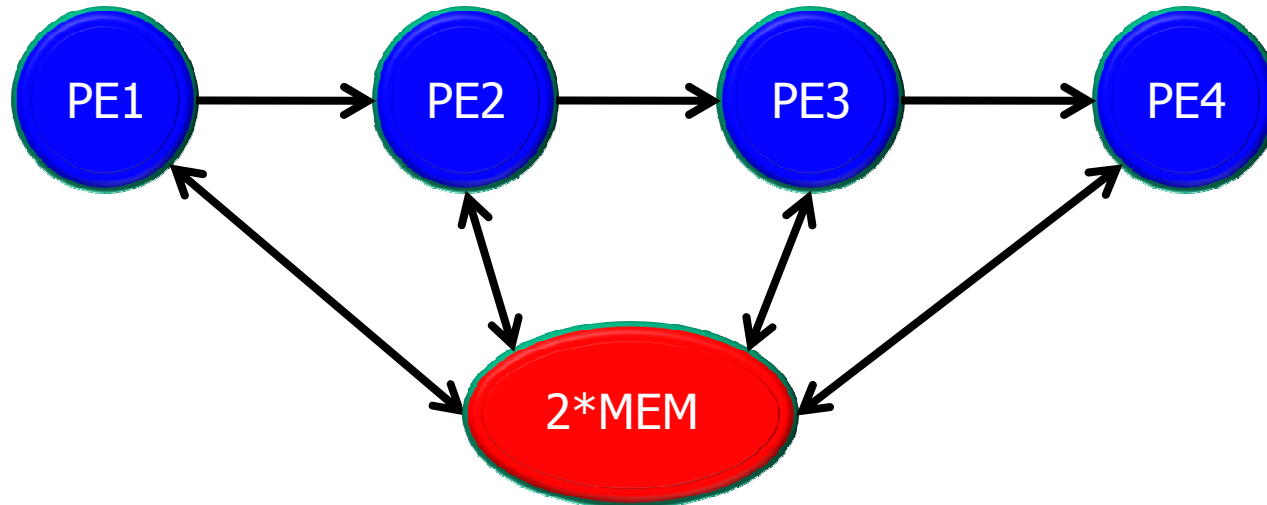
- Optimal mapping (π_1):

$$Cost(\pi_1) = \sum_{1 \leq i, j \leq N} [b_{i \rightarrow j} \cdot Dist(i, j)] = 9$$

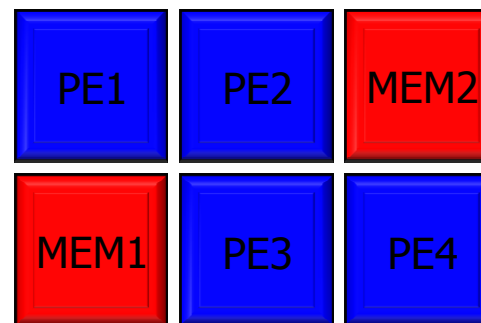


Motivation - Example #1 (cont.)

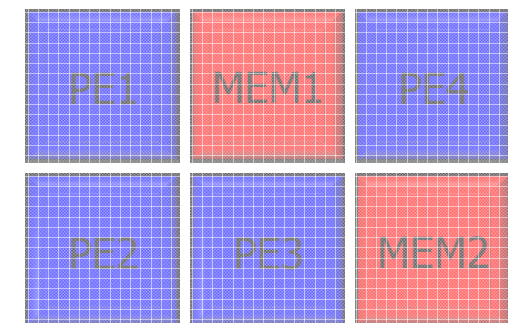
- Let the mapping algorithm assign the flows!



- Optimal mapping (π_2):

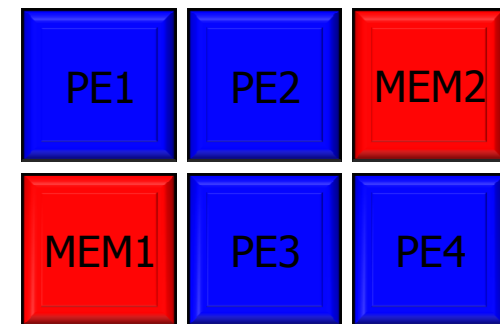
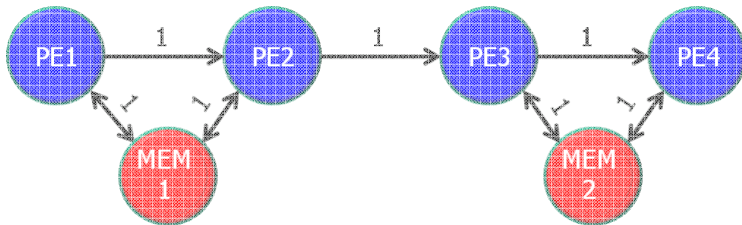
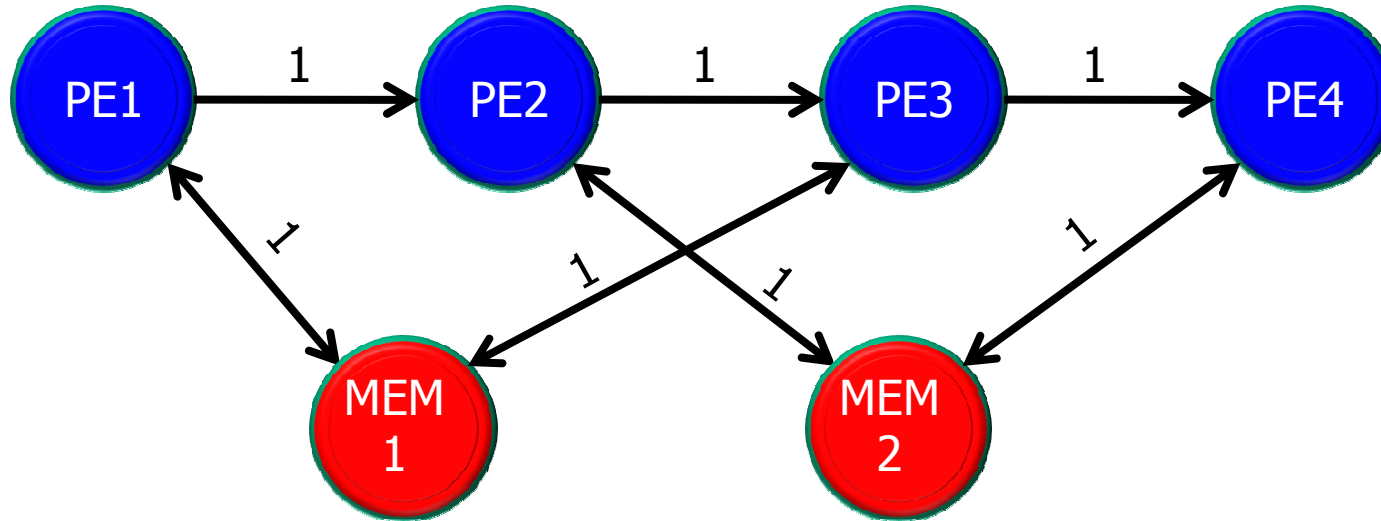


Cost(π_2)=7



Cost(π_1)=9

Motivation - Example #1 (cont.)

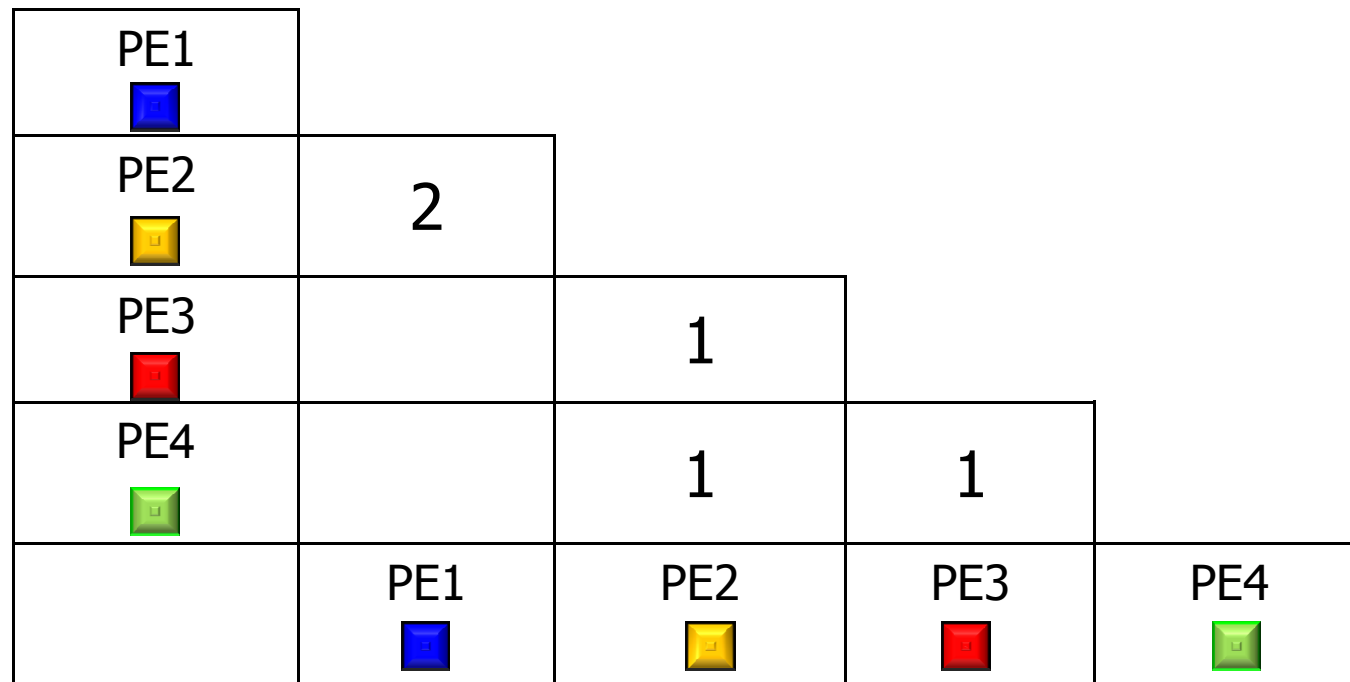


$\text{Cost}(\pi_2)=7$

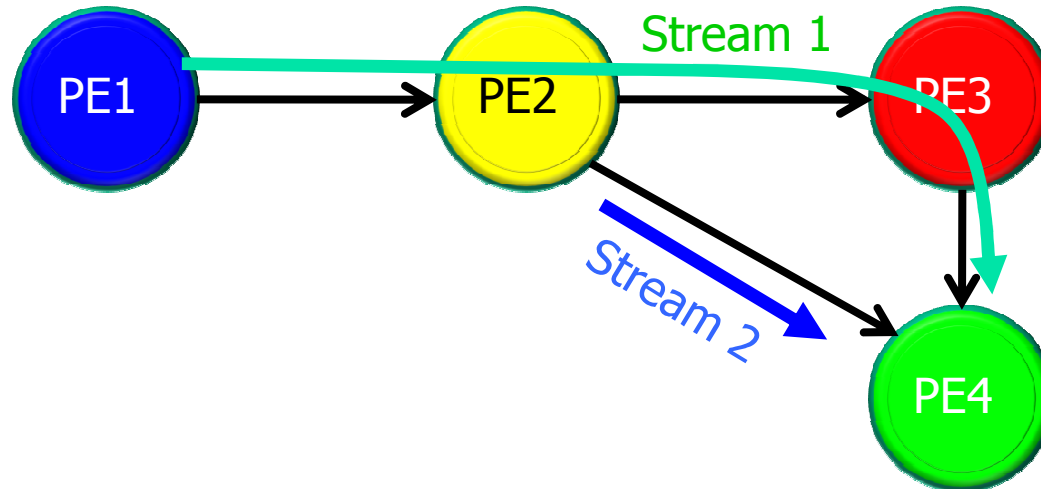
- The mapping algorithm should be aware of replicated modules!

Classic Performance Constraints

- Pair-wise **point-to-point** requirements
- For example, in a 4-module system:

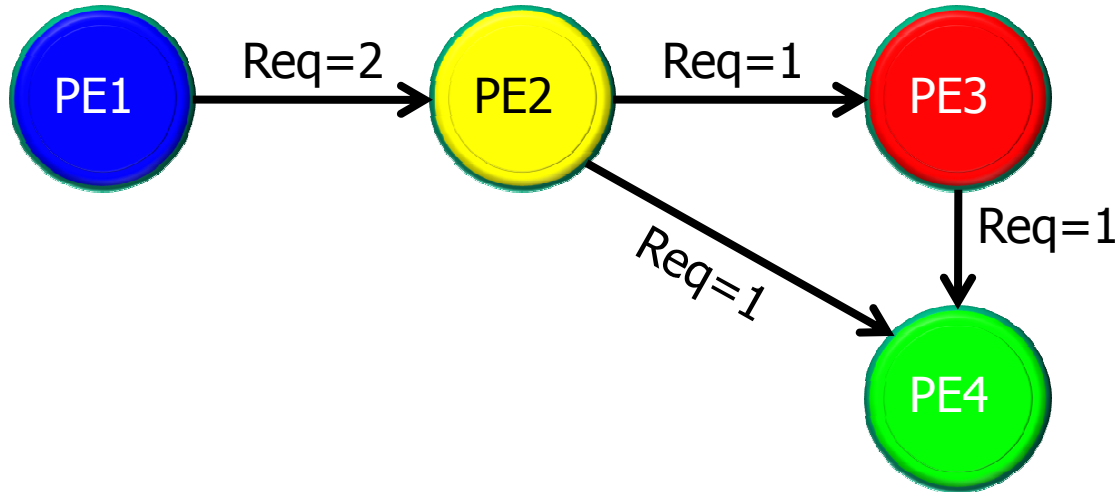


Motivation - Example #2

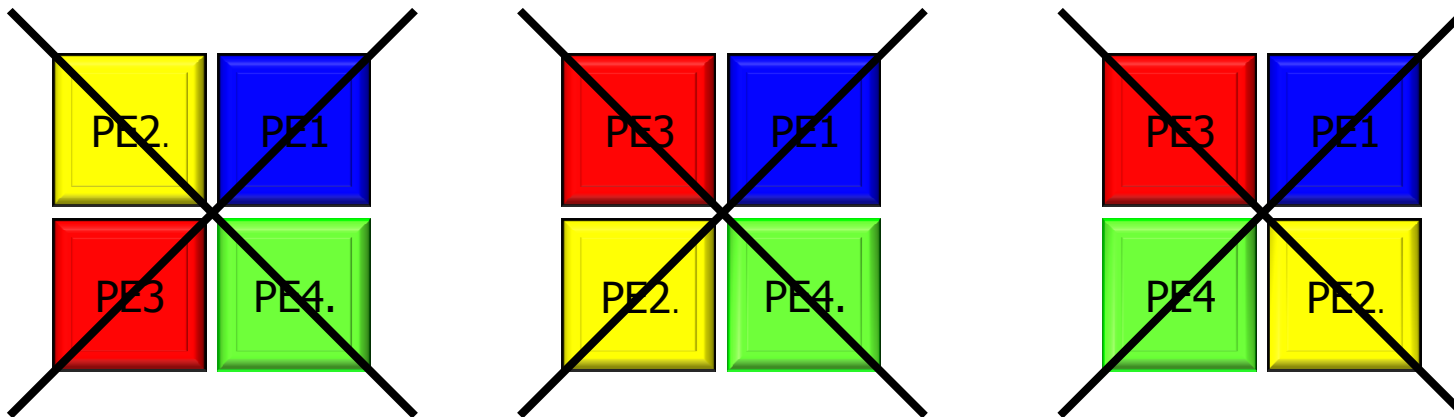


Stream ID	PEs	Timing Requirement
Stream 1	PE1 → PE2 → PE3 → PE4	4
Stream 2	PE2 → PE4	1

Example #2 – Pair-wise req.

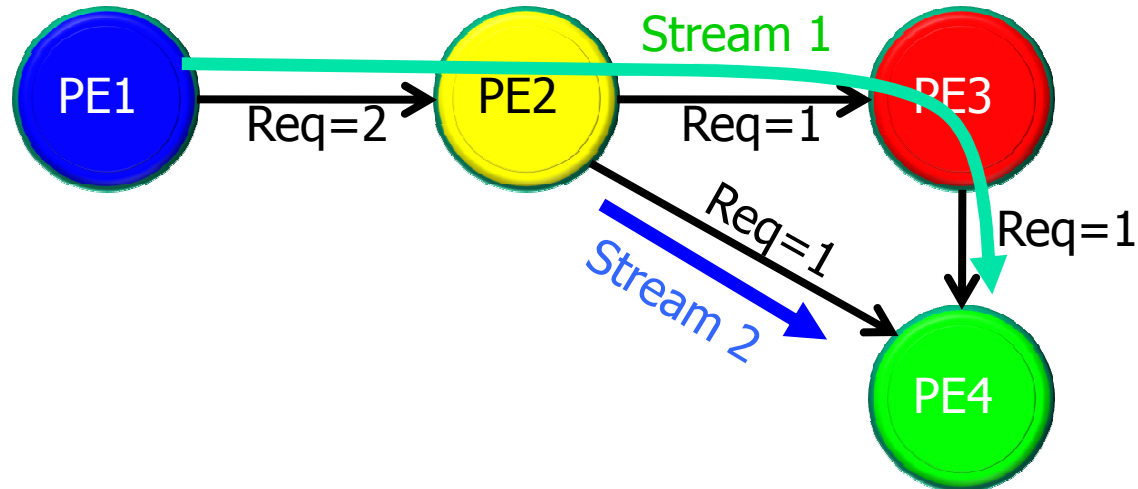


PE2	2		
PE3.		1	
PE4		1	1
	PE1	PE2	PE3

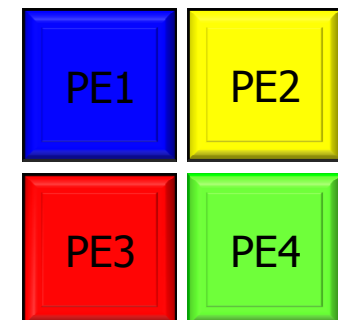


- No feasible mapping!

Application-Level Requirements



Stream ID	PEs	Requirement
Stream 1	PE1→PE2→PE3→PE4	4
Stream 2	PE2→PE4	1

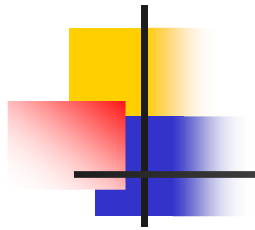


- A feasible mapping does exist!
- It's better to work with the application level requirements



This Work

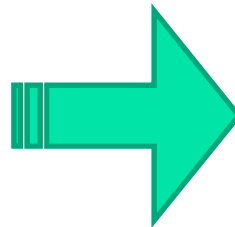
- Find efficient mappings by **extending the formulation of the mapping problem**
 - Adding degrees of freedom
- Degree of freedom #1
 - Leverage existence of replicated modules
- Degree of freedom #2
 - Replace p2p constraints with end-to-end, application-level requirements



Modifying the Formulation (1)

- Leverage existence of replicated modules
 - Allow the mapping algorithm to allocate flows to the best replicated module

Flow	BW	Time Req.
$PE_1 \rightarrow DSP_3$	100	3
$PE_2 \rightarrow DSP_4$	200	12
$PE_2 \rightarrow SRAM_1$	100	15
$PE_3 \rightarrow SRAM_2$	100	5
...



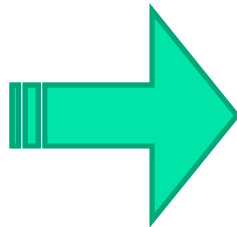
Flow	BW	Time Req.
$PE_1 \rightarrow \langle \text{ANY DSP} \rangle$	100	3
$PE_2 \rightarrow \langle \text{ANY DSP} \rangle$	200	12
$PE_2 \rightarrow \langle \text{ANY SRAM} \rangle$	100	15
$PE_3 \rightarrow \langle \text{ANY SRAM} \rangle$	100	5
...

Modifying the Formulation (2)

- Replace p2p constraints with end-to-end, application-level requirements

∞									
2	∞								
3	4	4							
3	1	3	∞						
3	7	7	4	∞					
∞	4	2	∞	3	∞				
5	∞	3	2	7	∞	4			
∞	6	5	1	∞	2	∞	∞		
1	∞	3	5	∞	∞	3	7	7	
1	3	∞	2	1	2	3	∞	3	∞

P2P timing req.



Stream ID	Stream's PEs	E2E Req.
1	PE ₁ →PE ₃ →PE ₉ →PE ₄ →PE ₁₀	23
2	PE ₅ →PE ₂ →PE ₃ →PE ₈ →PE ₇ →PE ₆ →PE ₁₀	12
3	PE ₅ →PE ₃ →PE ₉	15
4	PE ₇ →PE ₈ →PE ₂ →PE ₃	20
5	PE ₁ →PE ₂	2
...

E2E timing req.

- In this paper, for synthetic task graphs
 - Did so for a real application too



Outline

- The Era of Many Module SoC
- Revisiting the Mapping Problem
- **Cross-Entropy Optimization**
- Evaluation



Cross Entropy Optimization

- Modern optimization heuristic
 - Good at combinatorial optimization problems
- Akin to evolutionary algorithms
 - Generation of new solutions is based on sampling and estimation
- Inherently a global search method
 - Reduced risk of getting trapped in a local minimum



Cross Entropy Optimization

- Given an initial parameter vector $v=v_0$, sample a random population of K solutions x_1, x_2, \dots, x_k from the distribution given by $f(x;v)$.
- Evaluate the costs $S(x_i), i=1, \dots, K$.
- Using the ρK ($0 < \rho < 1$) elite (lowest cost) samples, obtain a new density function $f(x;v)$ by calculating a new vector v via Maximum Likelihood (ML) estimation.
- Repeat steps 1-3 with the new vector v unless maximum number of iterations is reached or no improvement is obtained for a predefined number of iterations.

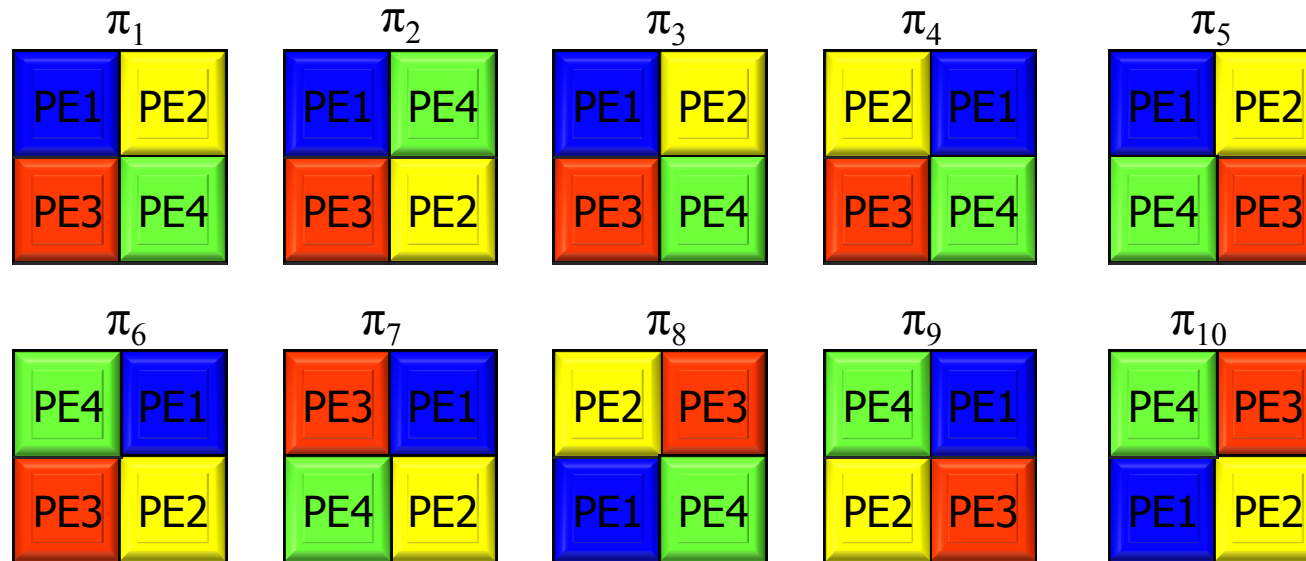
■ For example:

1. Generate 10 random mappings: $\pi_1, \pi_2, \dots, \pi_{10}$
2. Find 3 lowest cost mappings: π_2, π_5, π_7
3. Examine those 3 best mappings:
 - A. For each tile, calculate the probability core PE_i is mapped to that tile
 - B. Update probabilities accordingly



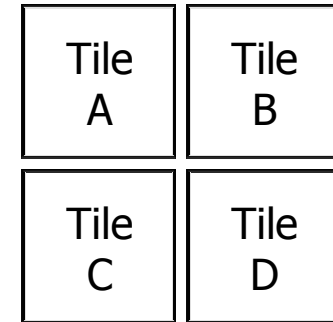
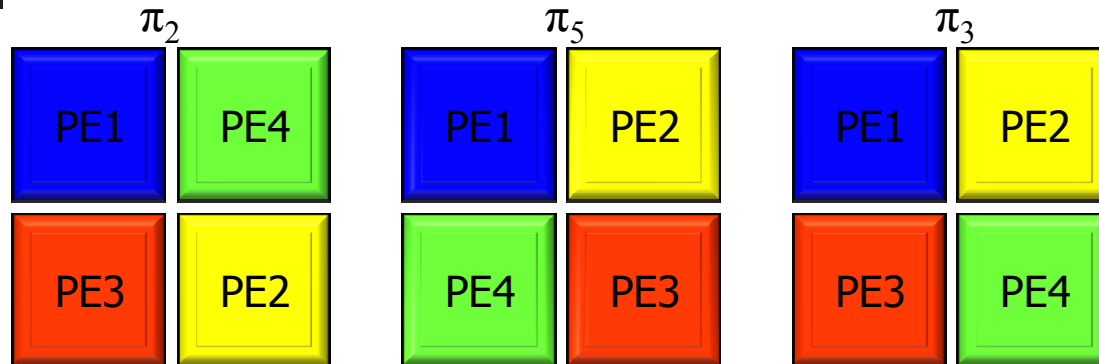
CE Example

Tile A	Tile B
Tile C	Tile D



$\text{Prob}(\text{TileA} \leftarrow \text{PE}_1) = \text{Prob}(\text{TileA} \leftarrow \text{PE}_2) = \text{Prob}(\text{TileA} \leftarrow \text{PE}_3) = \text{Prob}(\text{TileA} \leftarrow \text{PE}_4) = 0.25$
 $\text{Prob}(\text{TileB} \leftarrow \text{PE}_1) = \text{Prob}(\text{TileB} \leftarrow \text{PE}_2) = \text{Prob}(\text{TileB} \leftarrow \text{PE}_3) = \text{Prob}(\text{TileB} \leftarrow \text{PE}_4) = 0.25$
 $\text{Prob}(\text{TileC} \leftarrow \text{PE}_1) = \text{Prob}(\text{TileC} \leftarrow \text{PE}_2) = \text{Prob}(\text{TileC} \leftarrow \text{PE}_3) = \text{Prob}(\text{TileC} \leftarrow \text{PE}_4) = 0.25$
 $\text{Prob}(\text{TileD} \leftarrow \text{PE}_1) = \text{Prob}(\text{TileD} \leftarrow \text{PE}_2) = \text{Prob}(\text{TileD} \leftarrow \text{PE}_3) = \text{Prob}(\text{TileD} \leftarrow \text{PE}_4) = 0.25$

Updating Probabilities



- $\text{Prob}(\text{TileA} \leftarrow \text{PE}_1) = 1$
- $\text{Prob}(\text{TileB} \leftarrow \text{PE}_2) = 2/3$
- $\text{Prob}(\text{TileB} \leftarrow \text{PE}_4) = 1/3$
- $\text{Prob}(\text{TileC} \leftarrow \text{PE}_3) = 2/3$
- $\text{Prob}(\text{TileD} \leftarrow \text{PE}_2) = 1/3$
- $\text{Prob}(\text{TileD} \leftarrow \text{PE}_3) = 1/3$
- $\text{Prob}(\text{TileD} \leftarrow \text{PE}_4) = 1/3$
- Following iteration uses these updates probabilities
- Gradually, probabilities converge to 0/1



Outline

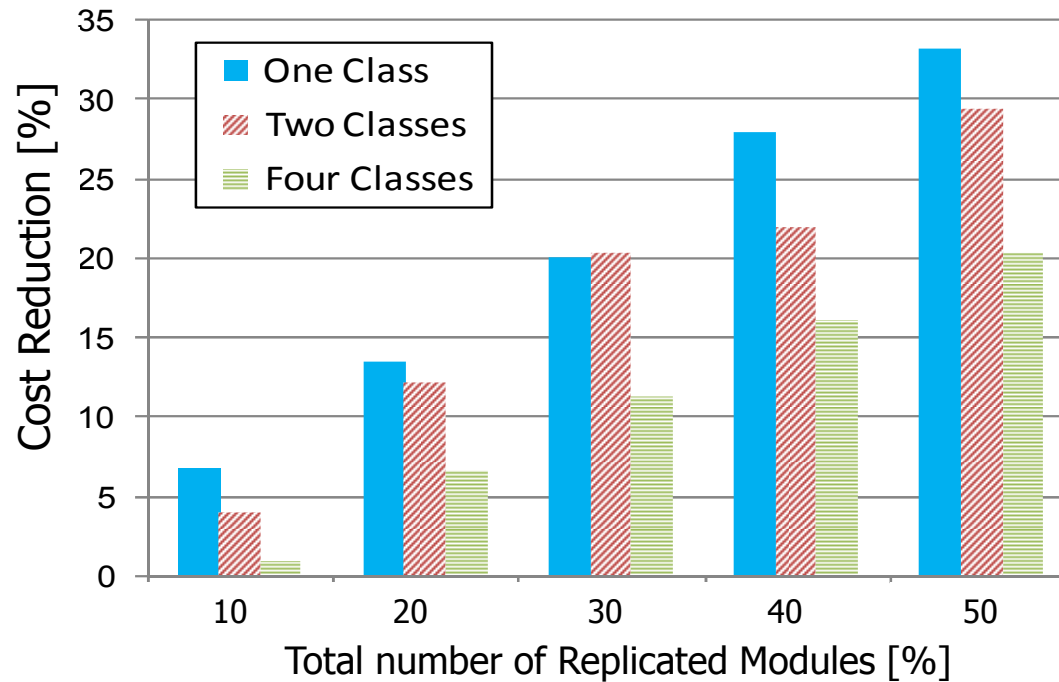
- The Era of Many Module SoC
- Revisiting the Mapping Problem
- Cross-Entropy Optimization
- **Evaluation**



Evaluation

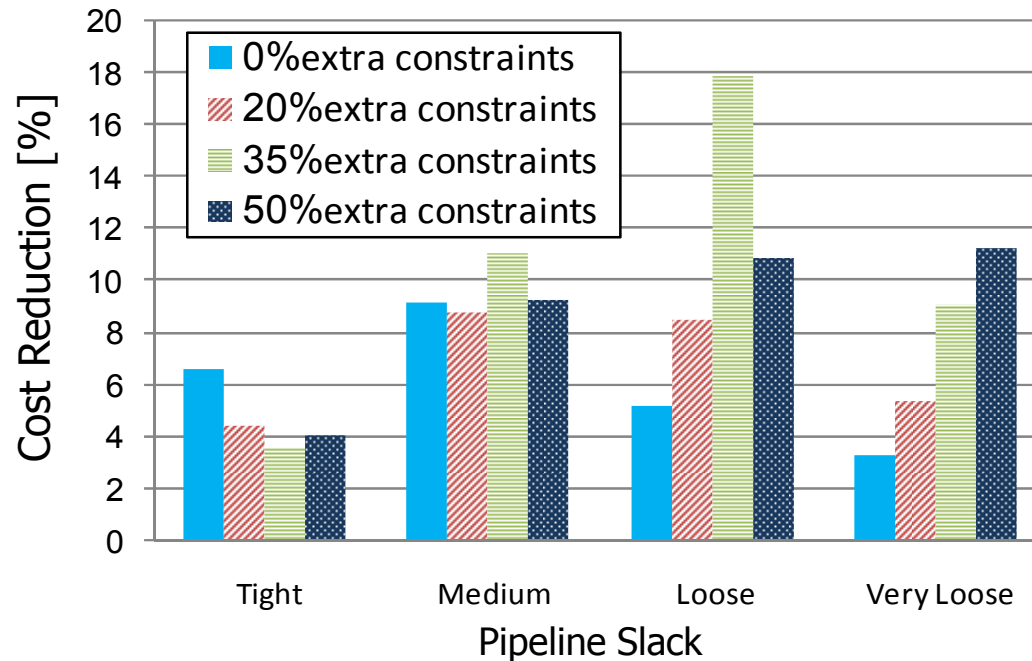
- Scenario
 - 6x6 mesh NoC
 - Synthetic, randomized SoC
 - Task graphs (and task-to-core mapping)
 - Varying number of replicated modules
 - Varying timing constraints
 - (Real application in DATE10 paper)
- Compare with best cost of classic mapping
 - Averaging multiple runs

Accounting for Replication



- “Class”: a group of identical PEs
 - Total number of replicated cores=
 $\{\text{Number of classes}\} * \{\text{class size}\}$

Application-Level Requirements

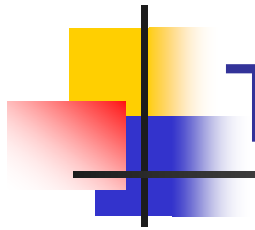


- SoCs with a pipeline data path and background P2P traffic
 - Varying pipeline slack
 - Different amounts of background constraints



Conclusions and Future Work

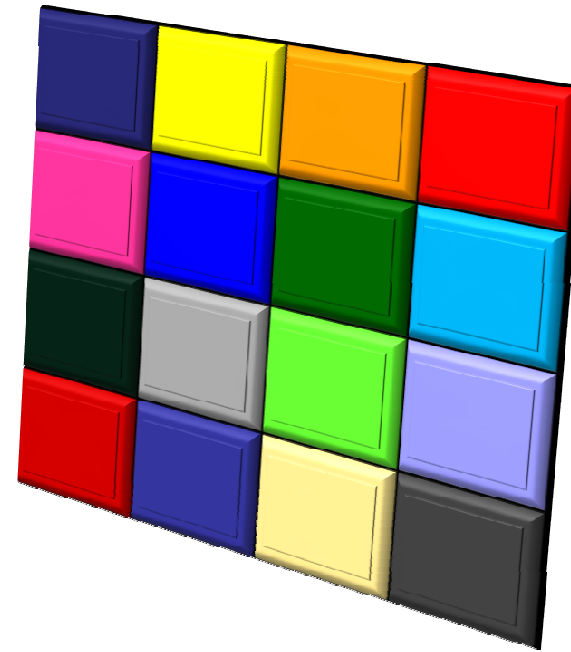
- We are going into the era of “Many module SoC”
- Extend the mapping to account for
 - Classes of replicated modules
 - Application-level requirements
- Meaningful power savings
- But mapping is an example
 - Routing? Task assignment? Link design? Topology selection?



The Era of Many-Module SoC

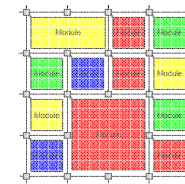
Thank you!

Questions?



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QNoC
Research
Group



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