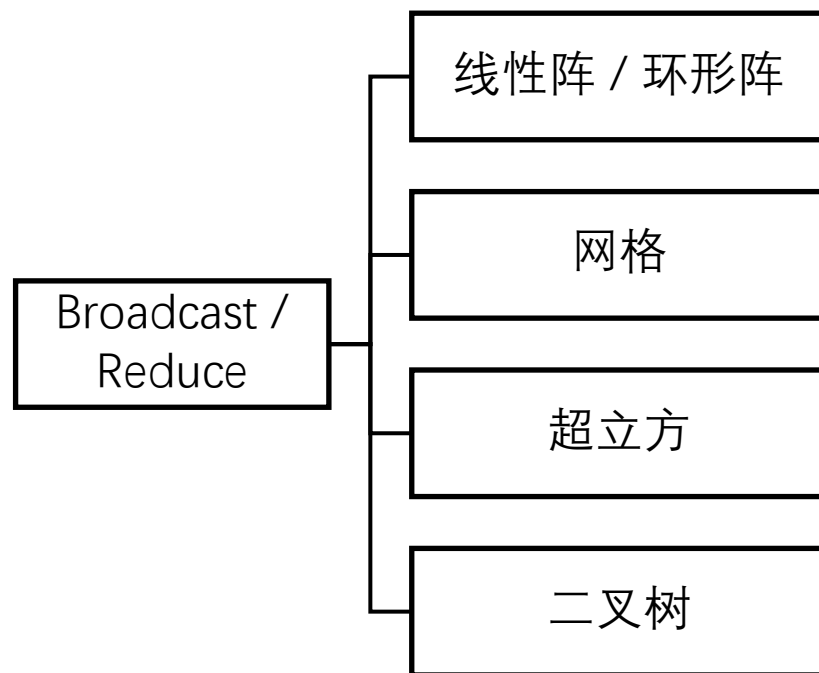
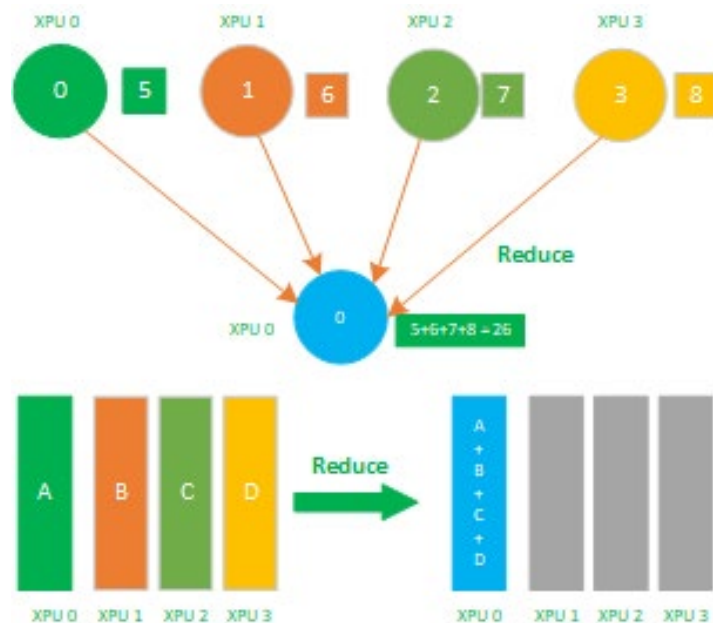
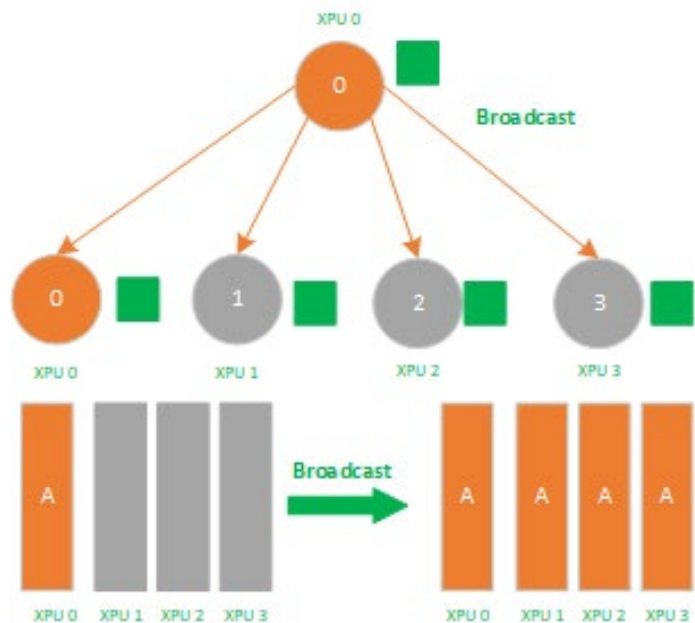
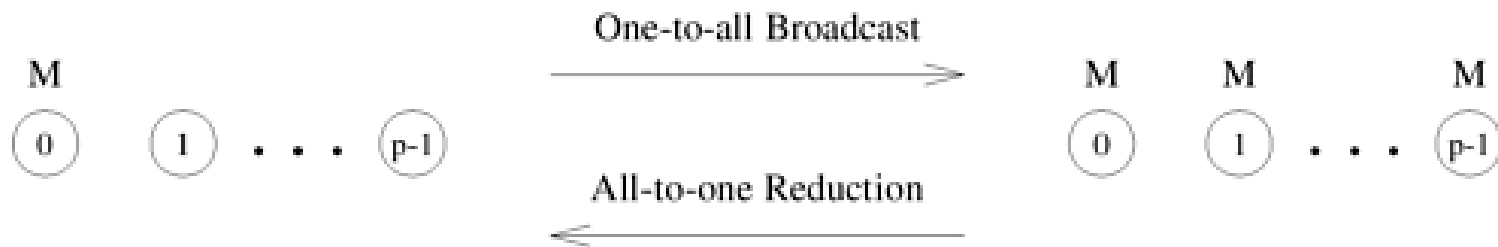


Chapter 4

Basic Communication Operations

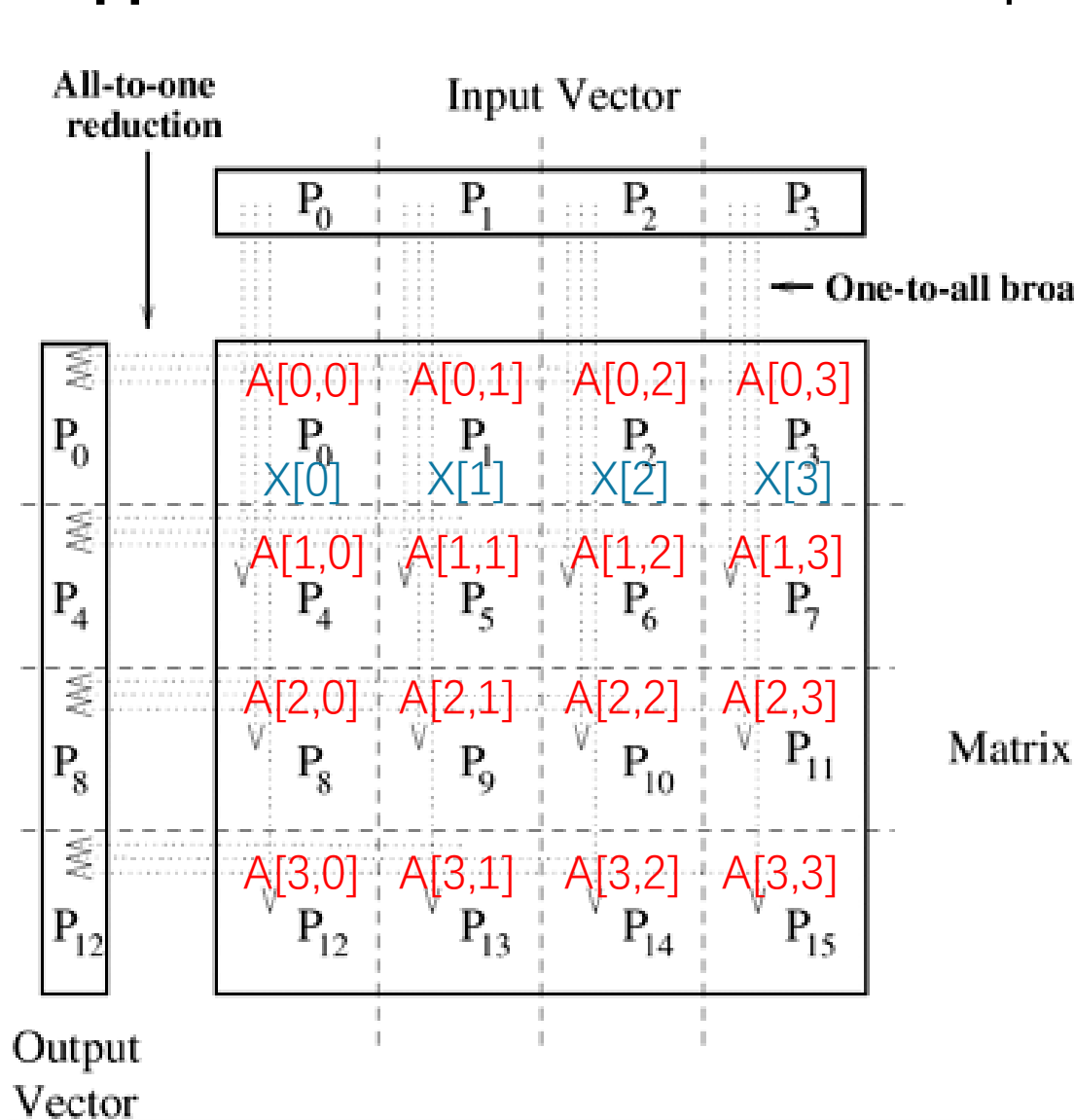
2024/06/24

4.1 One-to-All Broadcast / All-to-One Reduction



4.1 One-to-All Broadcast / All-to-One Reduction

Application Case: Matrix-Vector Multiplication



$$Y[i] = \underbrace{A[i,0]*X[0]}_{\text{位于P0}} + \underbrace{A[i,1]*X[1]}_{\text{位于P1}} + A[i,2]*X[2] + A[i,3]*X[3]$$

在P1, P5, P9, P13上计算

在P0, P4, P8, P12上计算

Step 1

P0/P1/P2/P3中对应的X数据Broadcast到对应的进程上

Step2

对应的进程进行计算

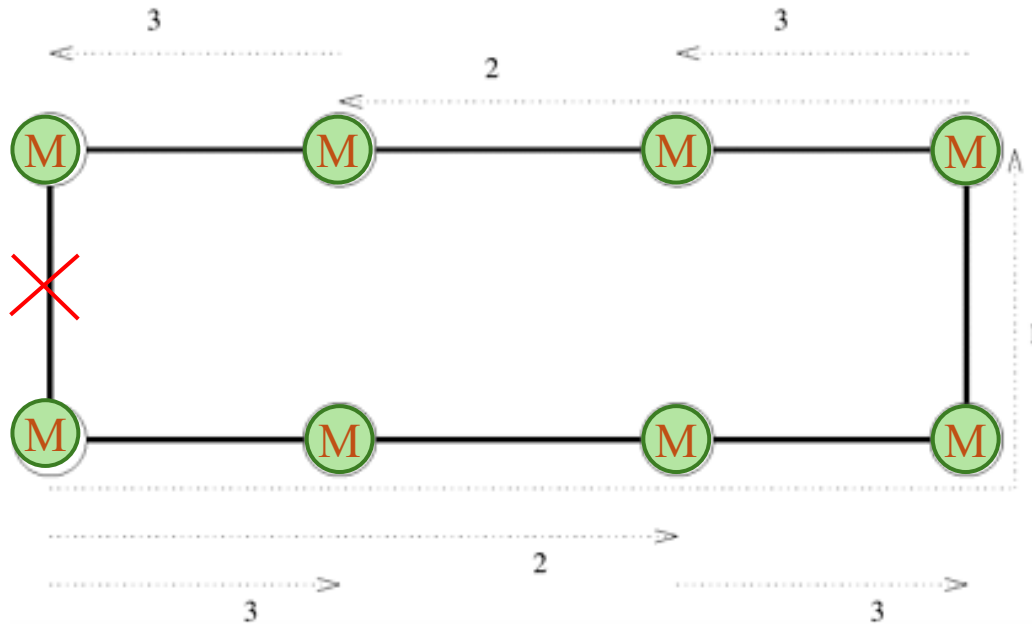
Step3

P1/P2/P3把计算结果Reduce到P0上
P5/P5/P7把计算结果Reduce到P4上

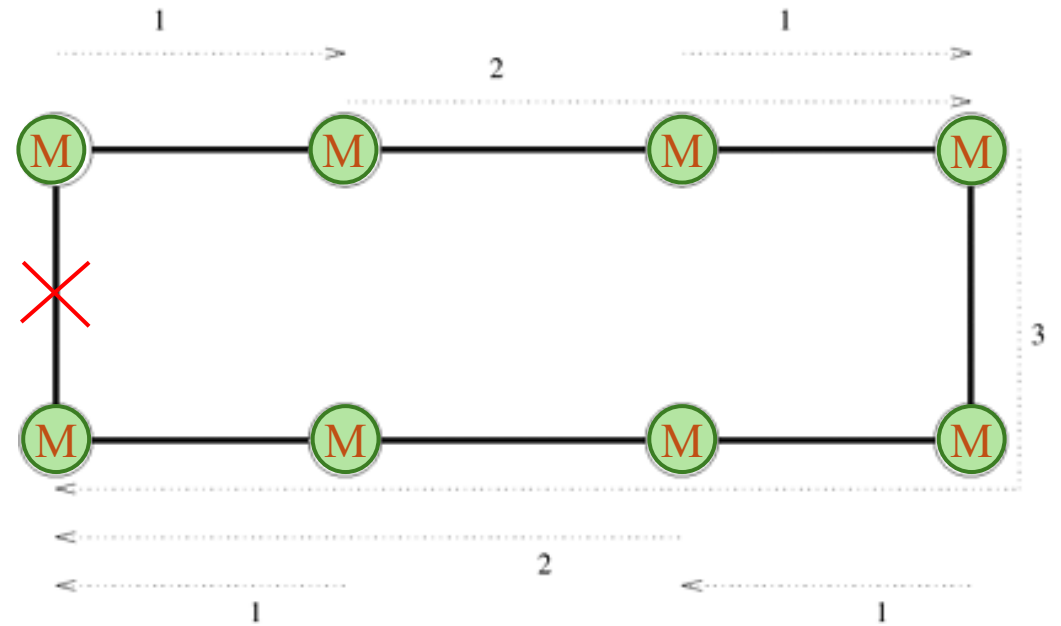
...

4.1 One-to-All Broadcast / All-to-One Reduction

Ring or Linear Array



Broadcast



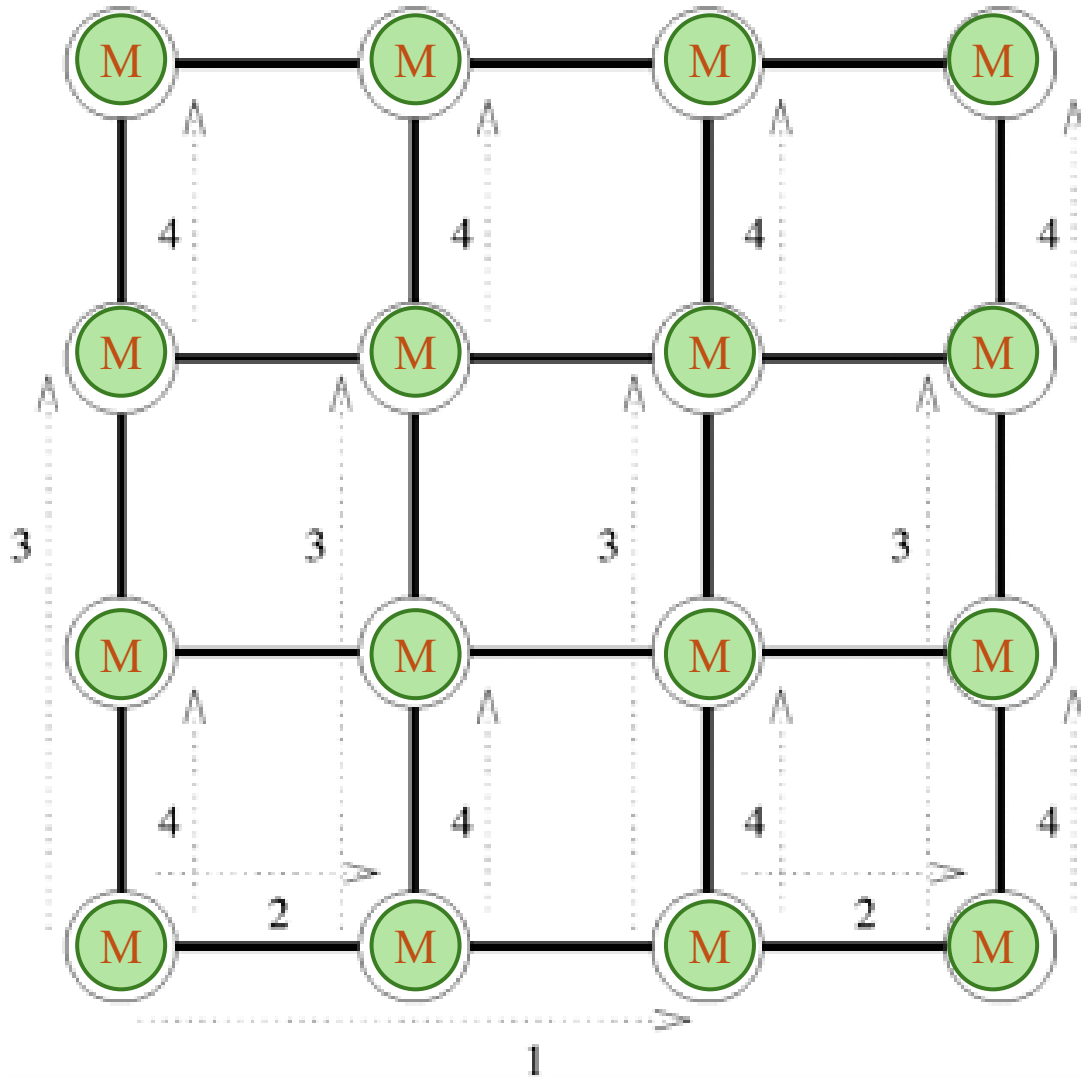
Reduce

$$T = (t_s + mt_w) + (t_s + mt_w) + (t_s + mt_w)$$

$$T = (t_s + mt_w) \log(p)$$

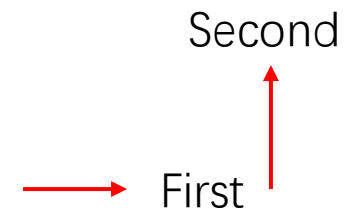
4.1 One-to-All Broadcast / All-to-One Reduction

Mesh Broadcast



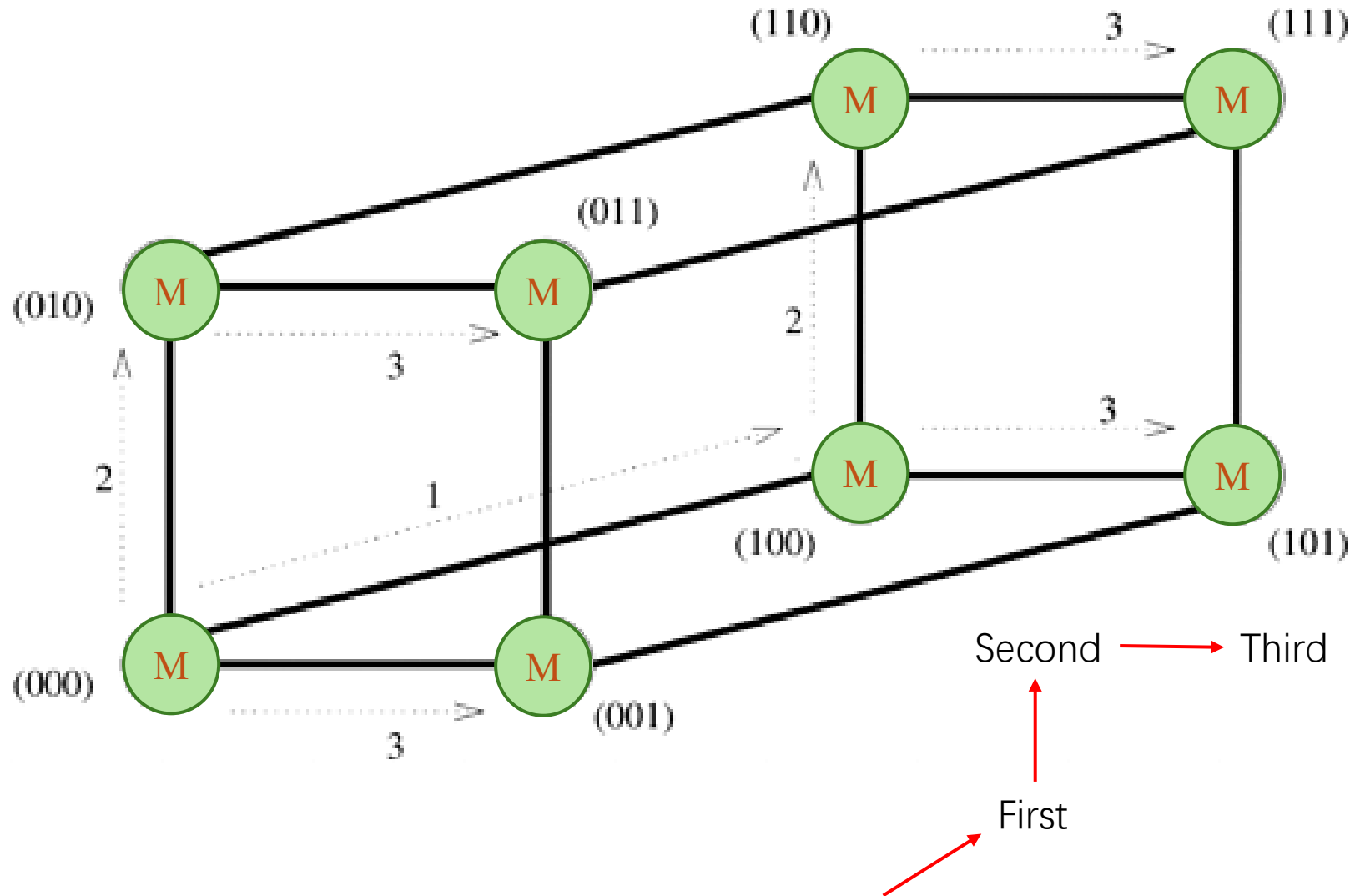
$$T = (t_s + mt_w) + (t_s + mt_w) + (t_s + mt_w) + (t_s + mt_w)$$

$$T = (t_s + mt_w) \log(p)$$



4.1 One-to-All Broadcast / All-to-One Reduction

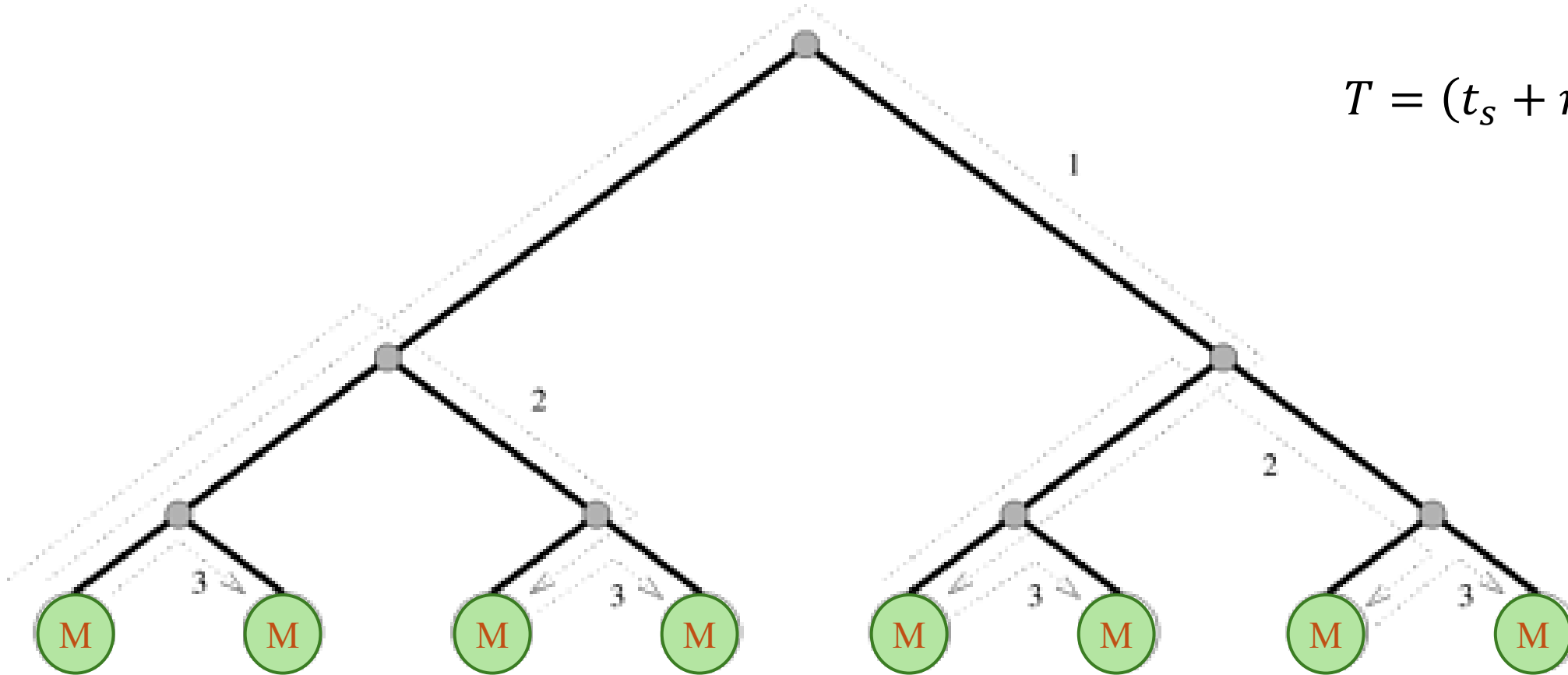
Hypercube Broadcast



$$T = (t_s + mt_w) \log(p)$$

4.1 One-to-All Broadcast / All-to-One Reduction

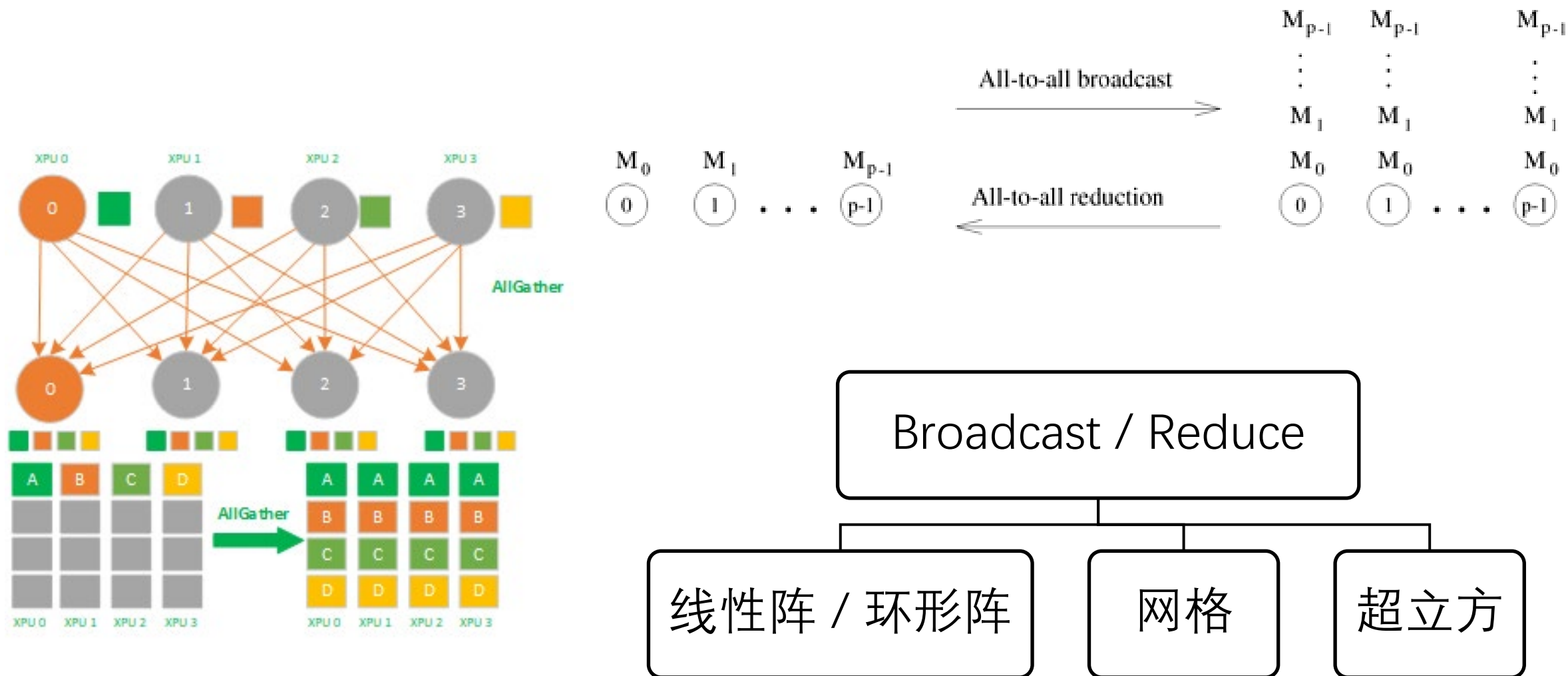
Balanced Binary Tree Broadcast



$$T = (t_s + mt_w) \log(p)$$

4.2 All-to-All Broadcast (All Gather) / All-to-All Reduction (Reduce Scatter)

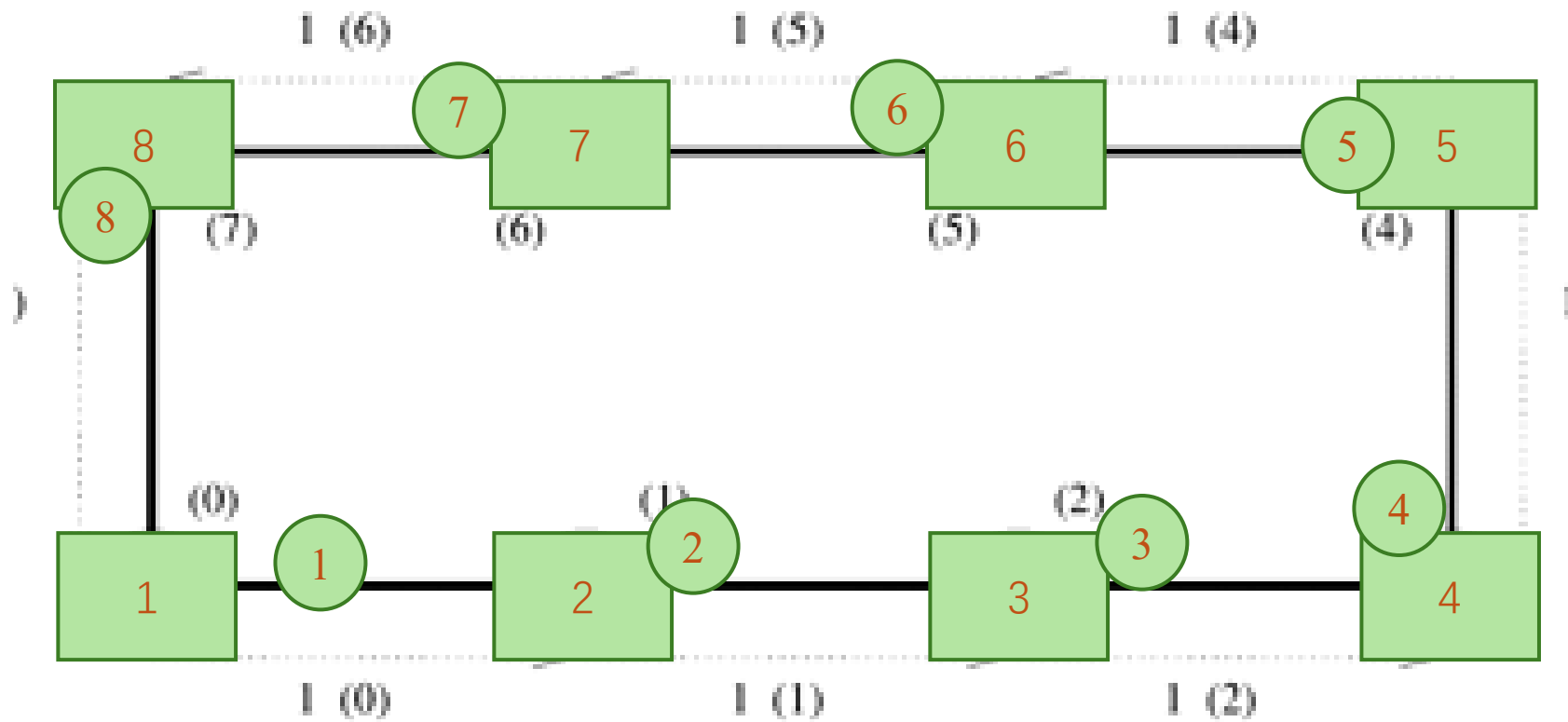
All-to-all broadcast is a generalization of one-to-all broadcast in which all p nodes simultaneously initiate a broadcast



4.2 All-to-All Broadcast (All Gather) / All-to-All Reduction (Reduce Scatter)

Linear Array and Ring Broadcast

The same procedure would also work on a linear array with bidirectional links.

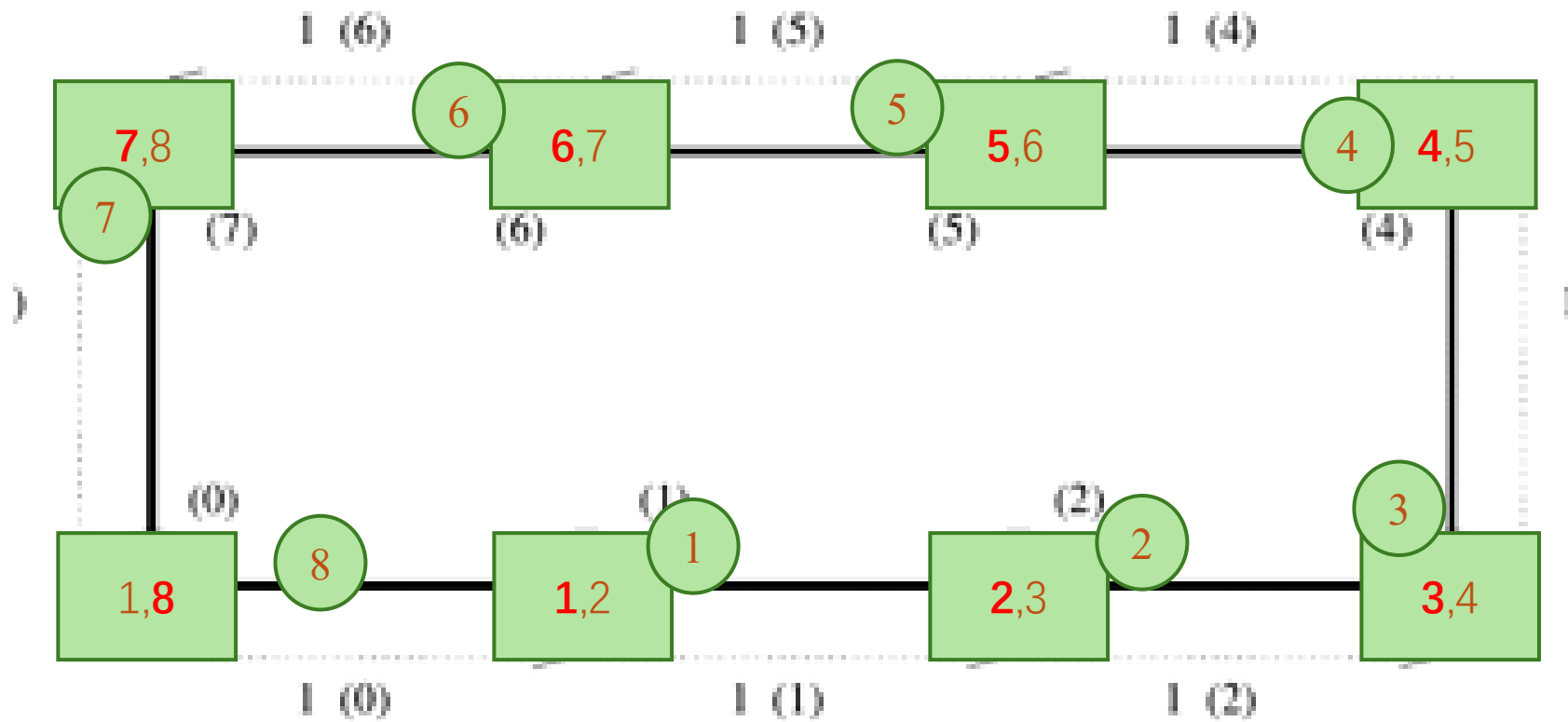


$$T = (t_s + mt_w)$$

4.2 All-to-All Broadcast (All Gather) / All-to-All Reduction (Reduce Scatter)

Linear Array and Ring Broadcast

The same procedure would also work on a linear array with bidirectional links.

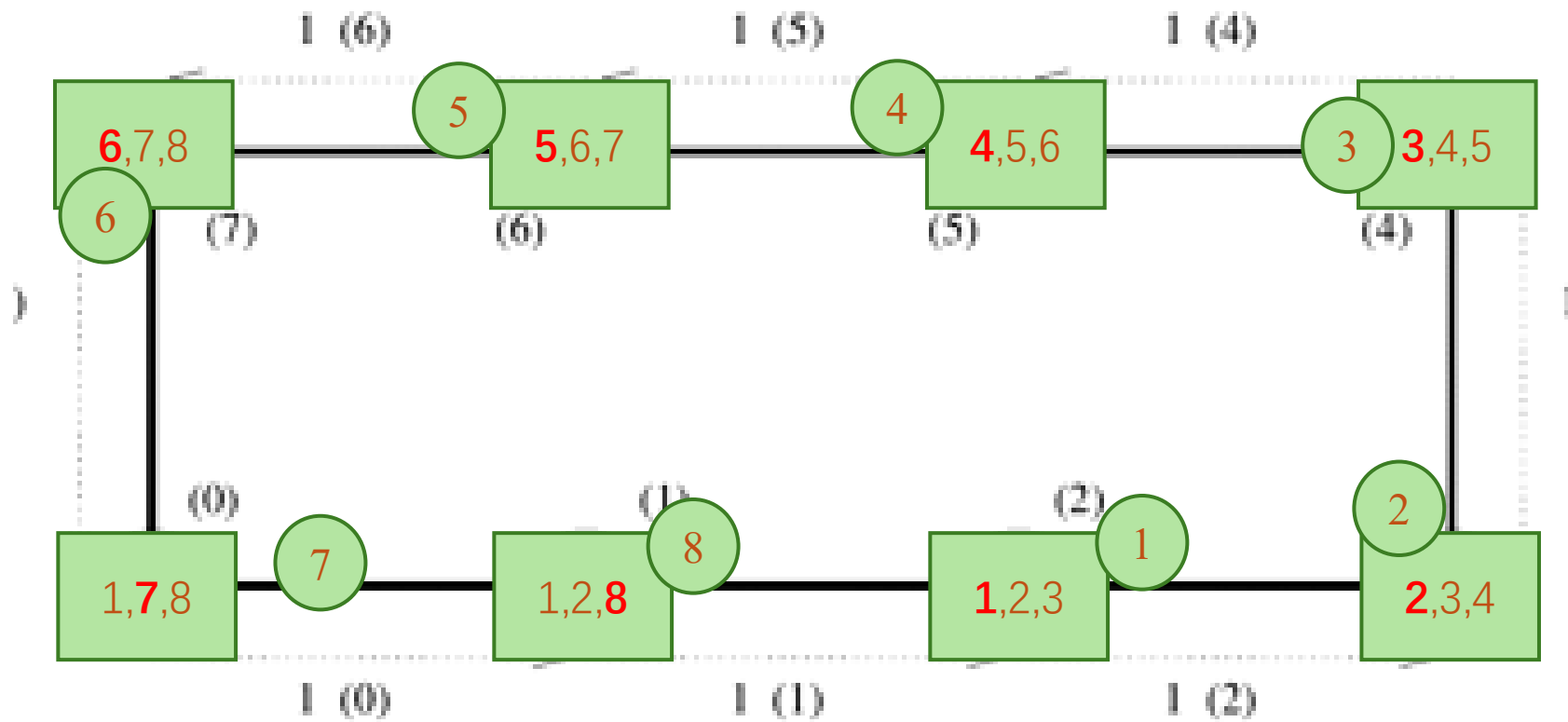


$$T = (t_s + mt_w) + (t_s + mt_w)$$

4.2 All-to-All Broadcast (All Gather) / All-to-All Reduction (Reduce Scatter)

Linear Array and Ring Broadcast

The same procedure would also work on a linear array with bidirectional links.

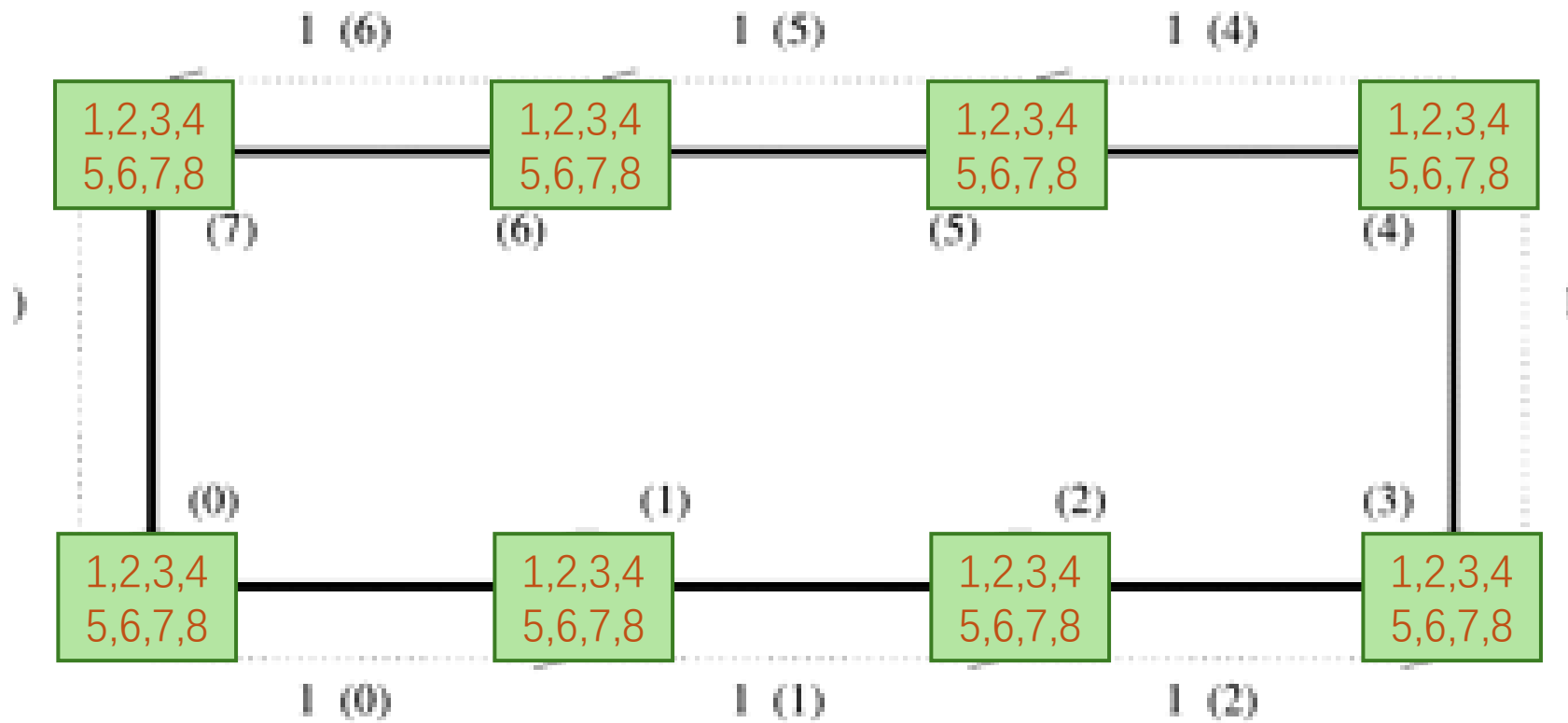


$$T = (t_s + mt_w) + (t_s + mt_w) + (t_s + mt_w)$$

4.2 All-to-All Broadcast (All Gather) / All-to-All Reduction (Reduce Scatter)

Linear Array and Ring Broadcast

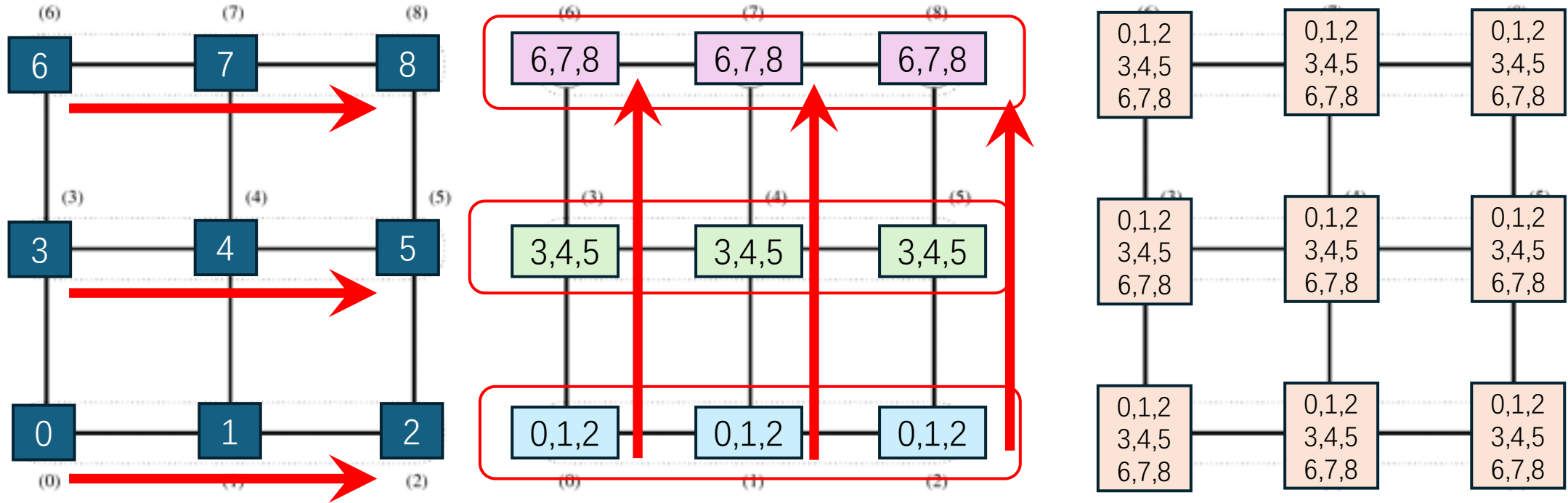
The same procedure would also work on a linear array with bidirectional links.



$$T = (t_s + mt_w)(p - 1)$$

4.2 All-to-All Broadcast (All Gather) / All-to-All Reduction (Reduce Scatter)

Mesh Broadcast



$$(t_s + m_1 t_w)(\sqrt{p} - 1)$$

$$m_1 = m$$

+

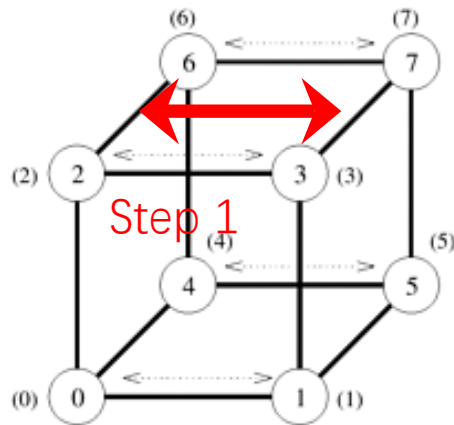
$$(t_s + m_2 t_w)(\sqrt{p} - 1)$$

$$m_2 = m_1 * \sqrt{p}$$

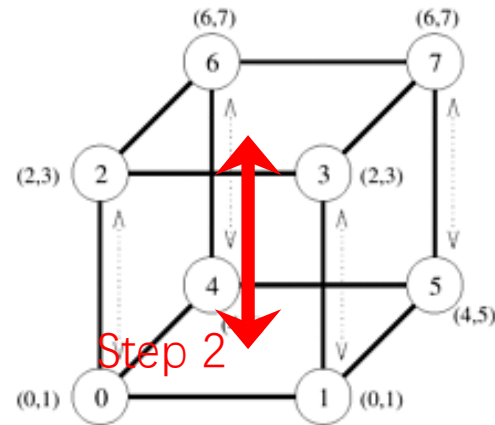
$$T = 2t_s(\sqrt{p} - 1) + mt_w(p - 1)$$

4.2 All-to-All Broadcast (All Gather) / All-to-All Reduction (Reduce Scatter)

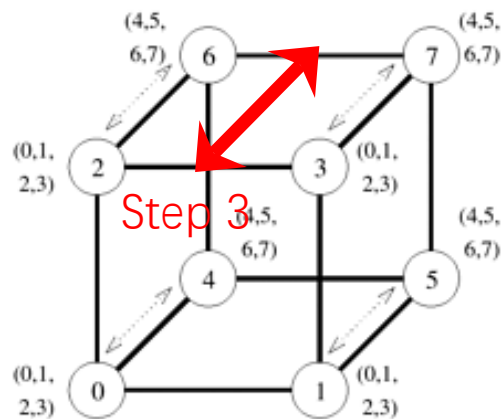
Hypercube Broadcast



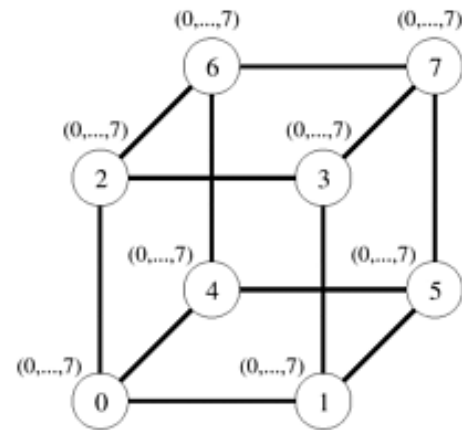
(a) Initial distribution of messages



(b) Distribution before the second step



(c) Distribution before the third step



(d) Final distribution of messages

Step 1

$$T_1 = (t_s + m_1 t_w), m_1 = m = 2^0 m$$

Step 2

$$T_2 = (t_s + m_2 t_w), m_2 = 2m_1 = 2^1 m$$

Step 3

$$T_3 = (t_s + m_3 t_w), m_3 = 2m_2 = 2^2 m$$

Step n

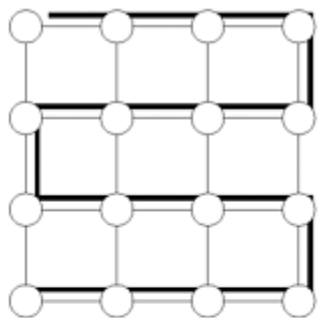
$$T_n = (t_s + m_n t_w), m_n = 2^{n-1} m$$

$$T = \sum_{i=1}^{\log p} (t_s + 2^{i-1} m t_w)$$

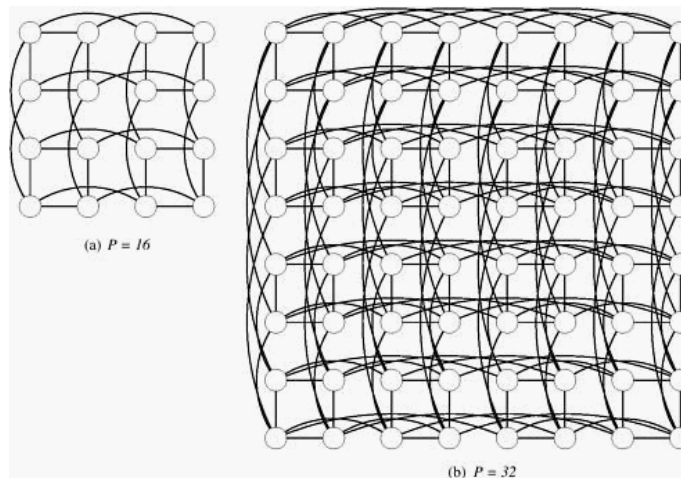
$$= t_s \log p + m t_w (p - 1)$$

4.2 All-to-All Broadcast (All Gather) / All-to-All Reduction (Reduce Scatter)

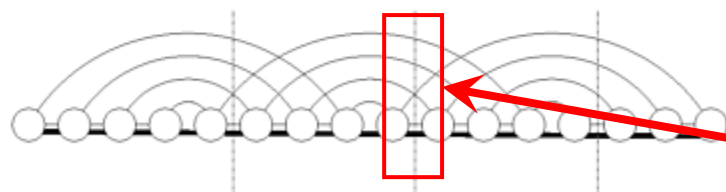
Caution



(a) Mapping a linear array into a 2D mesh (congestion 1).

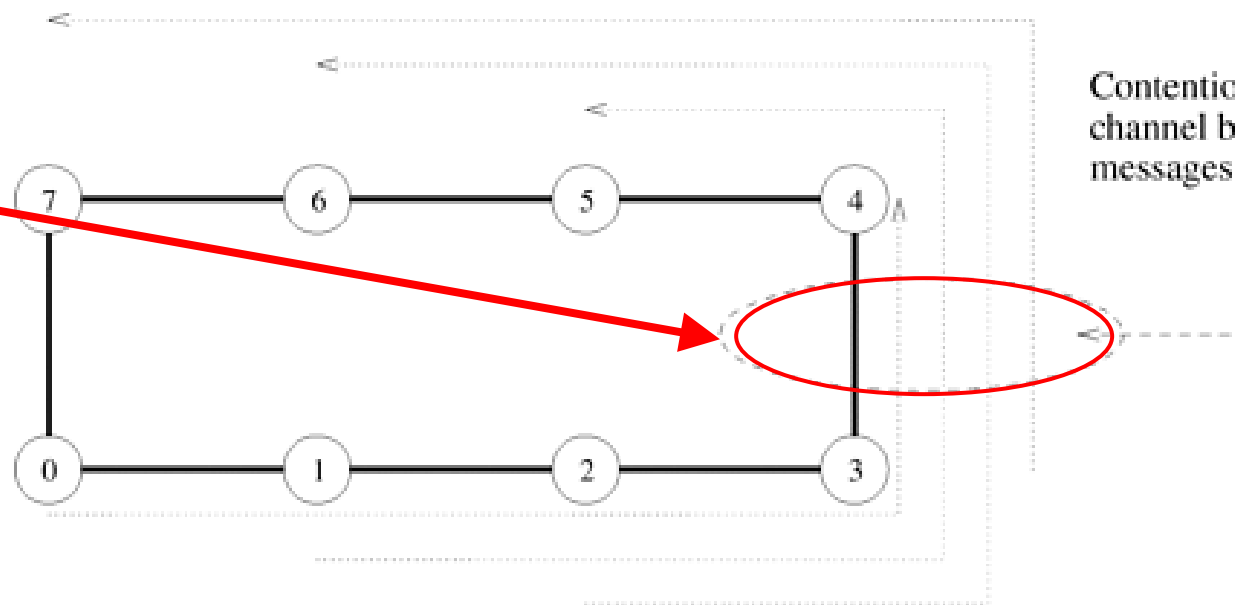


HyperCube到Mesh的映射



(b) Inverting the mapping – mapping a 2D mesh into a linear array (congestion 5)

Mesh到Linear的映射



Contention for a single channel by multiple messages

4.3 All-Reduce / Prefix-Sum

All Reduce

All-to-All Reduction: p个进程同时进行One-to-All Reduction, 且Reduce的目的地不同

All Reduce: All-to-All 的一种扩展

All Reduce = All-to-One Reduction + All-to-All Broadcast

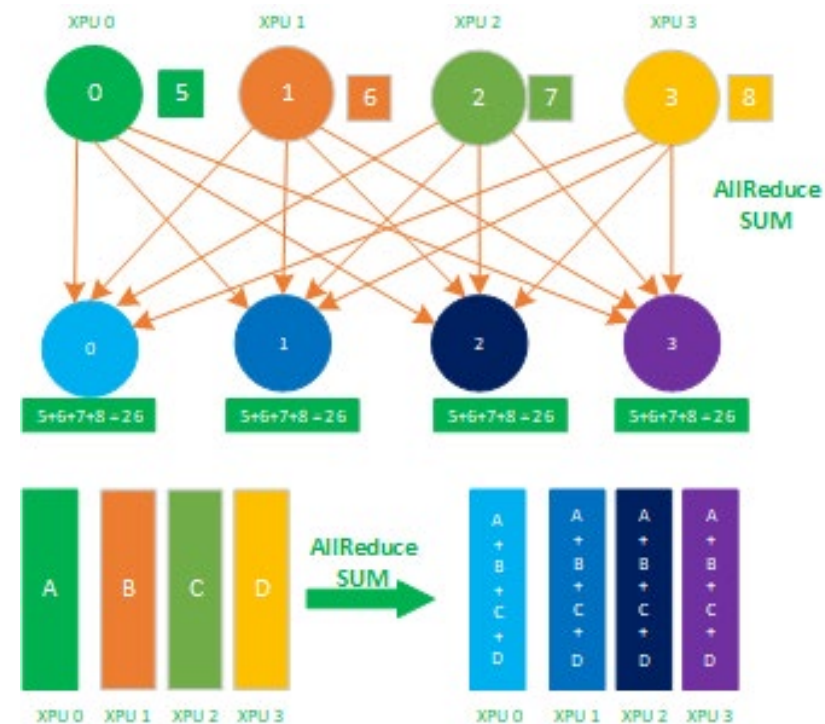
= All-to-All Broadcast * Reduce

All-to-All Broadcast

$$T = \sum_{i=1}^{\log p} (t_s + \underbrace{2^{i-1} m t_w}_{\text{通信的消息量不翻倍}})$$

All Reduce

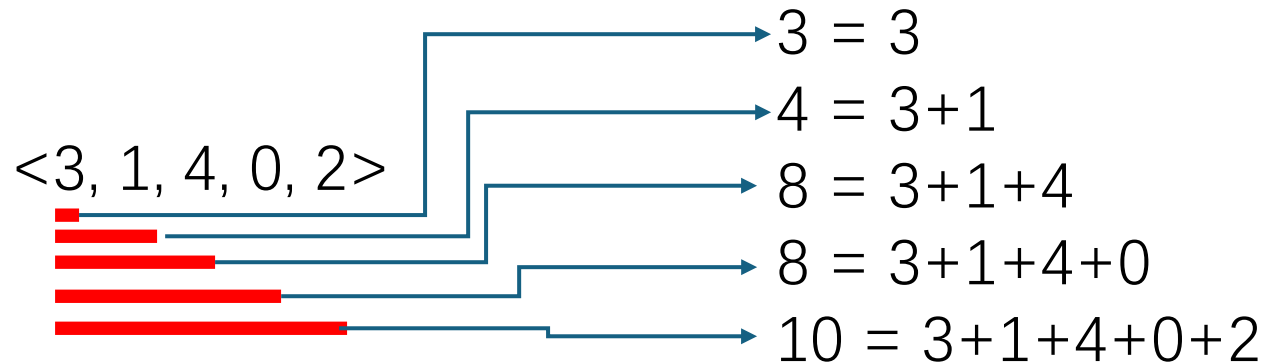
$$T = (t_s + m t_w) \log p$$



4.3 All-Reduce / Prefix-Sum

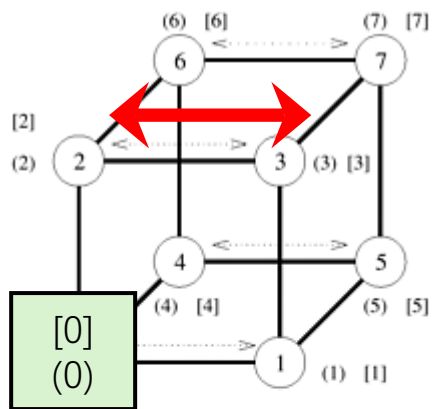
Prefix-Sum - Linear

Finding **prefix sums** (also known as the **scan** operation) is another important problem that can be solved by using a communication pattern similar to that used in **all-to-all broadcast** and **allreduce** operations.

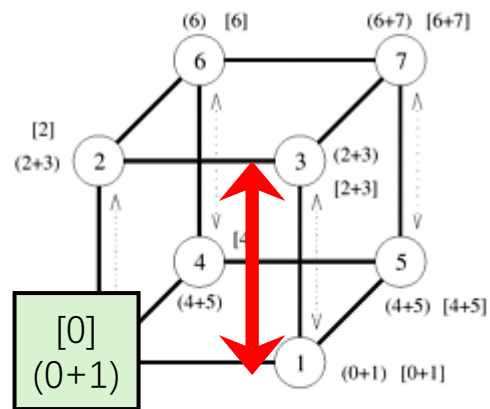


4.3 All-Reduce / Prefix-Sum

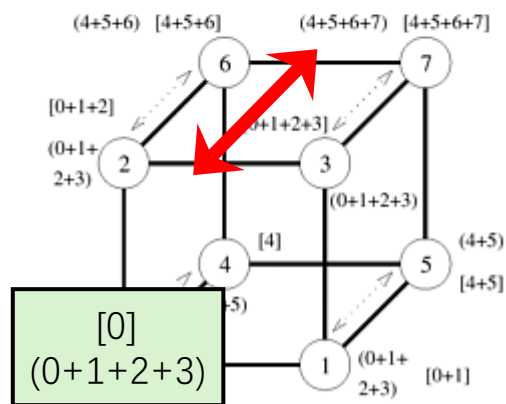
Prefix-Sum - Hypercube



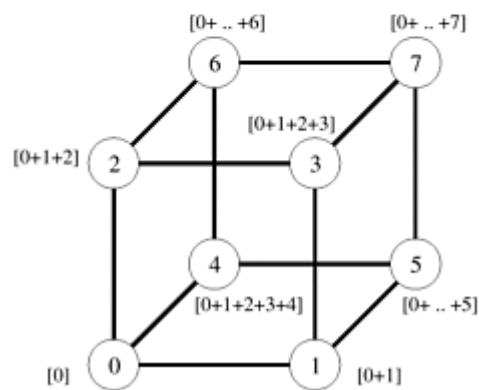
(a) Initial distribution of values



(b) Distribution of sums before second step



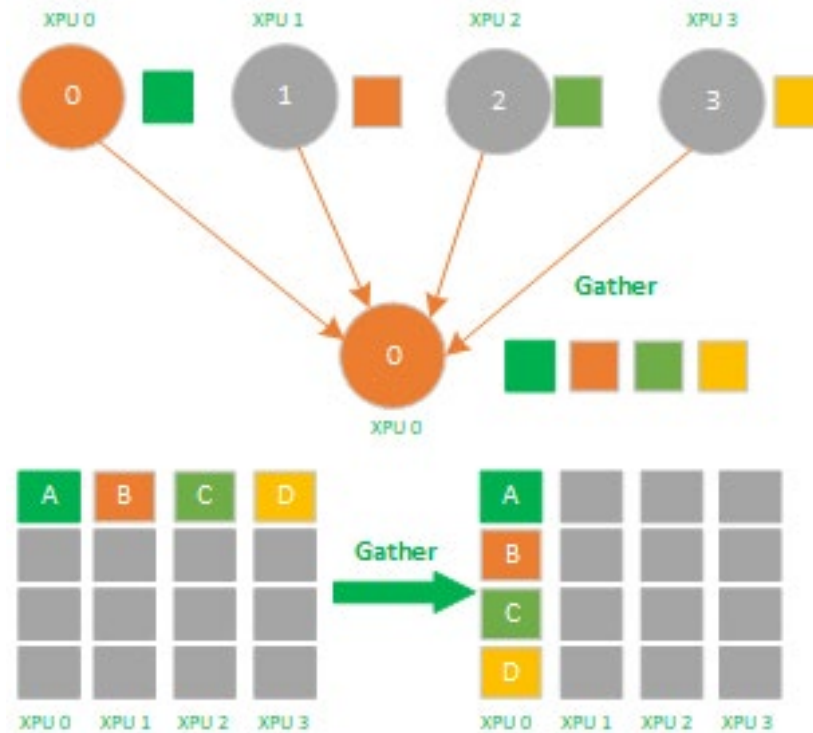
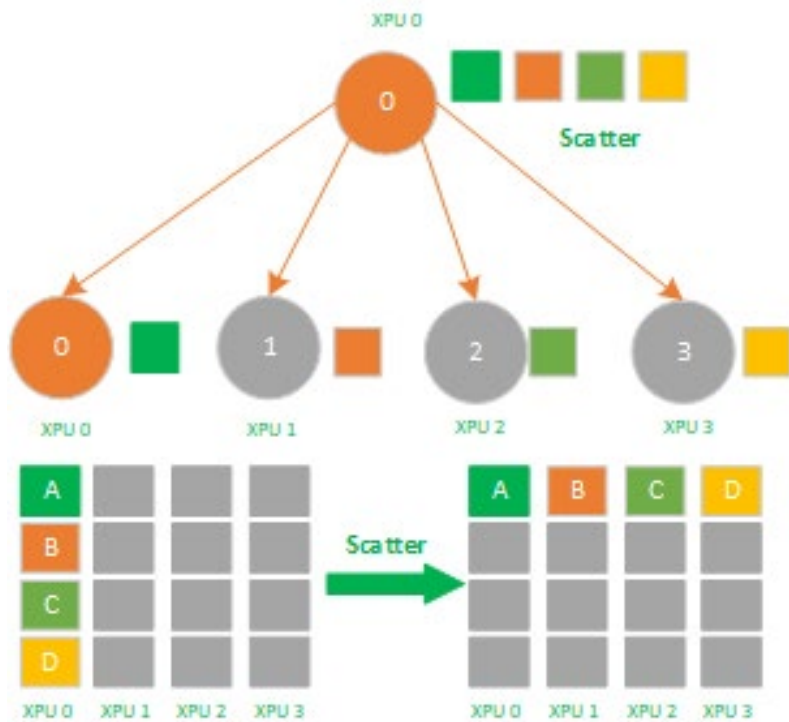
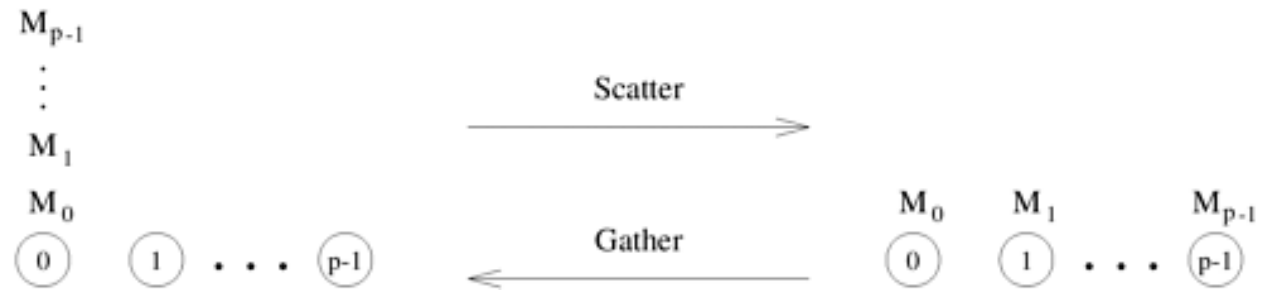
(c) Distribution of sums before third step



(d) Final distribution of prefix sums

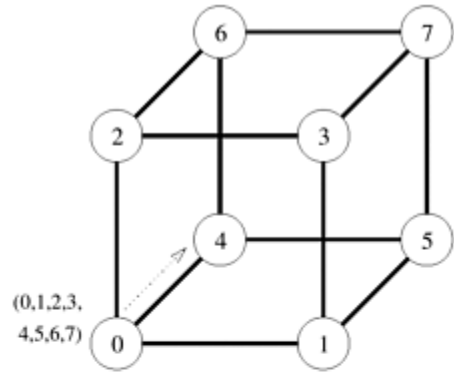
[方括号] 表示结果缓冲区中累积的本地Prefix-Sum
(圆括号) 表示下一步的传出信息缓冲区内容

4.4 Scatter (One-to-All Personalized Communication) / Gather

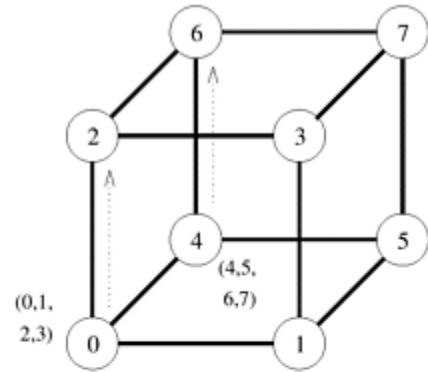


4.4 Scatter (One-to-All Personalized Communication) / Gather

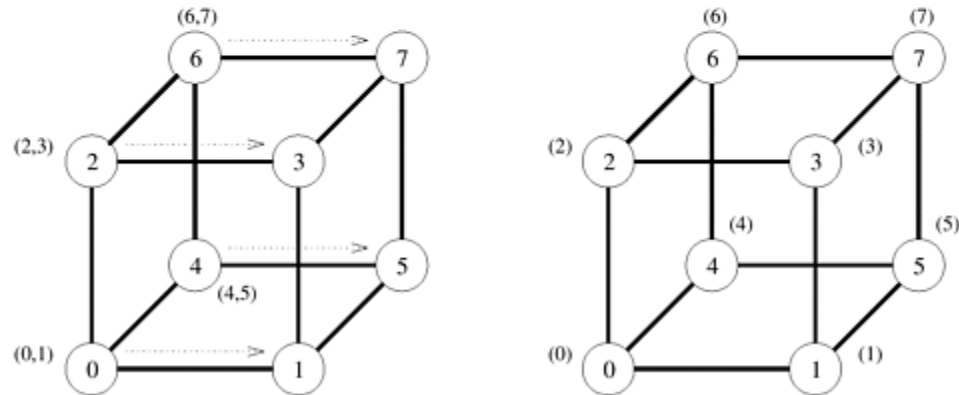
Hypercube



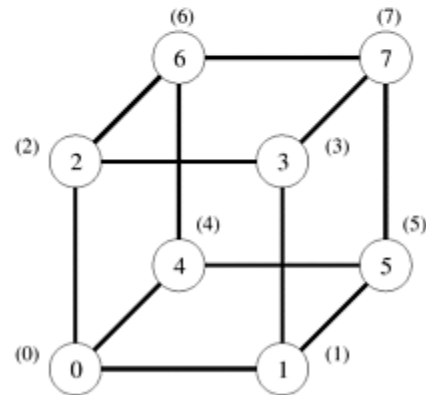
(a) Initial distribution of messages



(b) Distribution before the second step



(c) Distribution before the third step



(d) Final distribution of messages

Step 1

$$T_1 = (t_s + m \frac{p}{2} t_w)$$

Step 2

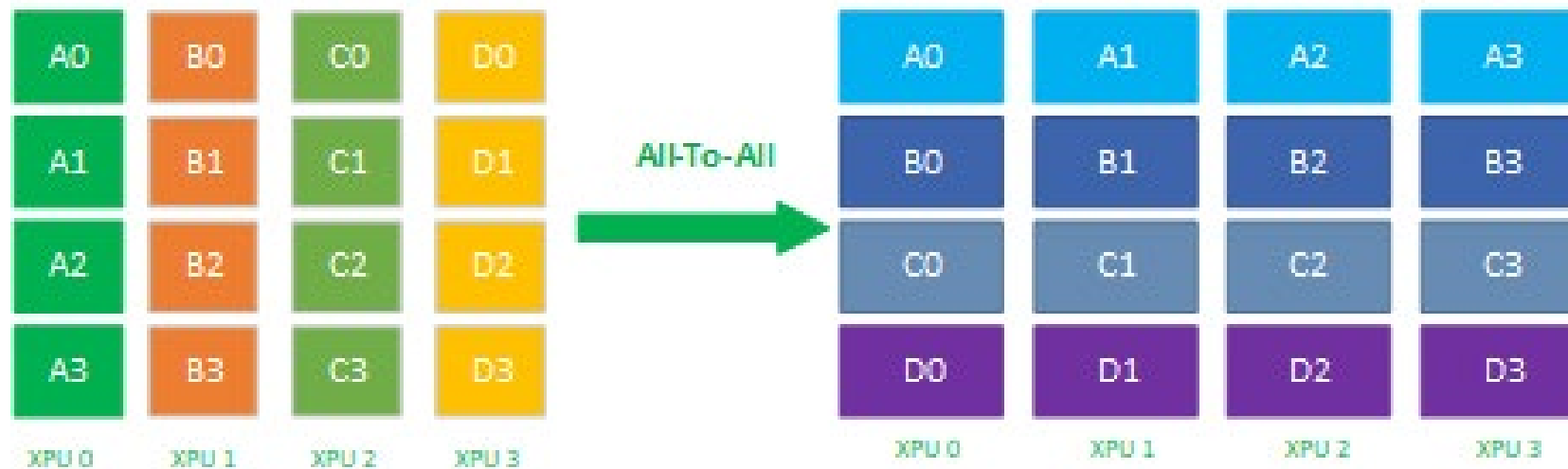
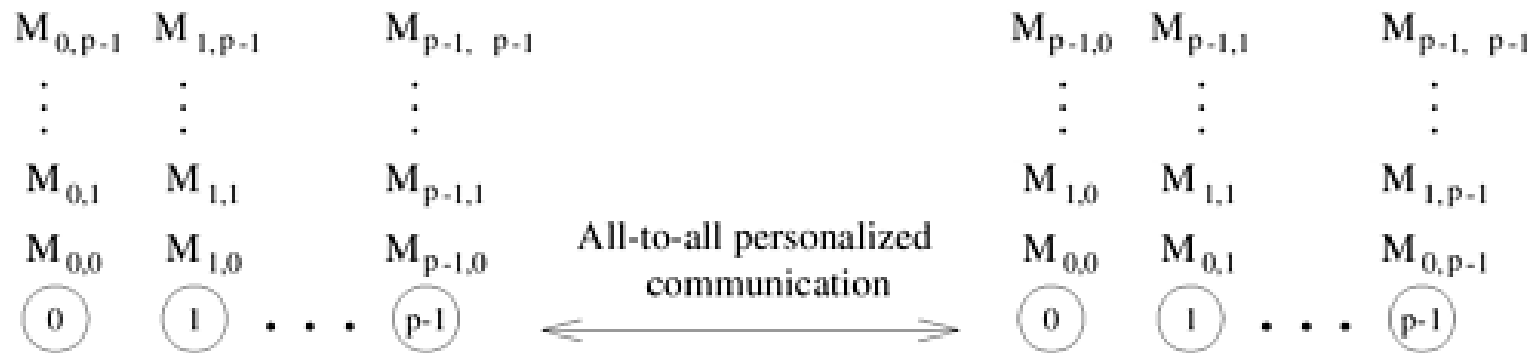
$$T_2 = (t_s + m \frac{p}{2^2} t_w)$$

Step 3

$$T_3 = (t_s + m \frac{p}{2^3} t_w)$$

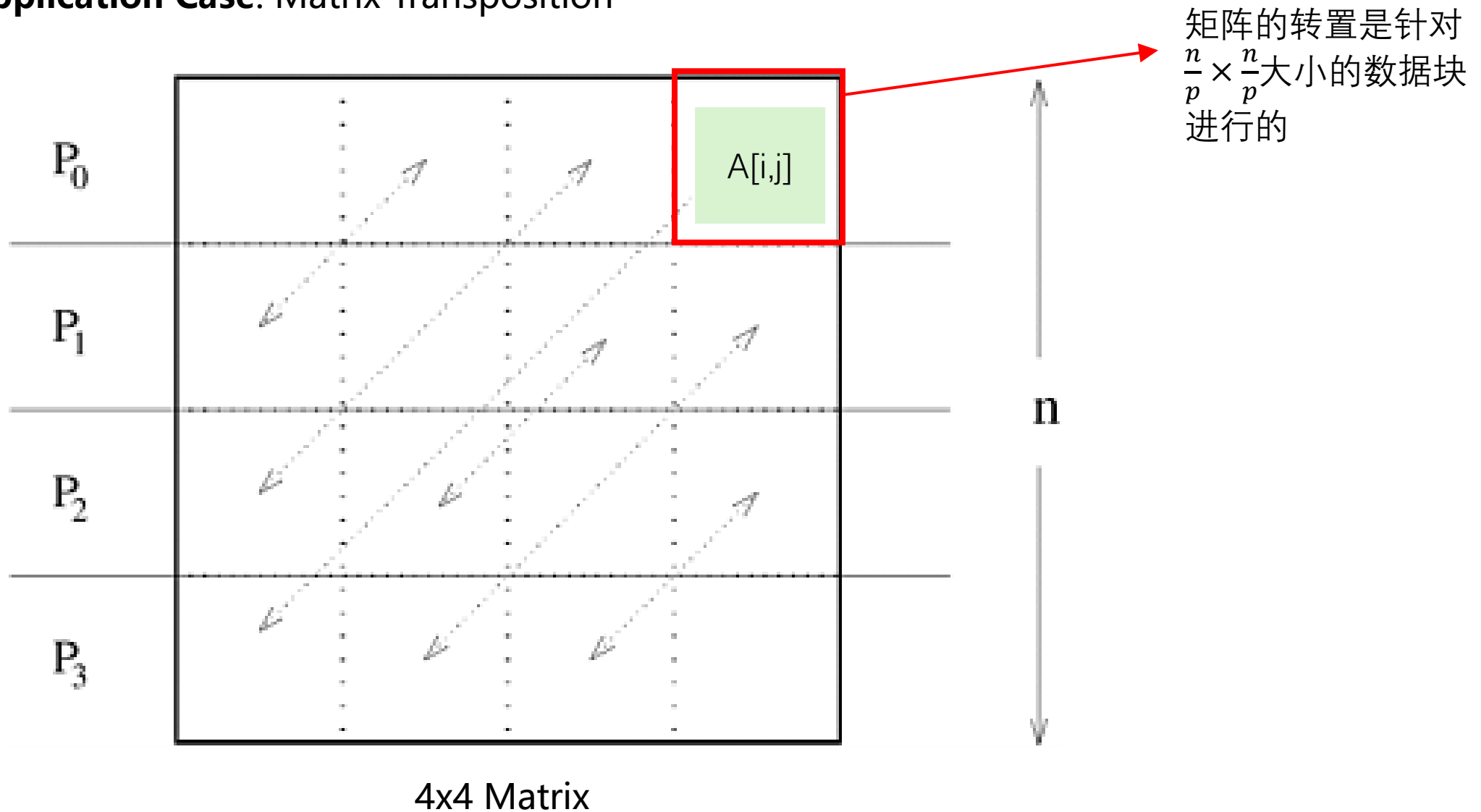
$$T = t_s \log p + m t_w (p - 1)$$

4.5 All-to-All Personalized Communication (Total Exchange)



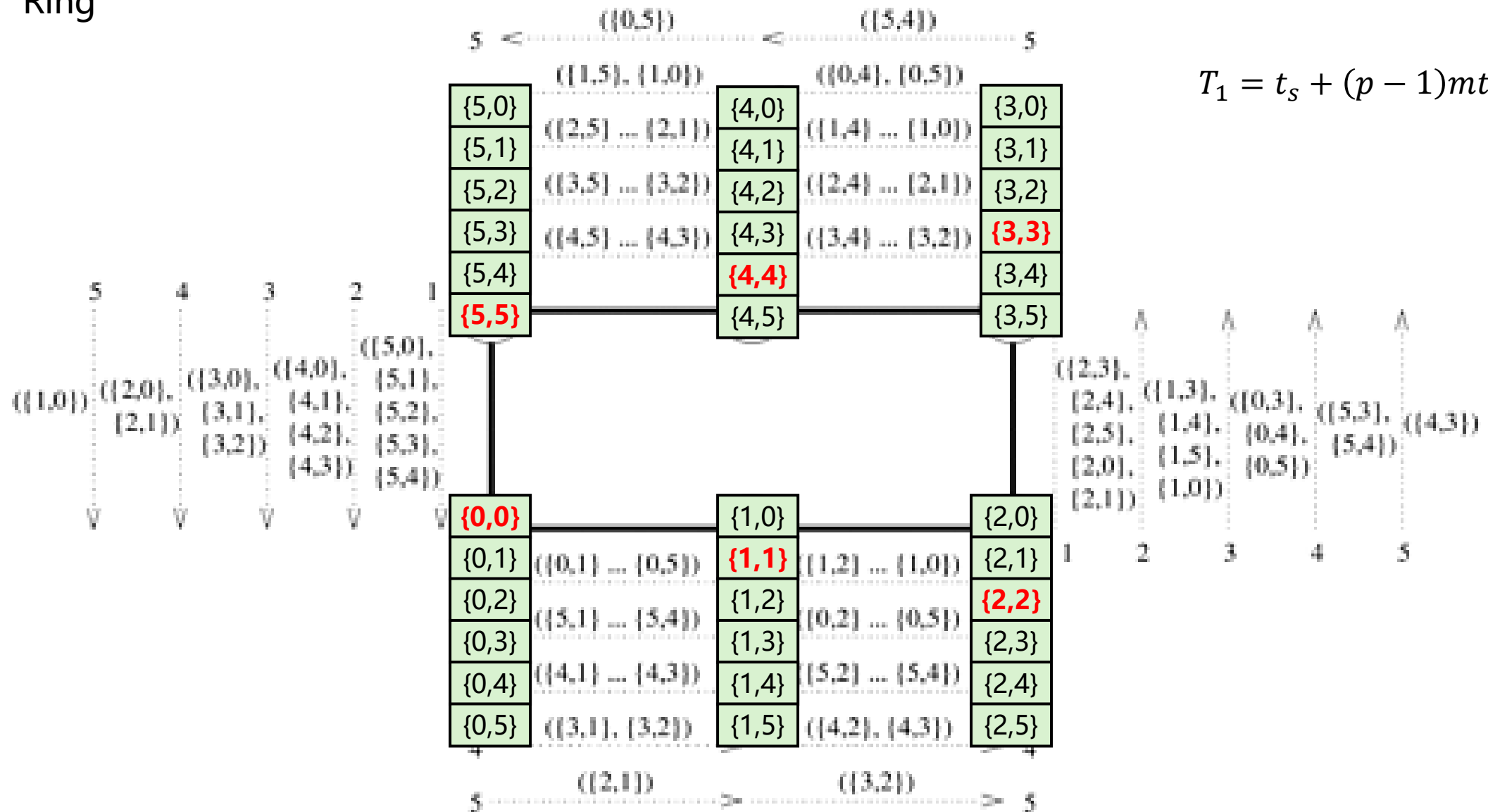
4.5 All-to-All Personalized Communication (Total Exchange)

Application Case: Matrix Transposition



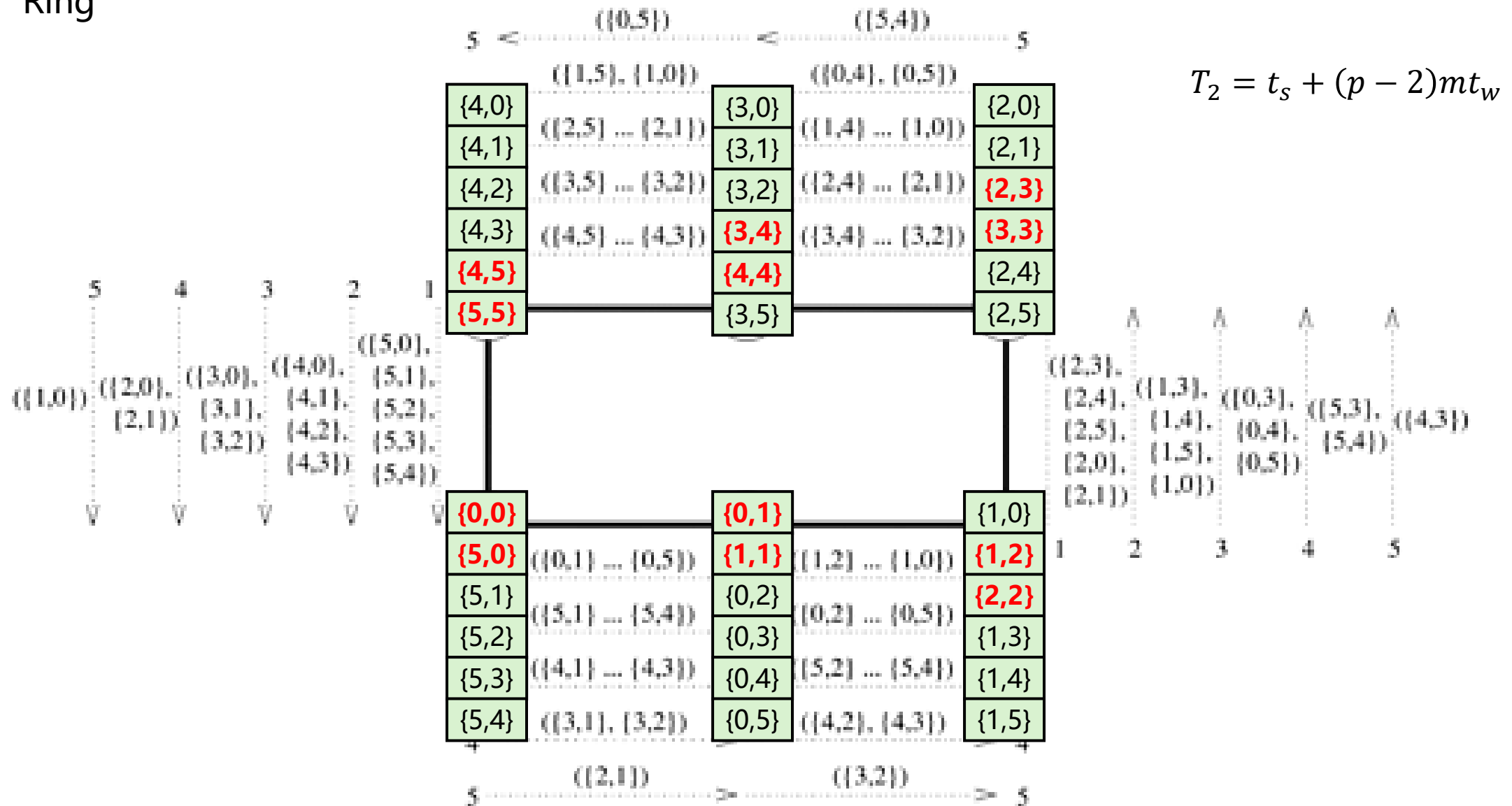
4.5 All-to-All Personalized Communication (Total Exchange)

Ring



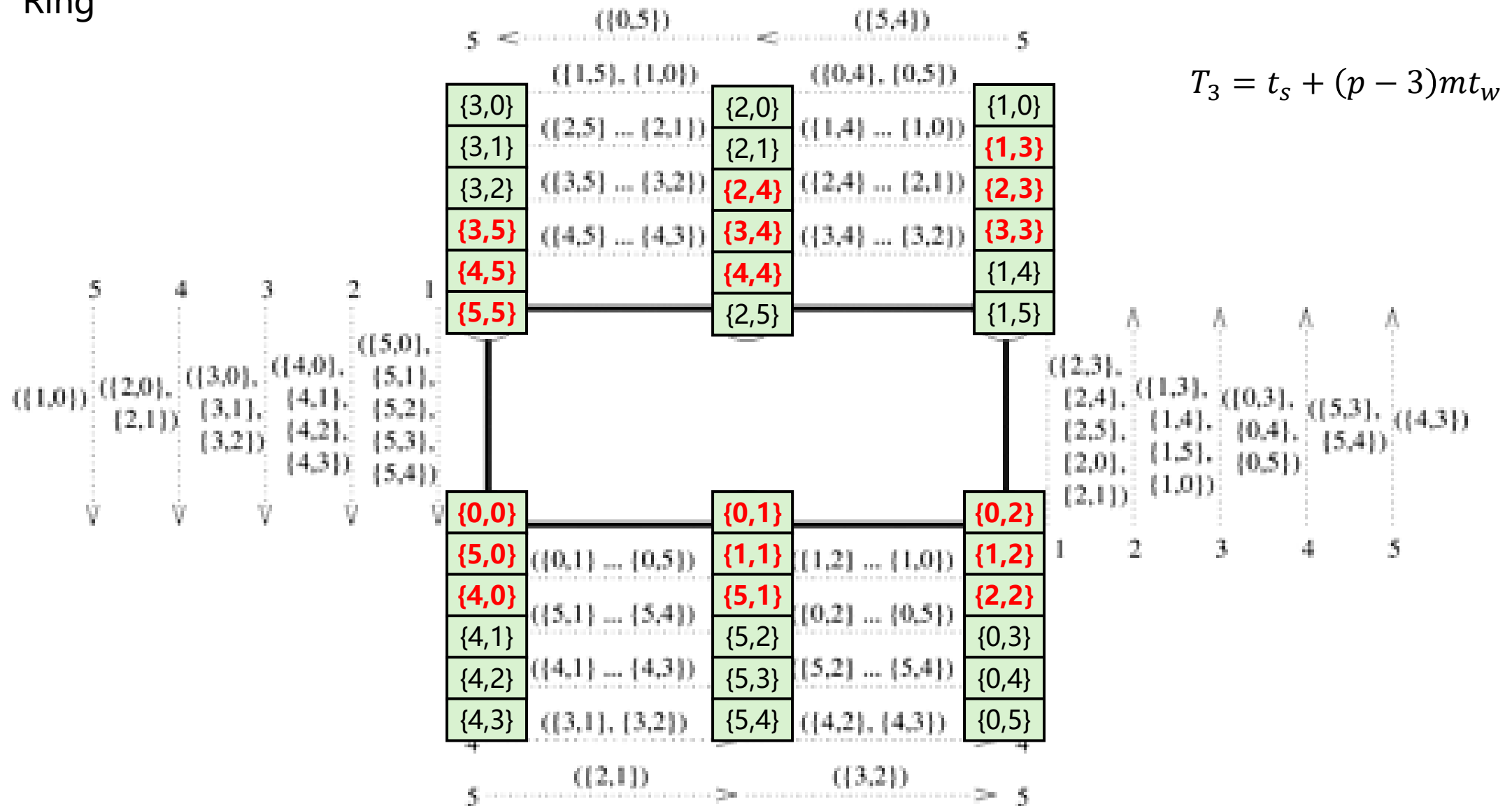
4.5 All-to-All Personalized Communication (Total Exchange)

Ring



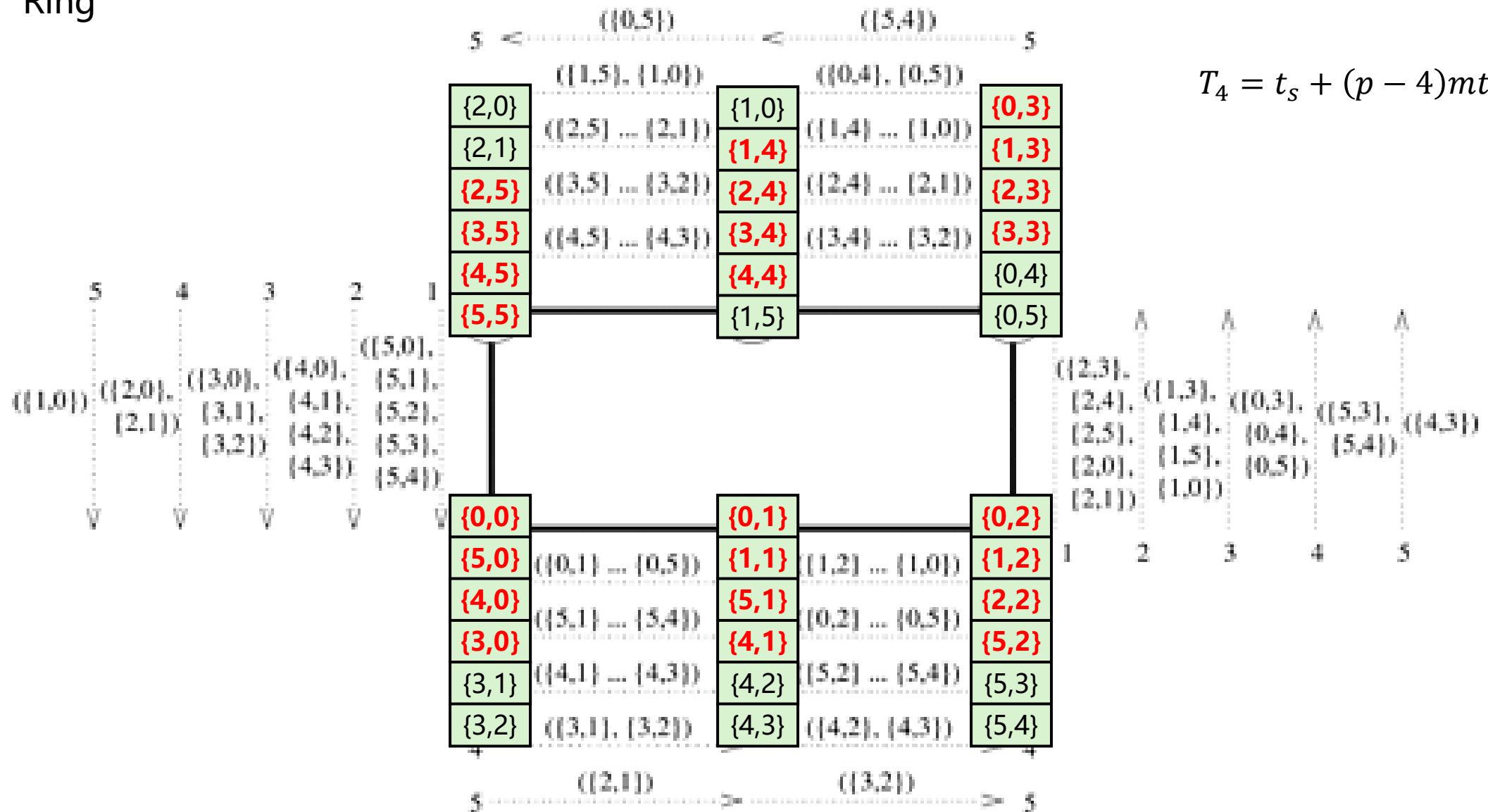
4.5 All-to-All Personalized Communication (Total Exchange)

Ring



4.5 All-to-All Personalized Communication (Total Exchange)

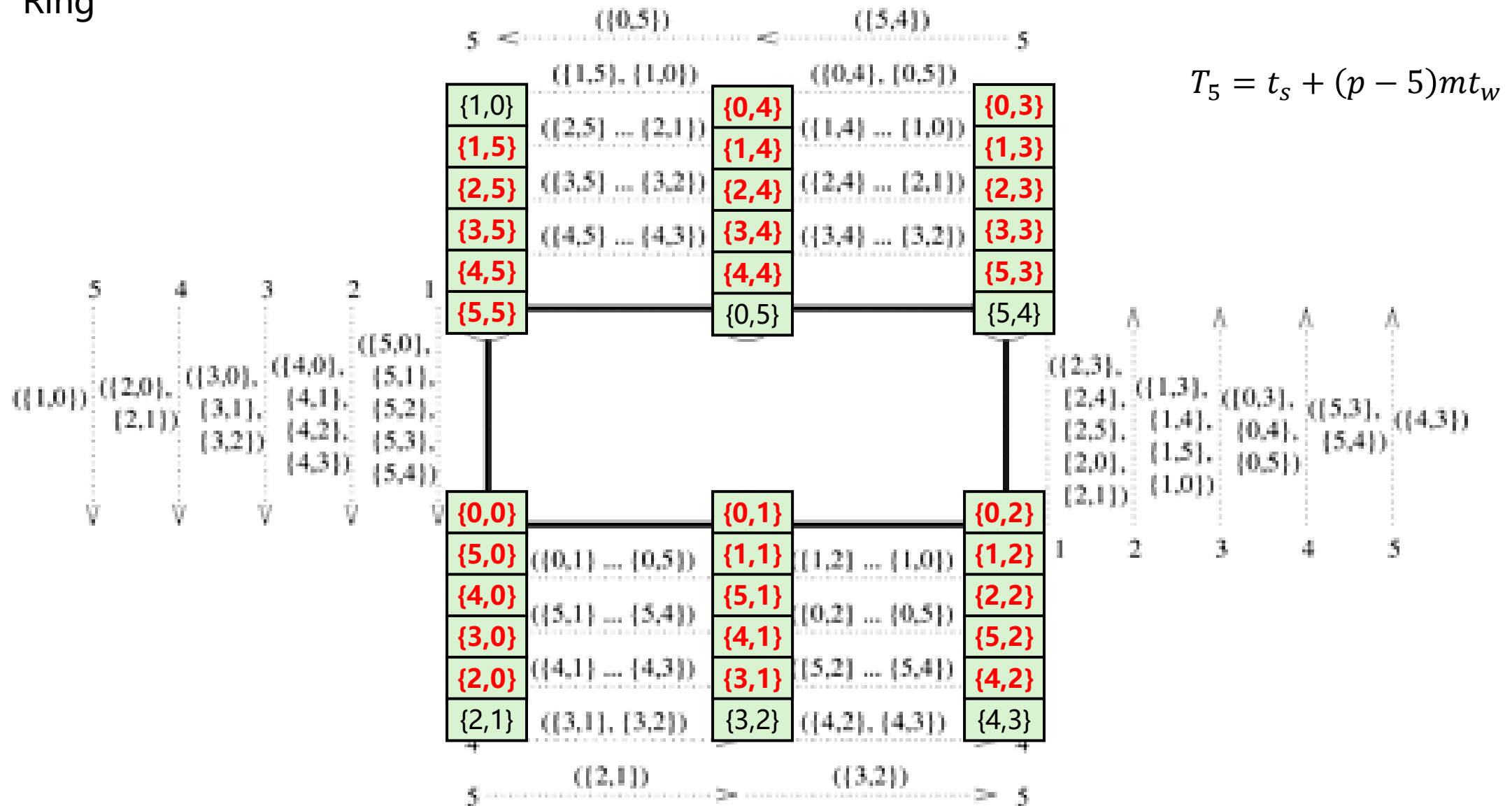
Ring



$$T_4 = t_s + (p - 4)mt_w$$

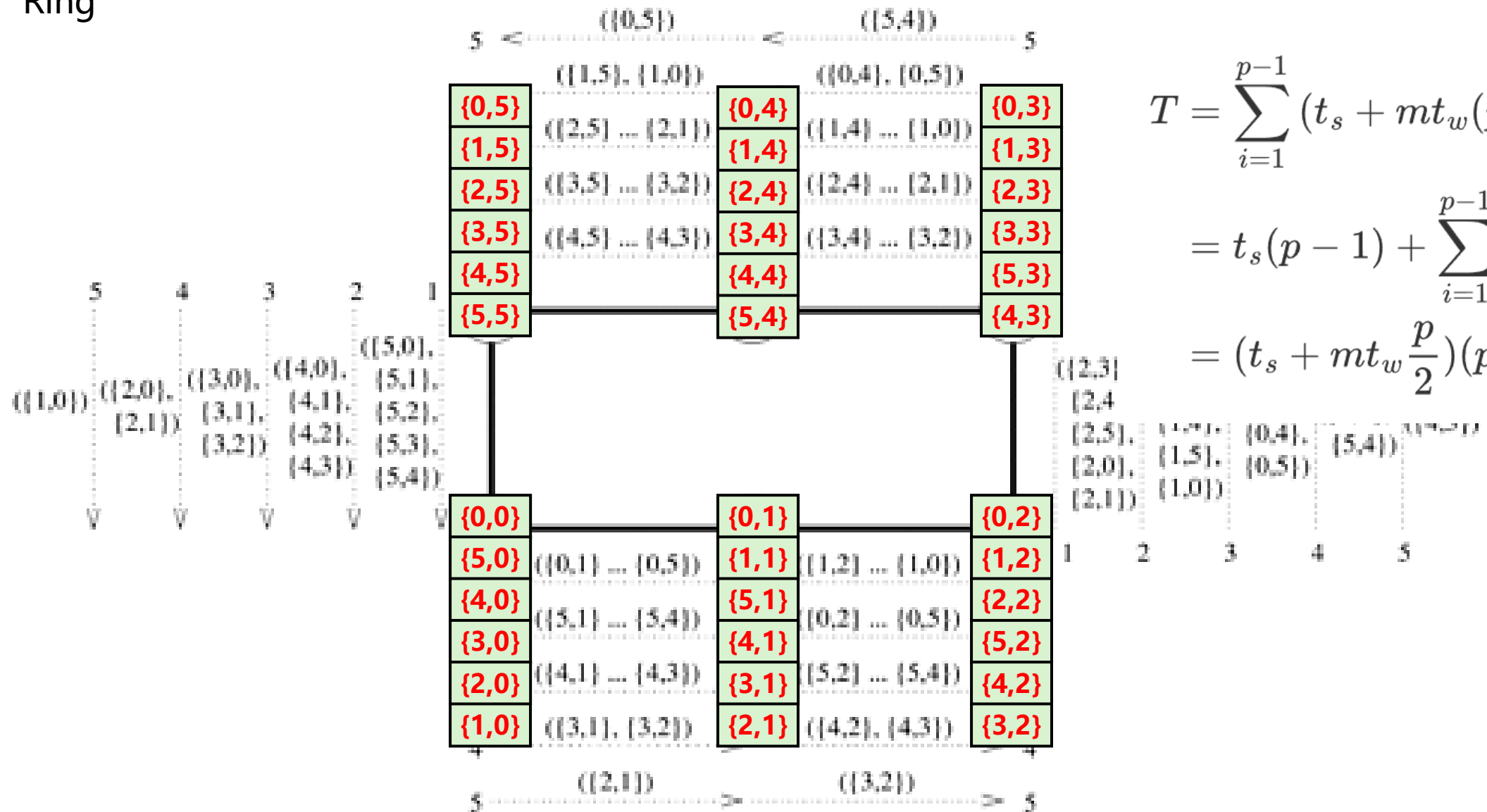
4.5 All-to-All Personalized Communication (Total Exchange)

Ring



4.5 All-to-All Personalized Communication (Total Exchange)

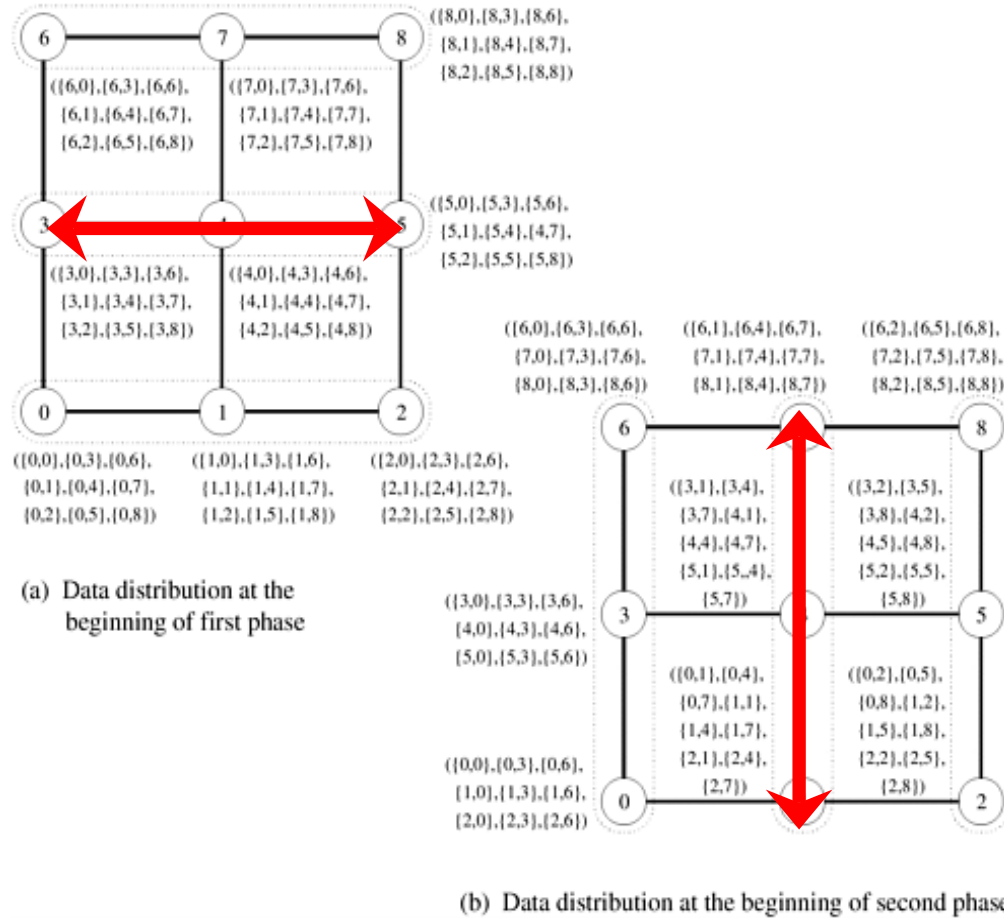
Ring



$$\begin{aligned}
 T &= \sum_{i=1}^{p-1} (t_s + mt_w(p - i)) \\
 &= t_s(p - 1) + \sum_{i=1}^{p-1} imt_w \\
 &= (t_s + mt_w \frac{p}{2})(p - 1)
 \end{aligned}$$

4.5 All-to-All Personalized Communication (Total Exchange)

Mesh



Ring

$$T = \sum_{i=1}^{p-1} (t_s + mt_w(p - i))$$

$$= t_s(p - 1) + \sum_{i=1}^{p-1} imt_w$$

$$= (t_s + mt_w \frac{p}{2})(p - 1)$$



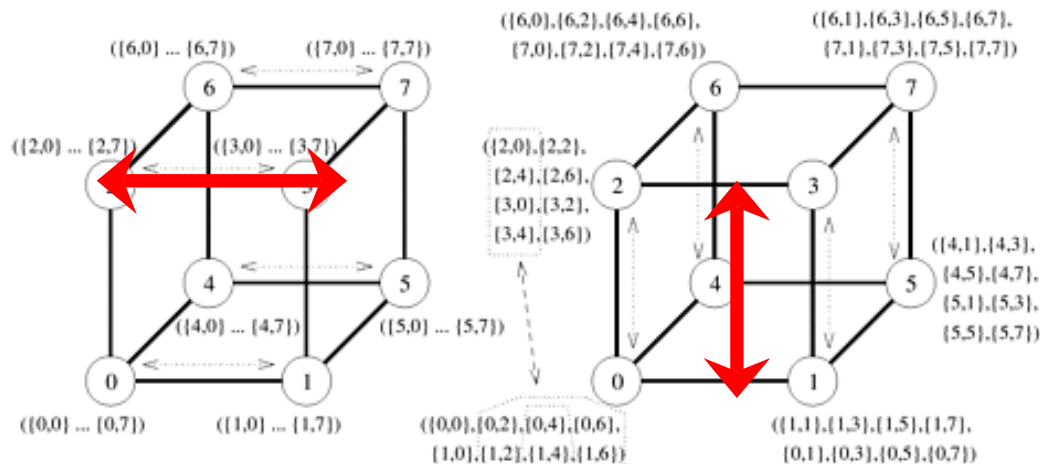
$2 \times \text{Ring}$ and $p \rightarrow \sqrt{p}$

Mesh

$$T = (2t_s + mt_w p)(\sqrt{p} - 1)$$

4.5 All-to-All Personalized Communication (Total Exchange)

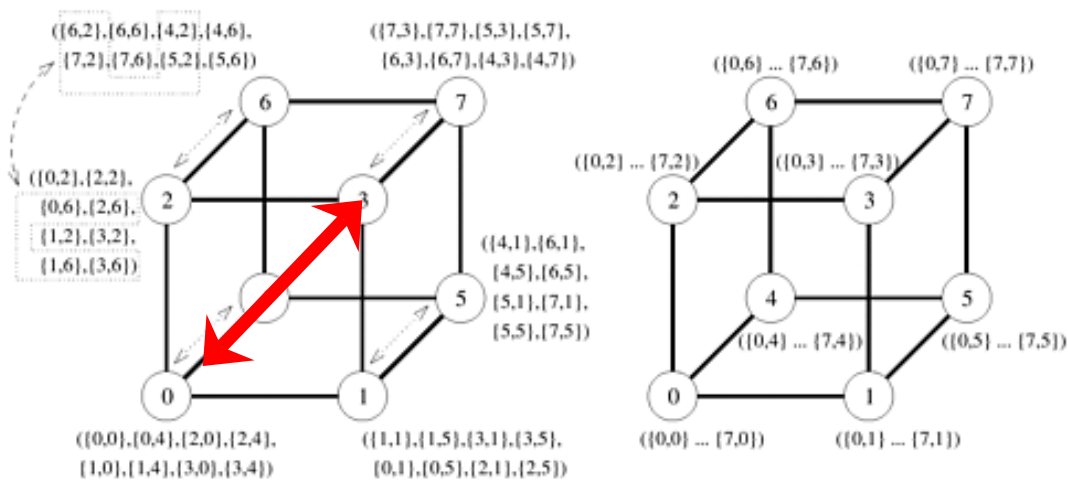
Hypercube



每步需要传输的数据量为 $\frac{mp}{2}$
 共计需要传输 $\log p$ 步

$$T = \left(t_s + \frac{mpt_w}{2} \right) \log p$$

未考虑传输数据后的排序和索引的时间



问题

- 以上All-to-All算法在Hypercube是最优的吗？
 - 一共进行 $\log p$ 步
 - 每步传输数据大小为 $mp/2$

4.5 All-to-All Personalized Communication (Total Exchange)

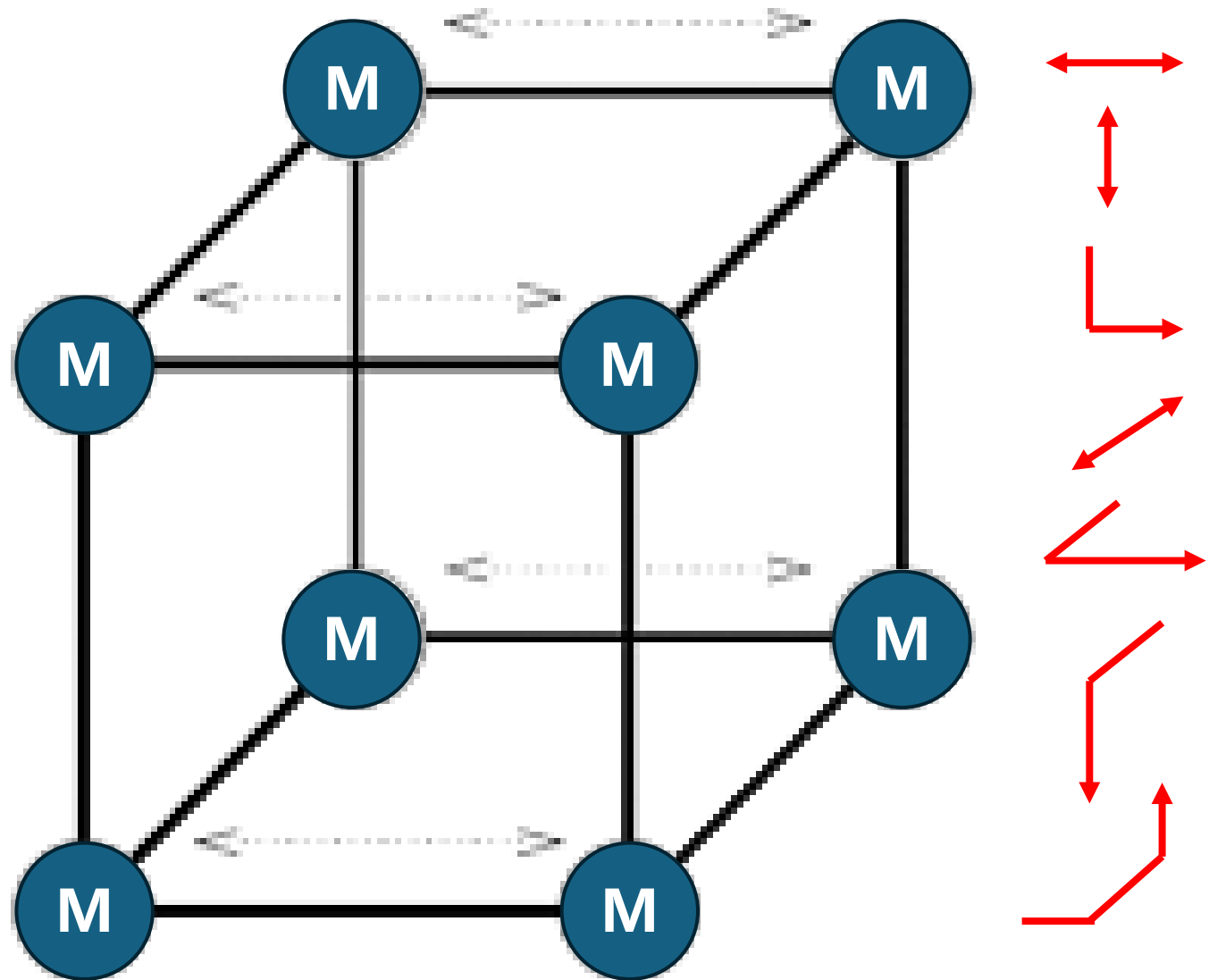
Hypercube - **An Optimal Algorithm**

All-to-All:

1. 每个节点与其他的 $p-1$ 个节点通信, 通信的大小为 m 个字节
2. 通信过程中没有阻塞

$$\begin{aligned} T &= (t_s + mt_w) \\ &+ (t_s + mt_w) + (t_s + mt_w) \\ &+ (t_s + mt_w) + (t_s + mt_w) \\ &+ (t_s + mt_w) + (t_s + mt_w) \end{aligned}$$

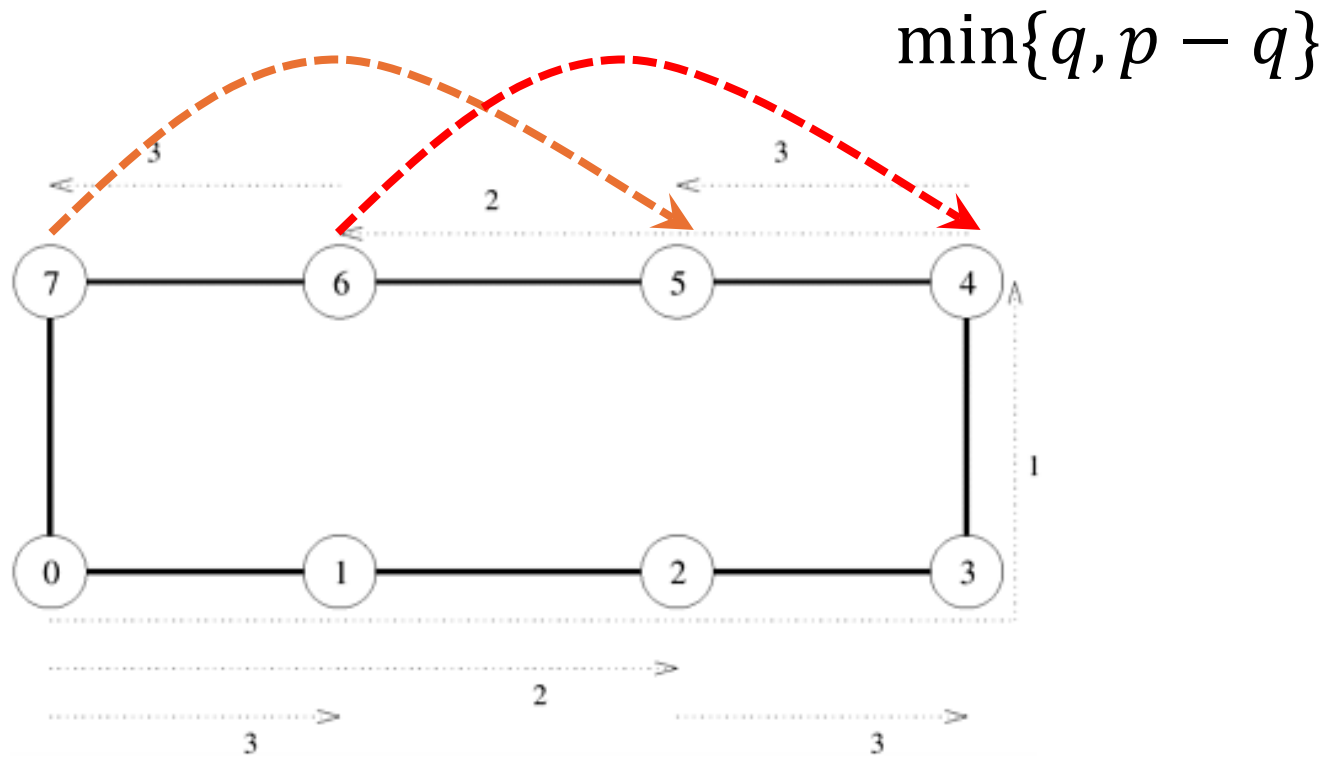
$$T = (t_s + t_w m)(p - 1)$$



4.6 Circular Shift

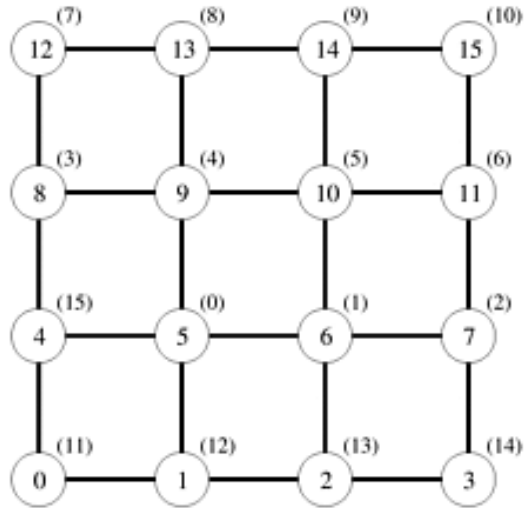
We define a **circular q -shift** as the operation in which node i sends a data packet to node $(i + q) \bmod p$ in a p -node ensemble

Ring



4.6 Circular Shift

Mesh - 4x4 Mesh 5-shift



循环移位的结果:

初始状态 $\langle 0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15 \rangle$

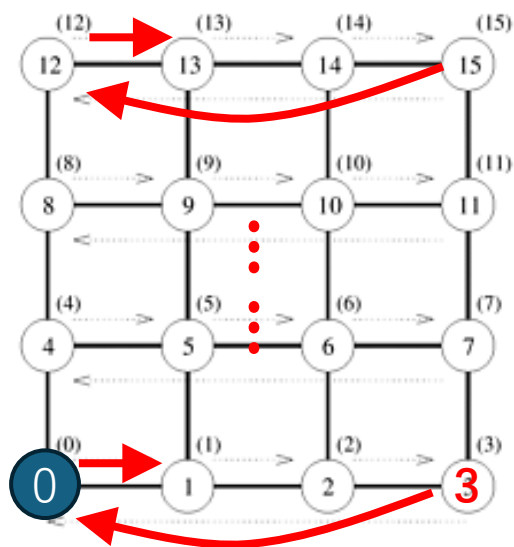
5-Shift $\langle 11, 12, 13, 14, 15, 0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10 \rangle$



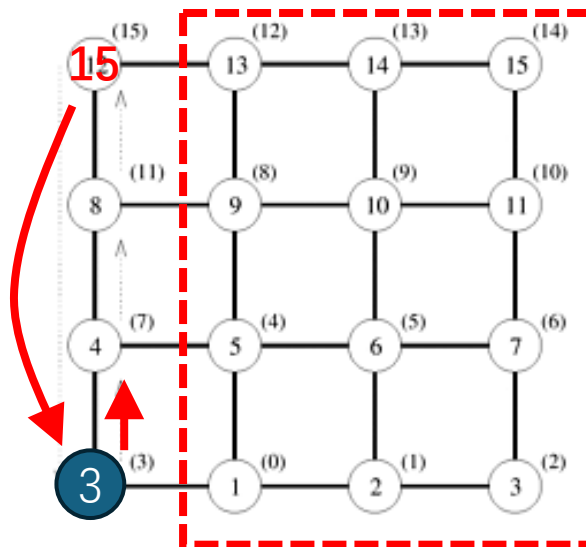
在Mesh上如何实现?

4.6 Circular Shift

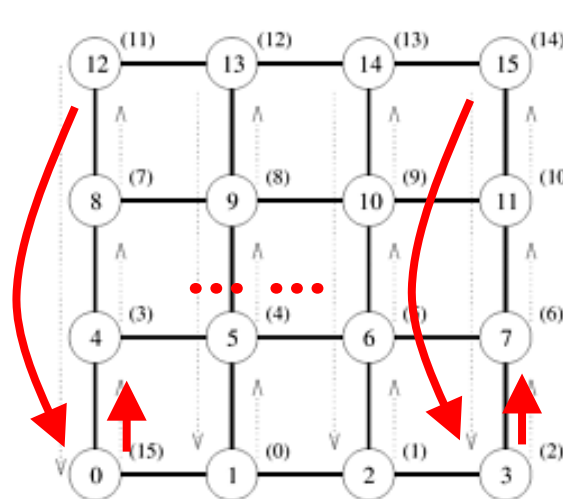
Mesh - 4x4 Mesh 5-shift



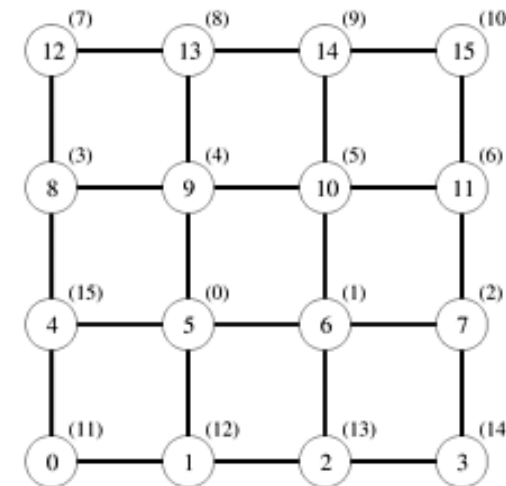
(a) Initial data distribution and the first communication step



(b) Step to compensate for backward row shifts



(c) Column shifts in the third communication step



(d) Final distribution of the data

→ 沿着行方向移动 $q \bmod \sqrt{p}$ 步

进位补偿

↑ 沿着列方向移动 $\lfloor \frac{q}{\sqrt{p}} \rfloor$ 步

During the circular row shifts, some of the data traverse the wraparound connection from the highest to the lowest labeled nodes of the rows.

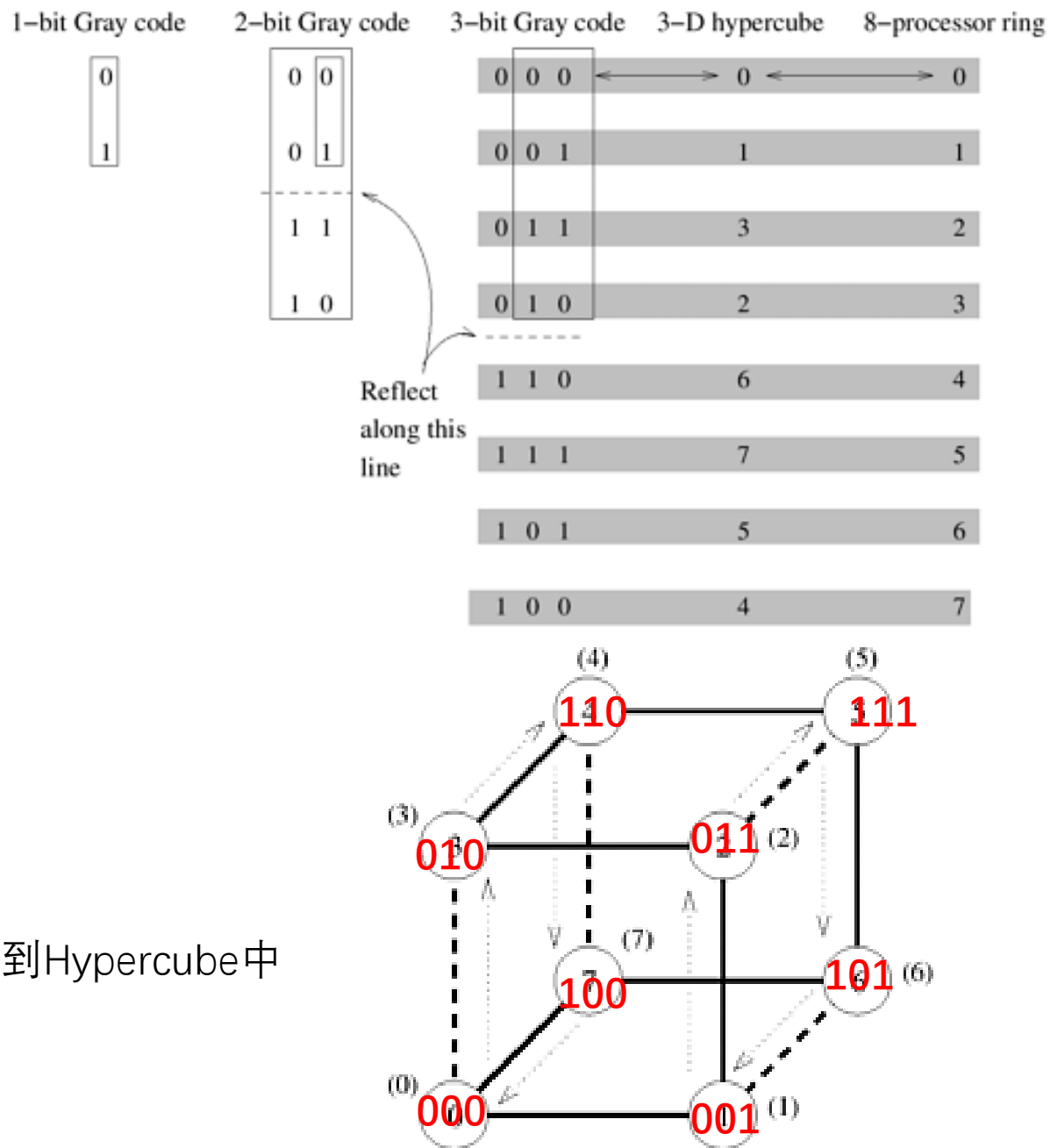
4.6 Circular Shift

Hypercube - 2^3 Hypercube 5-Shift

初始状态<0, 1, 2, 3, 4, 5, 6, 7>

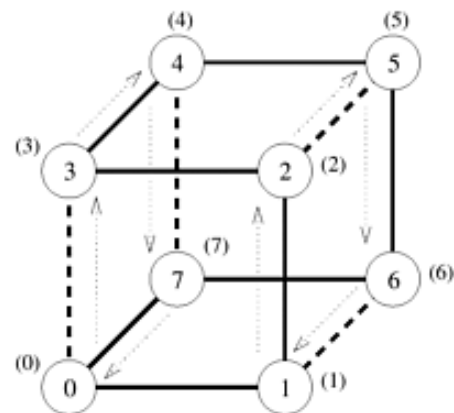
5-Shift <3, 4, 5, 6, 7, 0, 1, 2>

将Linear的序号使用二进制反射格雷码映射到Hypercube中



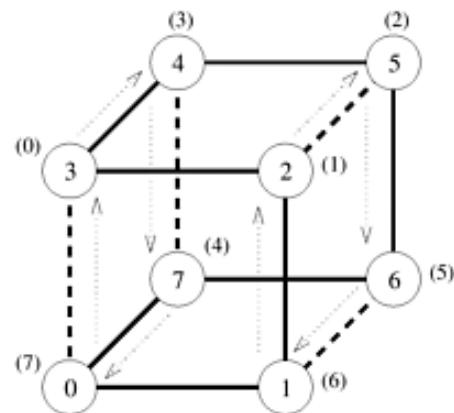
4.6 Circular Shift

Hypercube - 2³Hypercube 5-Shift

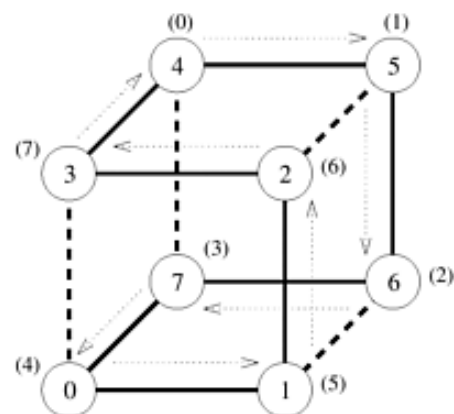


First communication step of the 4-shift

(a) The first phase (a 4-shift)



Second communication step of the 4-shift



(b) The second phase (a 1-shift)

(c) Final data distribution after the 5-shift

2-Shift → 0b010 → 分为1个阶段进行

5-Shift → 0b101 → 分为2个阶段进行 ←

7-Shift → 0b111 → 分为3个阶段进行

通信过程

= 2²移位 + 2⁰移位

= 4移位 + 1移位

= (2次通信) + 1次通信 ← 除了1移位都是两次通信

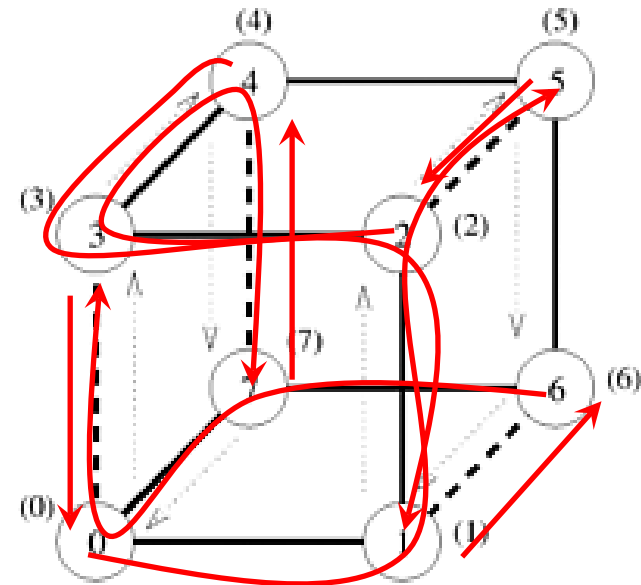
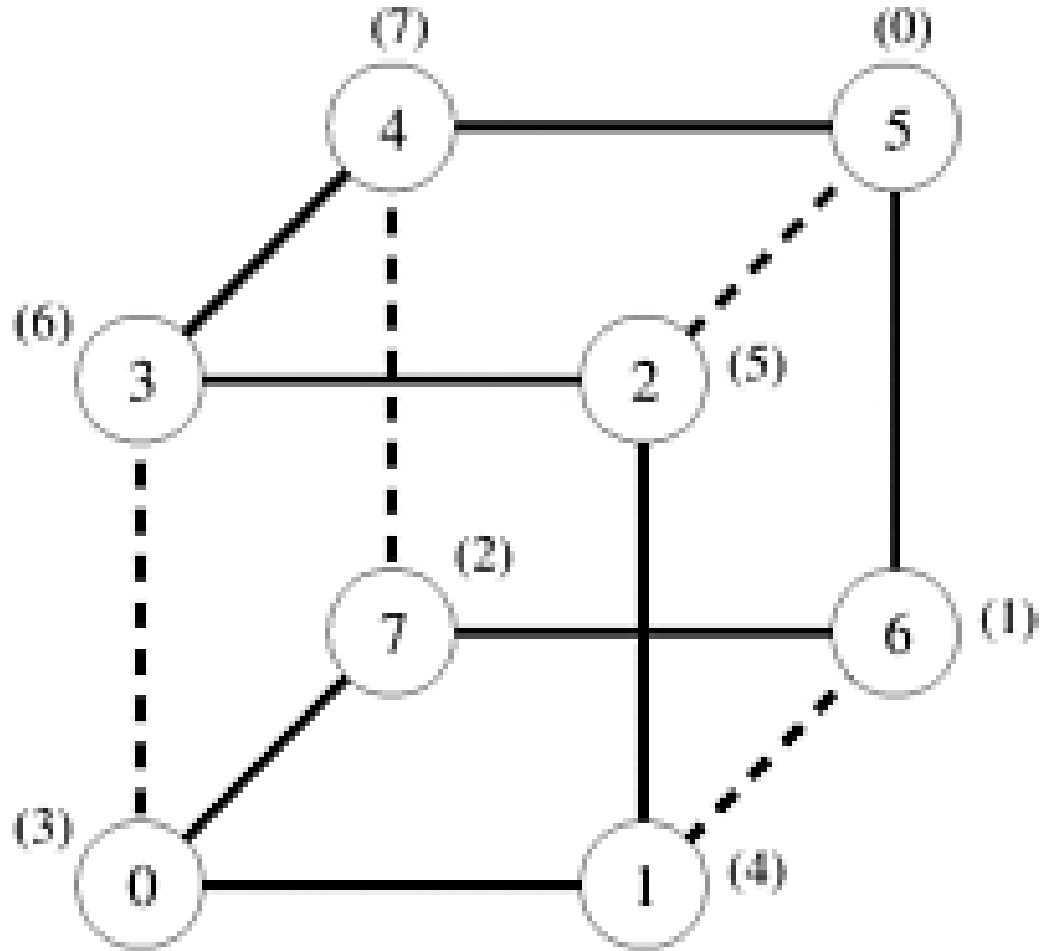
所有间隔为2⁰的节点构成一个子矩阵
循环一次

所有间隔为2²的节点构成一个子矩阵
循环两次

Ring

4.6 Circular Shift

Hypercube - 2^3 Hypercube 5-Shift



4.7 Improving the Speed of Some Communication Operations

1. Splitting and Routing Messages in Parts

- Note that the algorithms of this section rely on **m being large enough** to be split into p roughly equal parts.
- there is **a cut-off value for the message size m** and only the messages longer than the cut-off would benefit from the algorithms in this section.

2. All-Port Communication

- an ***all-port communication*** model permits simultaneous communication on all the channels connected to a node.

4.7 Improving the Speed of Some Communication Operations

Splitting and Routing Messages in Parts

Hypercube

One-to-All Broadcast and All-to-One Reduction

One-to-All Broadcast (m) = Scatter (m/p) + All-to-All Broadcast (m/p)

\rightarrow Hypercube $\{T\} = t_s \log p + \frac{m}{p} t_w (p - 1)$ *Slide Page 14*

\rightarrow Hpyercube $\{T\} = t_s \log p + \frac{m}{p} t_w (p - 1)$ *Slide Page 20*

$T = 2 \times (t_s \log p + t_w (p - 1) \frac{m}{p})$
 $\approx 2 \times (t_s \log p + \underline{mt_w})$

vs.

$T = (t_s + \underline{mt_w}) \log(p)$ *Slide Page 6*

$2mt_w$

$\log p \, mt_w$

4.7 Improving the Speed of Some Communication Operations

Splitting and Routing Messages in Parts

Hypercube

All Reduce

All Reduce (m) = All-to-All Reduction (m/p) + All-to-All Broadcast (m/p)

Hypercube $\{T\} = t_s \log p + \frac{m}{p} t_w (p - 1)$ Slide Page 14

Hypercube $\{T\} = t_s \log p + \frac{m}{p} t_w (p - 1)$ Slide Page 14

$$T = 2 \times (t_s \log p + t_w (p - 1) \frac{m}{p})$$
$$\approx 2 \times (t_s \log p + \underline{mt_w})$$

vs. $T = (t_s + \underline{mt_w}) \log p$ Slide Page 16

$2mt_w$

$\log p mt_w$

4.8 Summary

各种操作在超立方互连网络上的通信时间汇总

Operation	Hypercube Time	B/W Requirement
One-to-All Broadcast / All-to-One Reduction	$\min((t_s + mt_w) \log p, 2(t_s \log p + mt_w))$	$\Theta(1)$
All-to-All Broadcast / All-to-All Reduction	$t_s \log p + t_w m(p - 1)$	$\Theta(1)$
All Reduce	$\min((t_s + mt_w) \log p, 2(t_s \log p + t_w m))$	$\Theta(1)$
Scatter / Gather	$t_s \log p + t_w m(p - 1)$	$\Theta(1)$
All-to-All Personalized	$(t_s + t_w m)(p - 1)$	$\Theta(p)$
Circular Shift	$t_s + t_w m$	$\Theta(p)$

4.1 One-to-All Broadcast and All-to-One Reduction

3.1.2 任务交互 Task-Interaction

总结

问题

提示