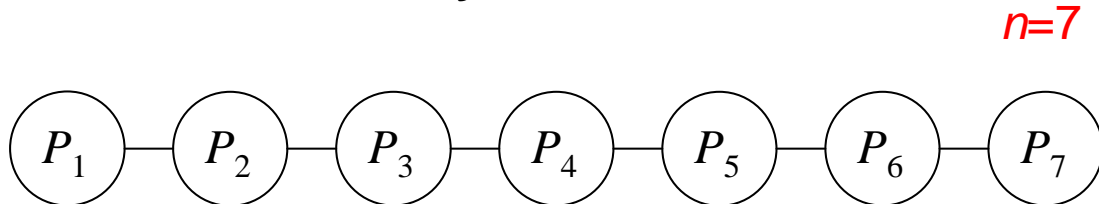


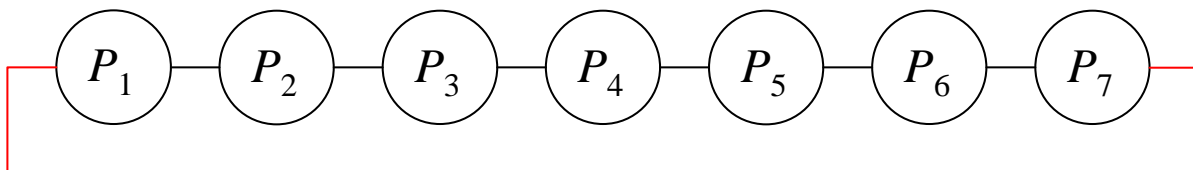
Linear Processor Arrays

■ Linear Processor Array



- (1) Diameter = $O(n)$ $n-1$
- (2) Maximum degree = 2 interconnection network
- (3) Link complexity = $O(n)$ $n-1$

■ Ring (a linear processor array with end-around connection)



■ Problems

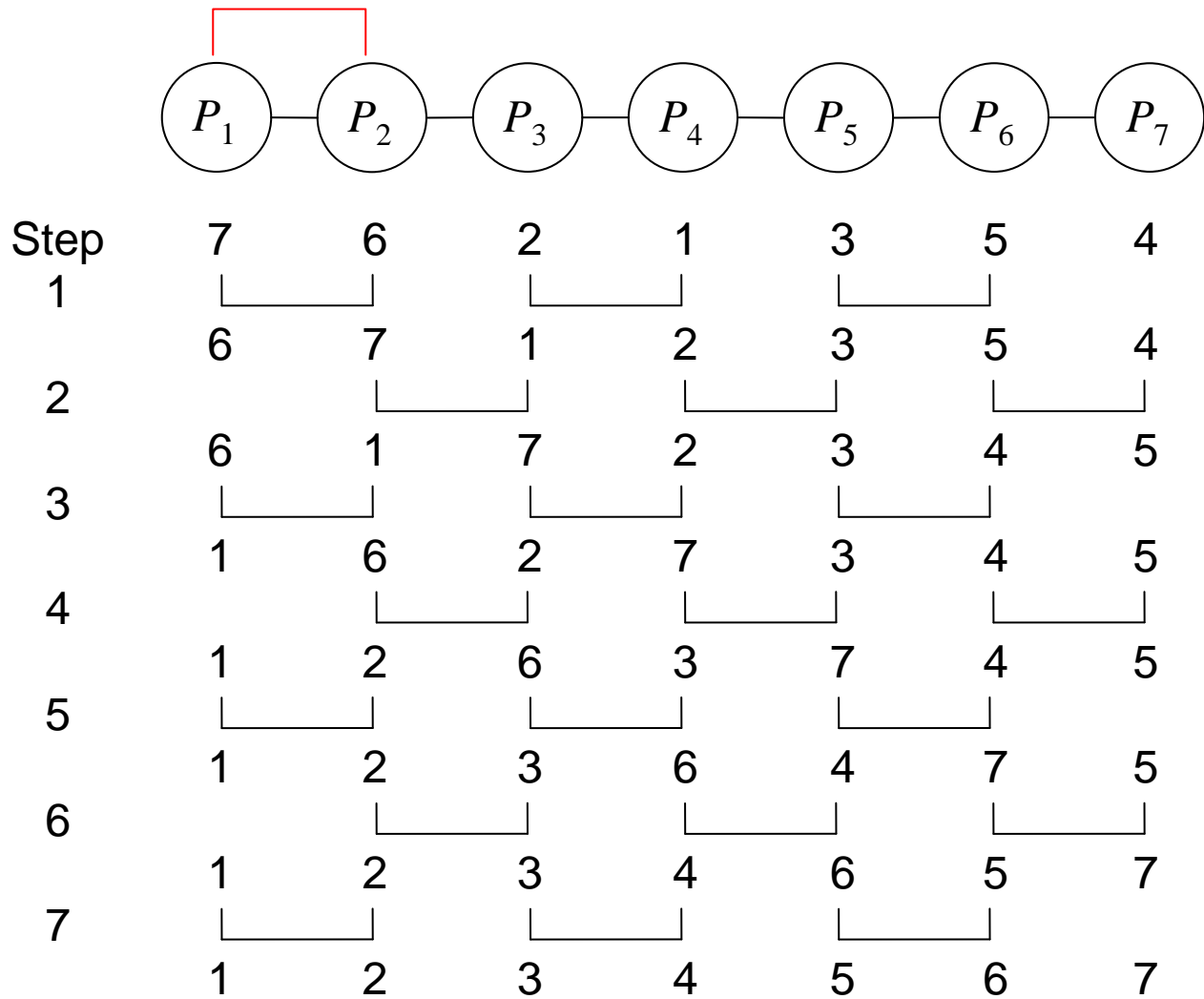
Problem LPA-1: Sorting ([Odd-Even Transposition Sort](#))

Input: $A[1 \dots n] = \{7, 6, 2, 1, 3, 5, 4\}$

Output: $A[1 \dots n] = \{1, 2, 3, 4, 5, 6, 7\}$

Model: LPA of size n . Initially, a_i is stored in P_i .

Compare & Exchange



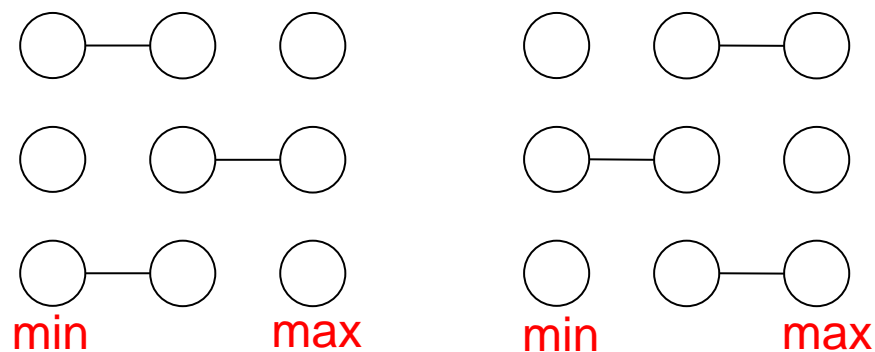
Theorem: Odd-Even transposition sort produces a sorted sequence of n data after n steps.

Proof: (by induction)

(1) $n=3$

case 1: odd-even

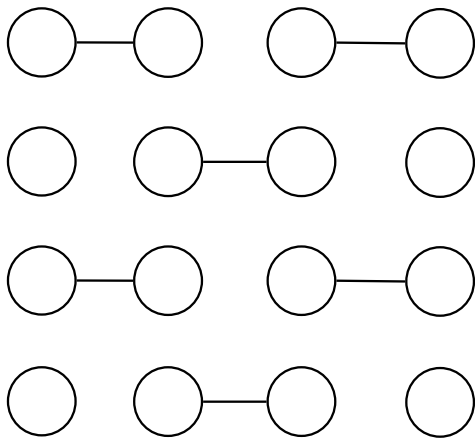
case 2: even-odd



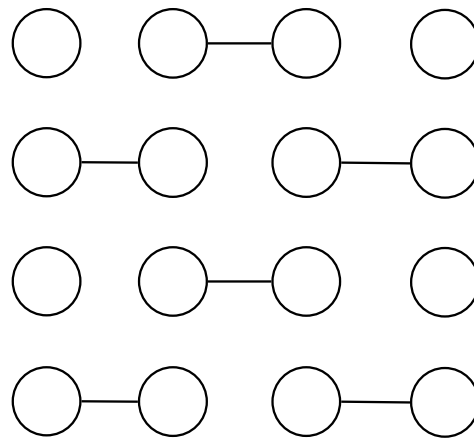
What is the sequential algorithm?
(Pipeline!)

(Assume $n = q \geq 3$ holds)

(2) $n=4$ case 1:odd-even



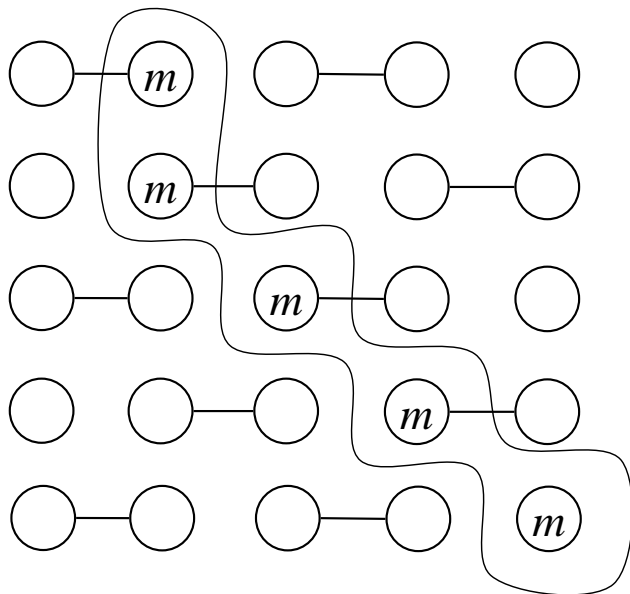
case 2:even-odd



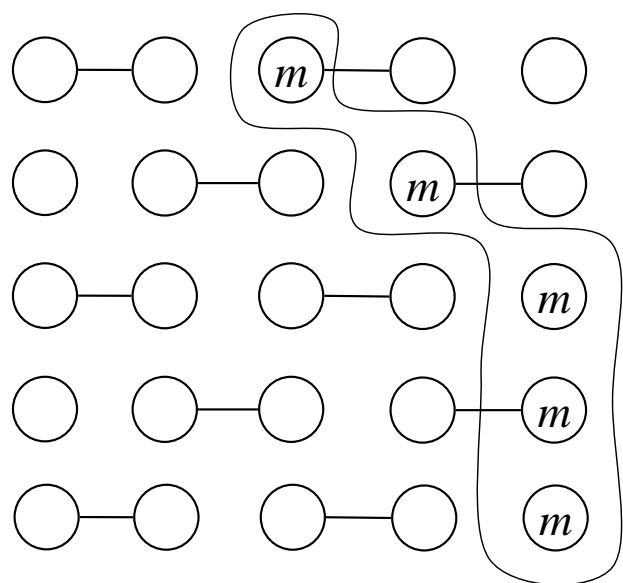
(Prove $n = q+1$ also holds)

(3) $n=5$ (Assume the maximum m is stored in P_k .)

case 1.1:odd-even, k is even

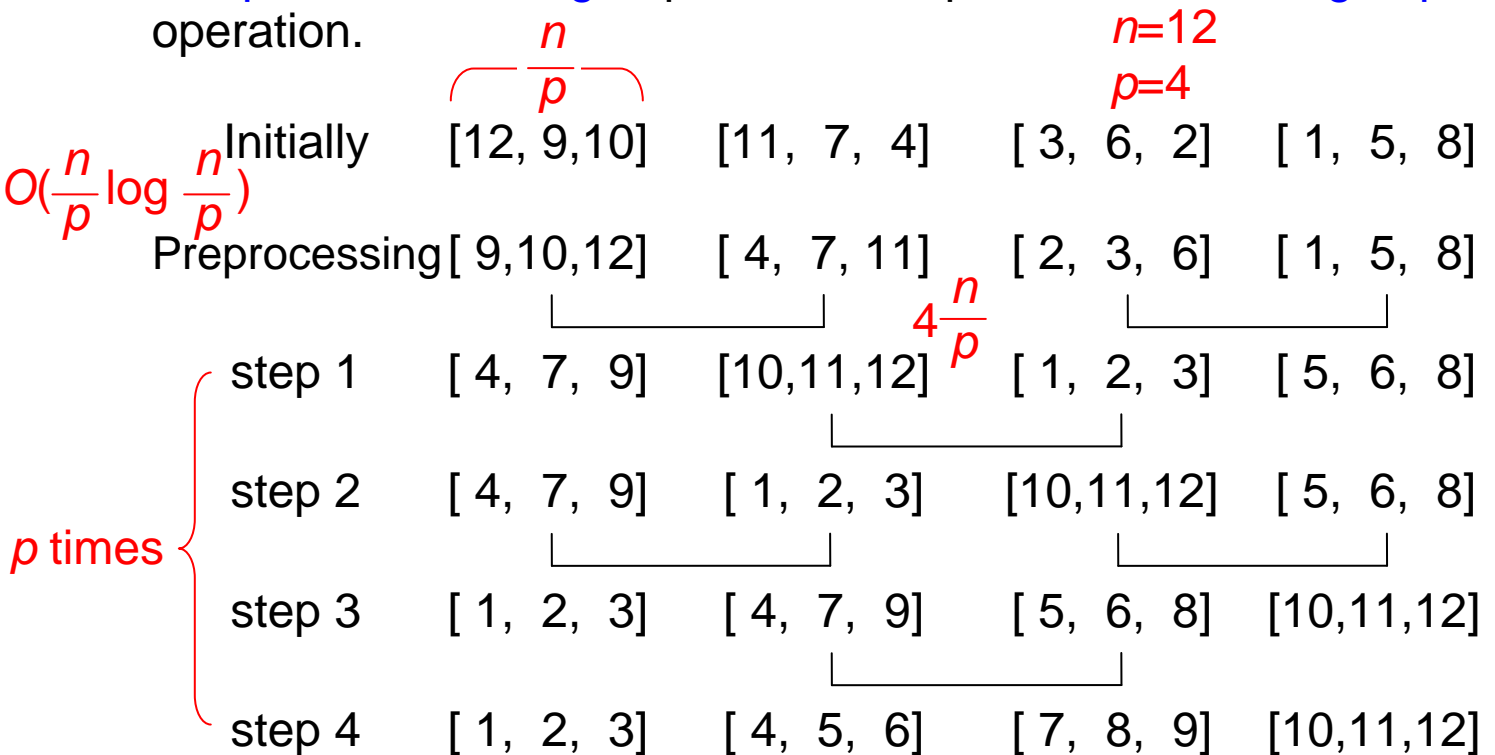


case 1.2:odd-even, k is odd



Problem LPA-2: Sorting (Merge-Splitting Sort)**Input:** $A[1\dots n]=\{12, 9, 10, 11, 7, 4, 3, 6, 2, 1, 8, 5\}$ **Output:** $A[1\dots n]=\{1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12\}$ **Model:** LPA of size p . Initially, $A[(i-1)*(n/p)+1 \dots i*(n/p)]$ are stored in P_i .(G. Baudet and D. Stevenson, "Optimal sorting algorithms for parallel computers," *IEEE Trans. Comput.* C-27 (1), 84-87.)

Merge-splitting sort is a generalization of odd-even transposition sort in which each processor holds a subsequence rather than a single data item, and the **comparison-exchange** operation is replaced with a **merge-split** operation.



$$* T(n) = O((n/p) \log (n/p)) + p * O(n/p) = O((n \log n)/p + n)$$

* For $p \leq \log n$, the proposed algorithm is cost-optimal.

$$\begin{aligned} \text{cost} &= (n \log n/p + n) \times p \\ &= n \log n + pn \end{aligned}$$

$\left\{ \begin{array}{l} 0/1 \\ \text{integer} \end{array} \right.$

Problem LPA-3: Integer Knapsack Problem

Input: $C, P[1 \dots n]$, and $W[1 \dots n]$ (capacity, profit, and weight)
 (C , and w_i are integers)

Output: MAX $\sum_{i=1}^n x_i p_i$

such that (1) $\sum_{i=1}^n x_i w_i \leq C$

(2) $x_i \geq 0$ are integers

Model: LPA of size n . Initially, p_i and w_i are stored in P_i , and C is stored in P_1 .

(1) Sequential Approach (dynamic programming)

Let $f(g, k)$ be the maximal value of the objective function using only the first k items with capacity limitation g . Clearly, $f(C, n)$ is the answer, and

$$f(g, k) = \text{Max} \left\{ \begin{array}{l} \text{not to take } k \\ f(g, k-1), \end{array} \begin{array}{l} \text{take } k \\ p_k + f(g-w_k, k) \end{array} \right\}. \quad (\text{if } g \geq w_k)$$

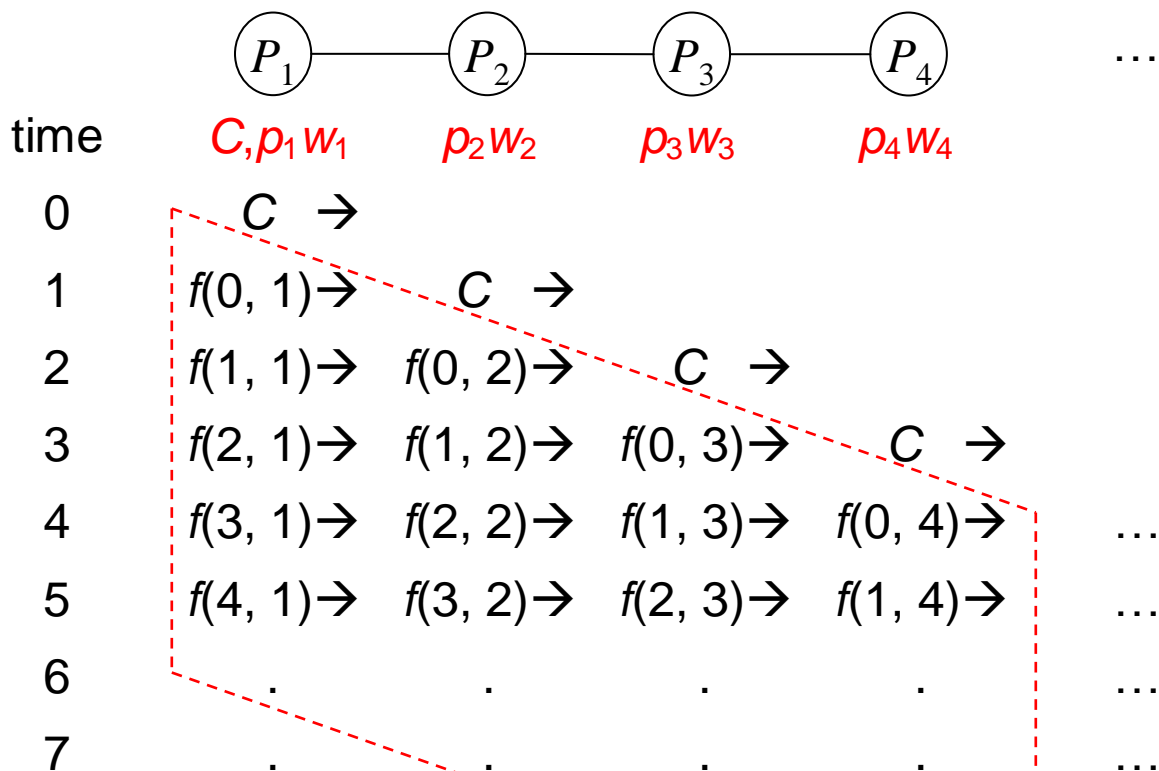
Iteration	$k=1$	$k=2$	$k=3$	$k=4$...
	$f(0,1)$	$f(0,2)$	$f(0,3)$	$f(0,4)$...
	$f(1,1)$	$f(1,2)$	$f(1,3)$	$f(1,4)$...
	$f(2,1)$	$f(2,2)$	$f(2,3)$	$f(2,4)$...
	$f(3,1)$	$f(3,2)$	$f(3,3)$	$f(3,4)$...
	$f(4,1)$	$f(4,2)$	$f(4,3)$	$f(4,4)$...

The algorithm

for $k=1$ to n
 for $g=0$ to C determine $f(g, k)$

$$\begin{array}{ccc}
 & & f(g-w_k, k) \\
 & & \downarrow \\
 f(g, k-1) & \longrightarrow & f(g, k)
 \end{array}$$

(2) Parallel Approach (pipelining)



* $T(n) = O(C + n - 1)$

*The 0/1 knapsack problem: $x_i = 0$ or 1 .
 For this problem, we have

$$f(g, k) = \mathbf{MAX}\{f(g, k-1), p_k + f(g-w_k, k-1)\}.$$

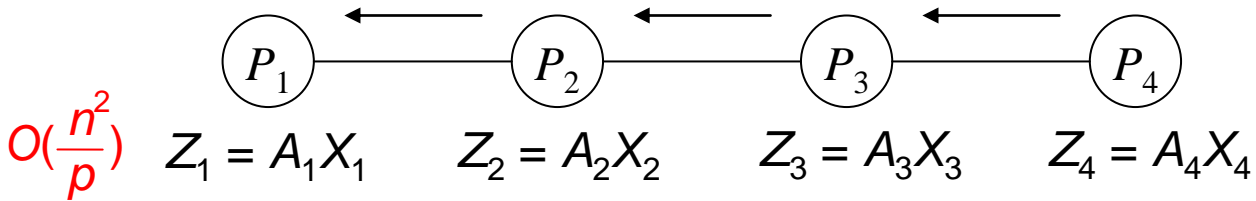
Problem LPA-4: Matrix Vector Product**Input:** a matrix $A[1 \dots n, 1 \dots n]$ and a vector $X[1 \dots n]$.**Output:** $Y[1 \dots n]=AX$ **Model:** LPA of size p . Let A and X be partitioned into p blocks as follows:

$$A=(A_1, A_2, \dots, A_p) \text{ and } X=(X_1, X_2, \dots, X_p),$$

where each A_i is of size $n \times r$ and each X_i is of size r . Initially, A_i and X_i are stored in P_i .

Example: ($n=6, p=2$) $\frac{6}{2}=3$ $[A_1, A_2] \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} = A_1 X_1 + A_2 X_2$

$$\begin{aligned}
 Y = AX &= \begin{bmatrix} a_{1,1} & a_{1,2} & a_{1,3} & a_{1,4} & a_{1,5} & a_{1,6} \\ a_{2,1} & a_{2,2} & a_{2,3} & a_{2,4} & a_{2,5} & a_{2,6} \\ a_{3,1} & a_{3,2} & a_{3,3} & a_{3,4} & a_{3,5} & a_{3,6} \\ a_{4,1} & a_{4,2} & a_{4,3} & a_{4,4} & a_{4,5} & a_{4,6} \\ a_{5,1} & a_{5,2} & a_{5,3} & a_{5,4} & a_{5,5} & a_{5,6} \\ a_{6,1} & a_{6,2} & a_{6,3} & a_{6,4} & a_{6,5} & a_{6,6} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \end{bmatrix} \\
 &= \begin{bmatrix} a_{1,1} & a_{1,2} & a_{1,3} \\ a_{2,1} & a_{2,2} & a_{2,3} \\ a_{3,1} & a_{3,2} & a_{3,3} \\ a_{4,1} & a_{4,2} & a_{4,3} \\ a_{5,1} & a_{5,2} & a_{5,3} \\ a_{6,1} & a_{6,2} & a_{6,3} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \begin{bmatrix} a_{1,4} & a_{1,5} & a_{1,6} \\ a_{2,4} & a_{2,5} & a_{2,6} \\ a_{3,4} & a_{3,5} & a_{3,6} \\ a_{4,4} & a_{4,5} & a_{4,6} \\ a_{5,4} & a_{5,5} & a_{5,6} \\ a_{6,4} & a_{6,5} & a_{6,6} \end{bmatrix} \begin{bmatrix} x_4 \\ x_5 \\ x_6 \end{bmatrix} \\
 &= \underbrace{A_1 X_1}_{Z_1} + \underbrace{A_2 X_2}_{Z_2}
 \end{aligned}$$

$(p=4)$ 

stage 1: Each processor P_i computes $Z_i = A_i X_i$. (Note that Z_i is a vector of order n .)

stage 2: for $i=p$ to 2

2.1 processor P_i sends Z_i to processor P_{i-1} , and then

2.2 processor P_{i-1} sets $Z_{i-1} = Z_{i-1} + Z_i$.

stage 3: Processor P_1 sets $Y = Z_1$.

partition

computation

$$* T(n) = O(n^2/p) + (p-1)O(n) = O(n^2/p + np)$$

communication

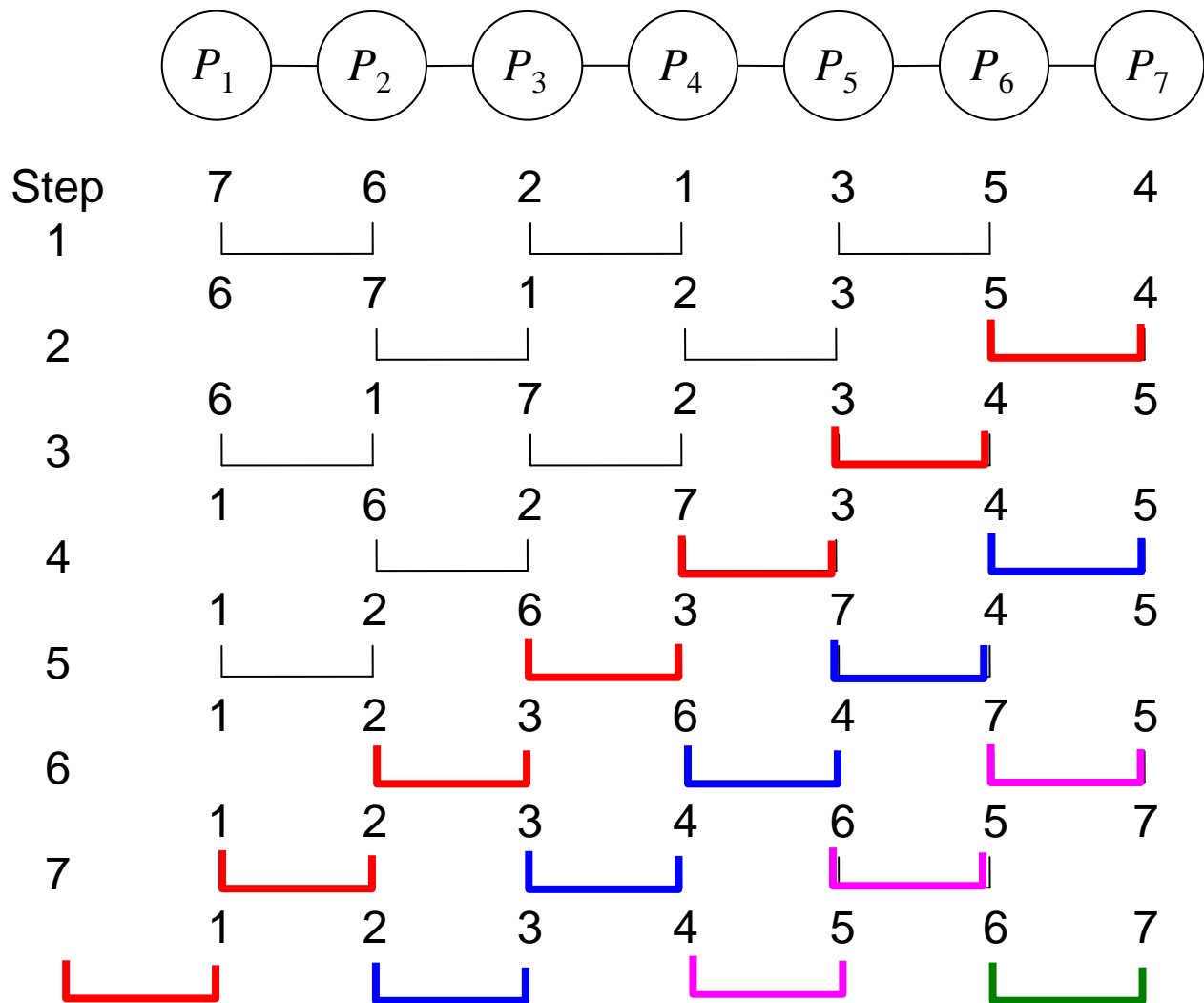
* For $p \leq n^{1/2}$, the proposed algorithm is cost-optimal.

Remark: Given a parallel algorithm with **computation time** $C(n)$ and **communication time** $E(n)$. (Note that total running time $T(n) = C(n) + E(n)$.) If $C(n) < E(n)$, it is possible to reduce the cost and even the total running time, by using fewer processors. In this case, using fewer processors may increase $C(n)$, and possibly decrease $E(n)$; thus $C(n)$ and $E(n)$ are balanced.

*	p	1	$n^{1/2}$	n
	$C(n) = n^*(n/p)$	n^2	$n^{3/2}$	n
	$E(n) = p*n$	1	$n^{3/2}$	n^2
	$T(n)$	n^2	$n^{3/2}$	n^2
	cost	n^2	n^2	n^3

partition + pipeline ???

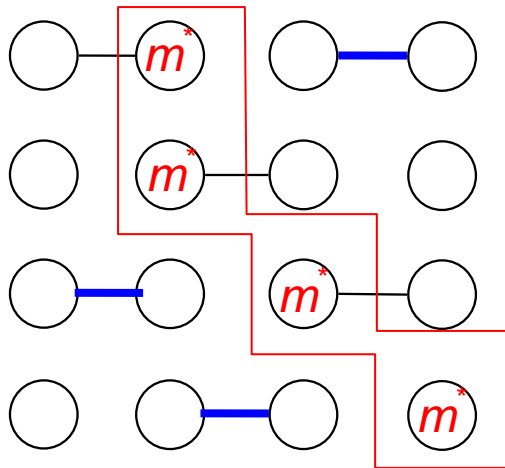
Appendix 1.



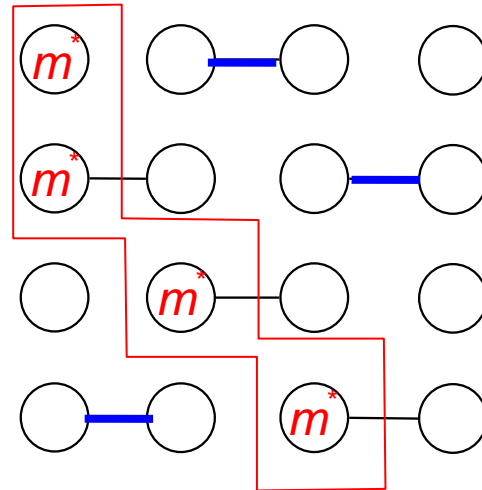
Appendix 2.

(Assume $n = q \geq 3$ holds)

(2) $n=4$ case 1:odd-even



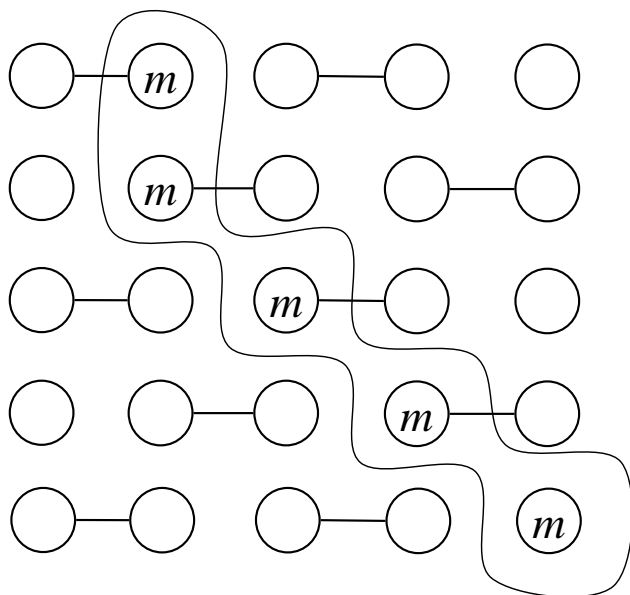
case 2:even-odd



(Prove $n = q+1$ also holds)

(3) $n=5$ (Assume the maximum m is stored in P_k .)

case 1.1:odd-even, k is even



case 1.2:odd-even, k is odd

