

SCHEDULING

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Outline

- The scheduling problem.
- Scheduling without constraints.
- Scheduling under timing constraints.
 - Relative scheduling.
- Scheduling under resource constraints.
 - The ILP model (Integer Linear Programming).
 - Heuristic methods (graph coloring, etc).

Timing constraints versus *resource constraints*

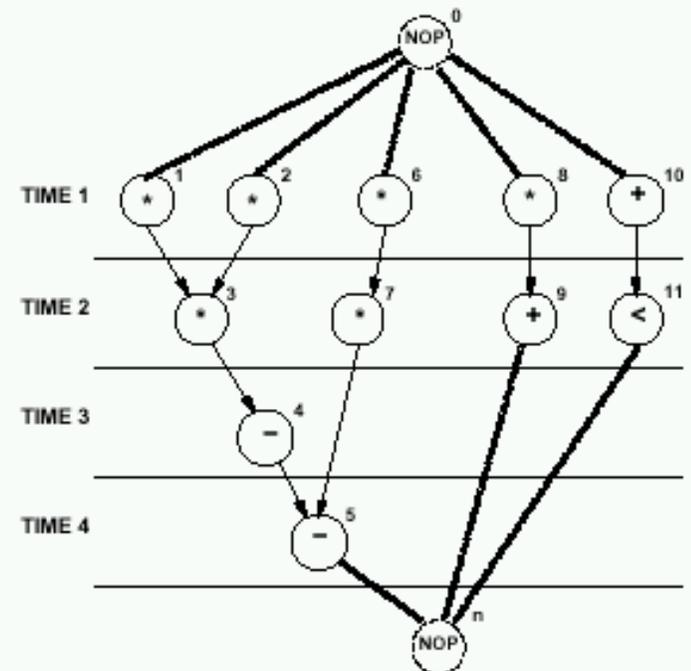
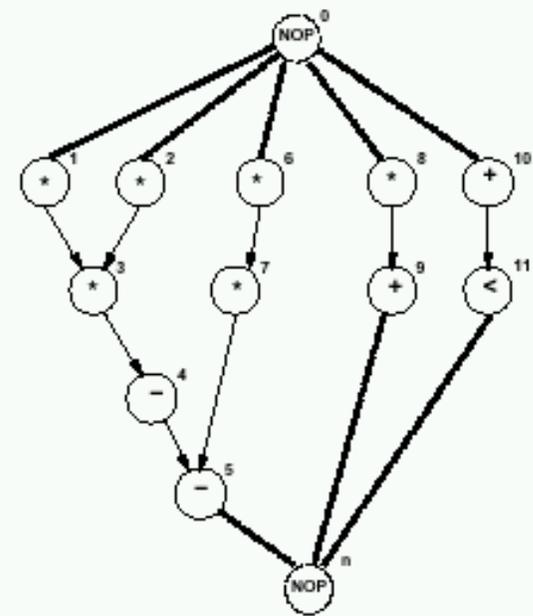
Scheduling

- Circuit model:
 - Sequencing graph.
 - Cycle-time is given.
 - Operation delays expressed in cycles.
- Scheduling:
 - Determine the start times for the operations.
 - Satisfying all the sequencing (timing and resource) constraints.
- Goal:
 - Determine area/latency trade-off.

Do you remember what is latency?

Example

This is As Soon as Possible Scheduling (ASAP). It can be used as a bound in other methods like ILP or when latency only is important, not area.



Taxonomy

- **Unconstrained** scheduling.
- Scheduling with timing constraints:
 - Latency.
 - Detailed timing constraints.
- Scheduling with **resource constraints**.
- Related problems:
 - Chaining. What is chaining?
 - Synchronization. What is synchronization?
 - Pipeline scheduling.

Simplest model

- All operations have bounded delays.
- All delays are expressed in **numbers of cycles** of a **single one-phase clock**.
 - Cycle-time is given.
- **No constraints** - no bounds on area.
- **Goal:**
 - Minimize **latency**.

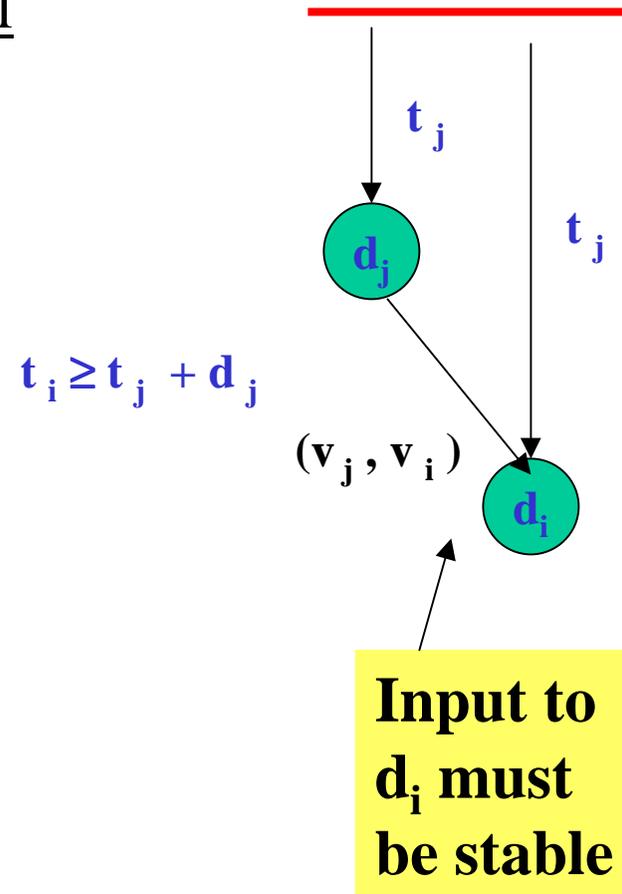
Minimum-latency unconstrained scheduling problem

- Given a set of operations \mathbf{V} with set of corresponding integer delays \mathbf{D} and a partial order on the operations \mathbf{E} :

- Find an integer labeling of the operations

$\varphi : \mathbf{V} \rightarrow \mathbf{Z}^+$, such that:

- $t_i = \varphi(v_i)$,
- $t_i \geq t_j + d_j \forall i, j$ such that $(v_j, v_i) \in E$
- and t_n is *minimum*.

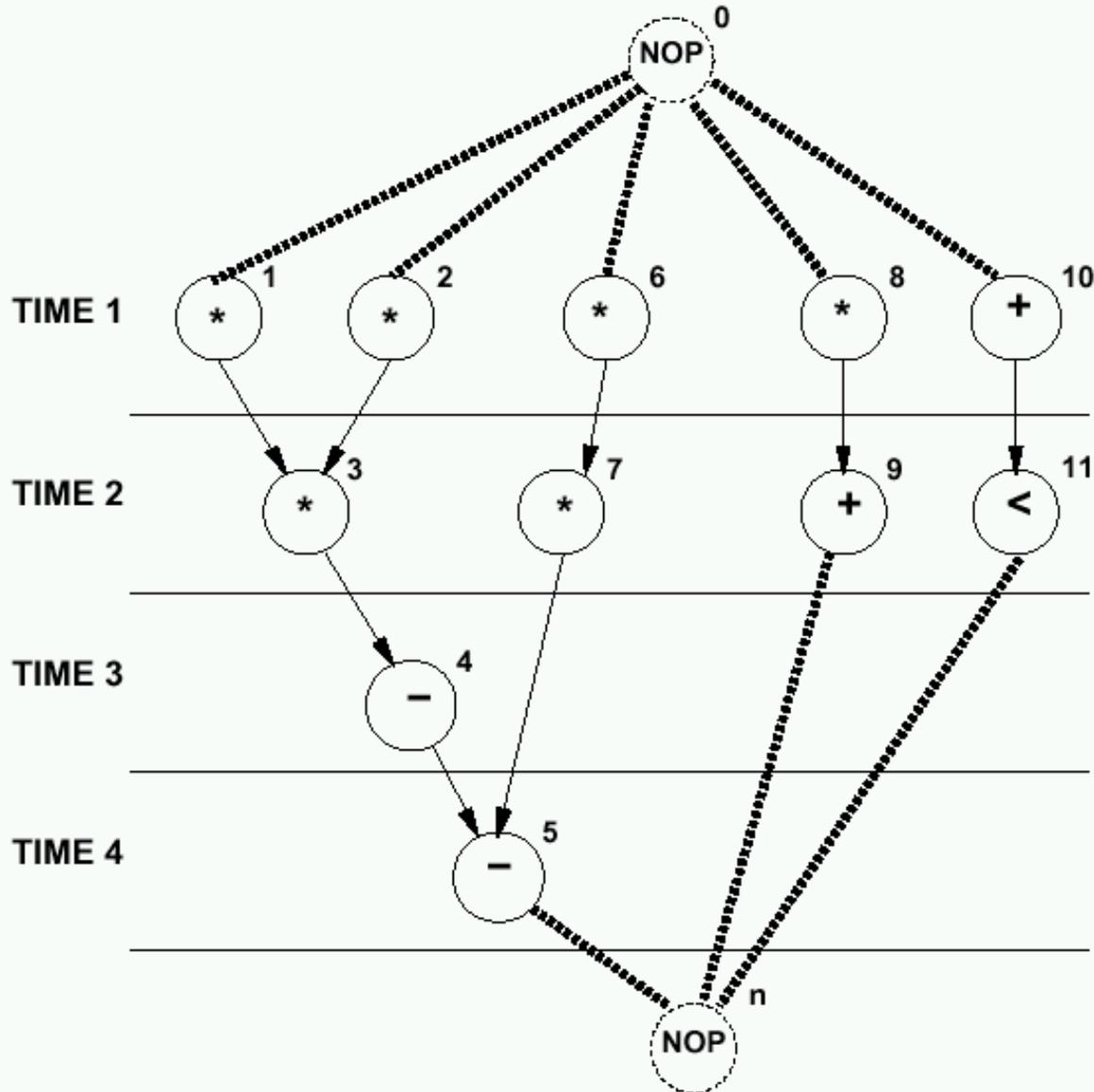


ASAP scheduling algorithm

```
ASAP (  $G_s(V, E)$  ){  
  Schedule  $v_0$  by setting  $t^S_0 = 1$ ;  
  repeat {  
    Select a vertex  $v_i$  whose predecessors are all scheduled;  
    Schedule  $v_i$  by setting  $t^S_i = \max_{j:(v_j, v_i) \in E} t^S_j + d_j$  ;  
  }  
  until ( $v_n$  is scheduled) ;  
  return ( $t^S$  );  
}
```

Similar to breadth-first search

Example - ASAP



- Solution
- Multipliers = 4
- ALUs = 2

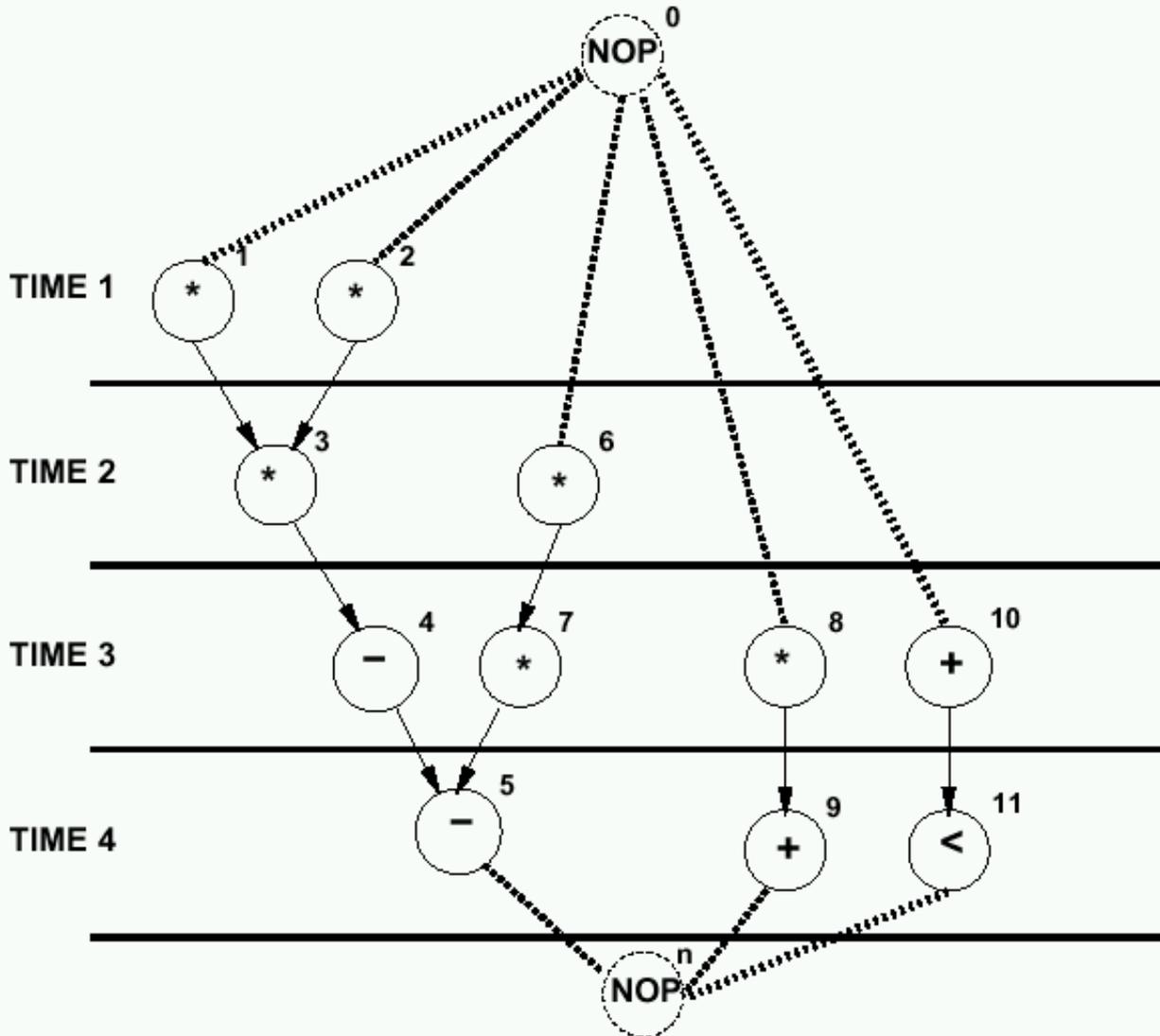
Latency Time=4

ALAP scheduling algorithm

```
ALAP(  $G_s(V, E), \bar{\lambda}$  ) {  
  Schedule  $v_n$  by setting  $t_n^L = \bar{\lambda} + 1$ ;  
  repeat {  
    Select vertex  $v_i$  whose succ. are all scheduled;  
    Schedule  $v_i$  by setting  $t_i^L = \min_{j:(v_i, v_j) \in E} t_j^L - d_i$  ;  
  }  
  until ( $v_0$  is scheduled) ;  
  return ( $\mathbf{t}^L$ );  
}
```

- As Late as Possible - ALAP
- Similar to depth-first search

Example ALAP



- Solution
- multipliers = 2
- ALUs = 3

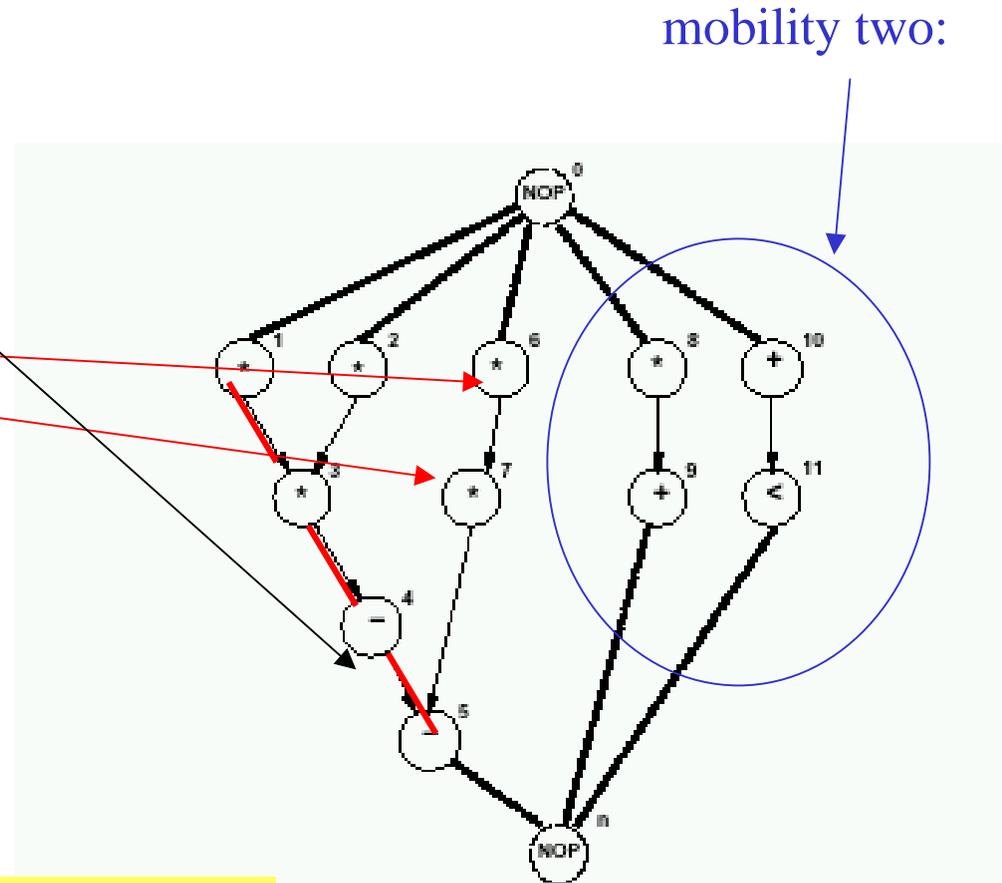
Latency Time=4

Remarks

- ALAP solves a **latency-constrained** problem.
- Latency bound can be set to **latency computed by ASAP algorithm.** <-- using bounds, also in other approaches
- **Mobility:**
 - Mobility is defined for **each operation.**
 - **Difference** between ALAP and ASAP schedule.
- What is **mobility?** **number of cycles that I can move upwards or downwards the operation**
- **Slack on the start time.**

Example of using mobility

- Operations with **zero** mobility:
 - $\{V_1, V_2, V_3, V_4, V_5\}$.
 - *They are on the critical path.*
- Operations with mobility **one**:
 - $\{V_6, V_7\}$.
- Operations with mobility **two**:
 - $\{V_8, V_9, V_{10}, V_{11}\}$.



1. Start from ALAP
2. Use mobility to improve

Last slide for today

Scheduling under detailed timing constraints

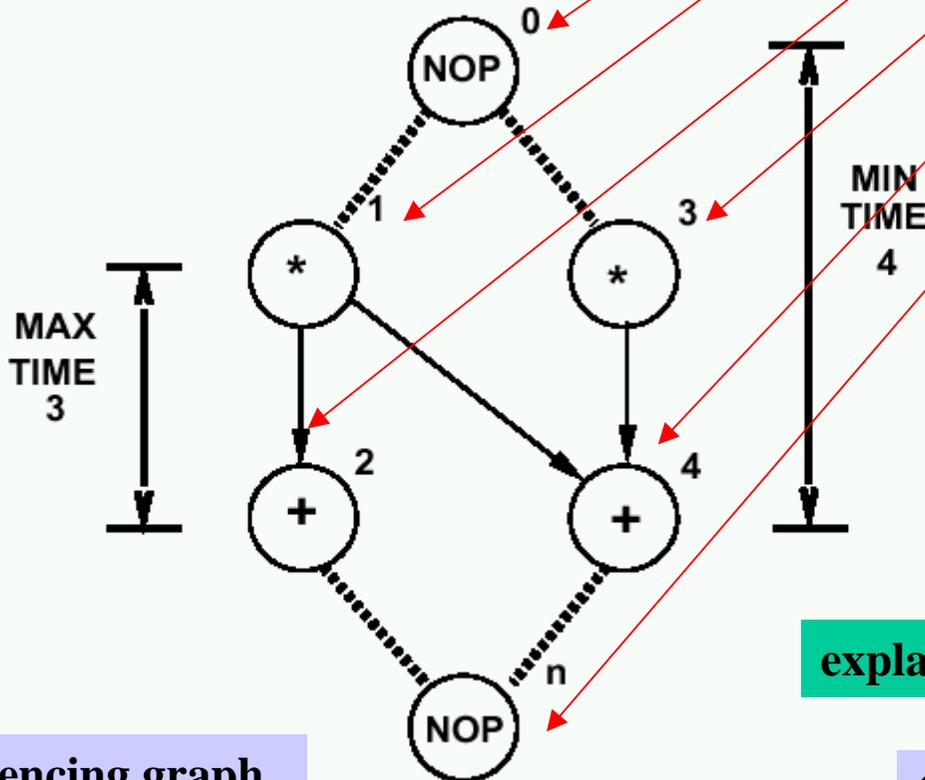
- Motivation:
 - Interface design.
 - Control over operation start time.
- Constraints:
 - Upper/lower bounds on start-time difference of any operation pair.
- **Feasibility** of a solution.

Constraint graph model

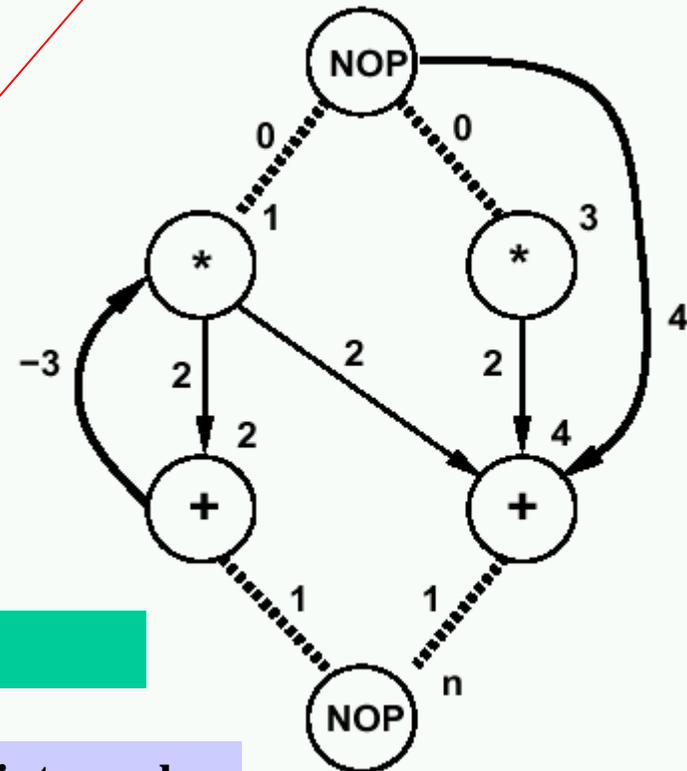
- Start from a sequencing graph.
- Model **delays** as **weights on edges**.
- Add **forward edges** for minimum constraints.
 - Edge (v_i, v_j) with weight $l_{ij} \Rightarrow t_j \geq t_i + l_{ij}$.
- Add **backward edges** for maximum constraints.
 - Edge (v_i, v_j) with weight:
 - $-u_{ij} \Rightarrow t_j \leq t_i + u_{ij}$
 - because $t_j \leq t_i + u_{ij} \Rightarrow t_i \geq t_j - u_{ij}$

Example of using constraint graph with minimum and maximum constraints

Vertex	Start time
v_0	1
v_1	1
v_2	3
v_3	1
v_4	5
v_n	6



explain



Methods for scheduling under detailed timing constraints

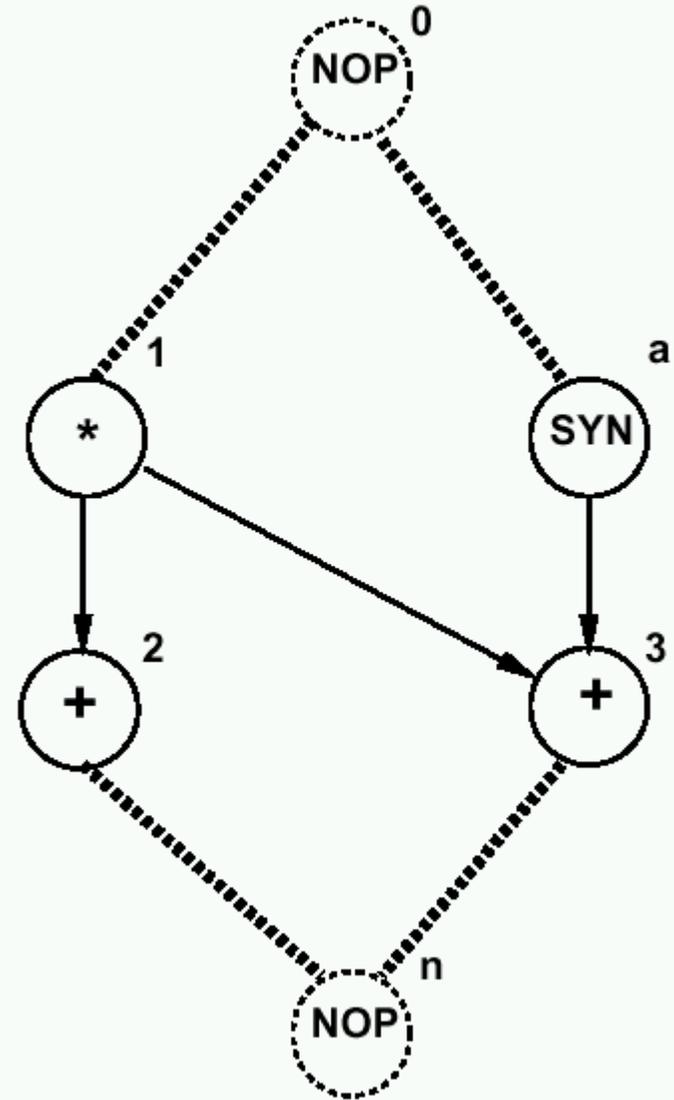
- Assumption:
 - All delays are fixed and known.
- Set of linear inequalities.
- Longest path problem.
- Algorithms for the longest path problem were discussed in Chapter 2:
 - Bellman-Ford,
 - Liao-Wong.

Method for scheduling with unbounded-delay operations

- Unbounded delays:
 - Synchronization.
 - Unbounded-delay operations (e.g. *loops*).
- *Anchors*.
 - Unbounded-delay operations.
- **Relative** scheduling:
 - Schedule operations **with respect to the anchors**.
 - **Combine** schedules.

Example of what?

- $t_3 = \max \{t_1 + d_1 ; t_a + d_a\}$



Relative scheduling method

- For each vertex:
 - Determine *relevant anchor set* $R(\cdot)$.
 - Anchors affecting start time.
 - Determine time offset *from anchors*.
- Start-time:
 - Expressed by:
$$t_i = \max_{a \in R(v_i)} \{t_a + d_a + t_i^a\}$$
 - Computed only at run-time because delays of anchors are unknown.

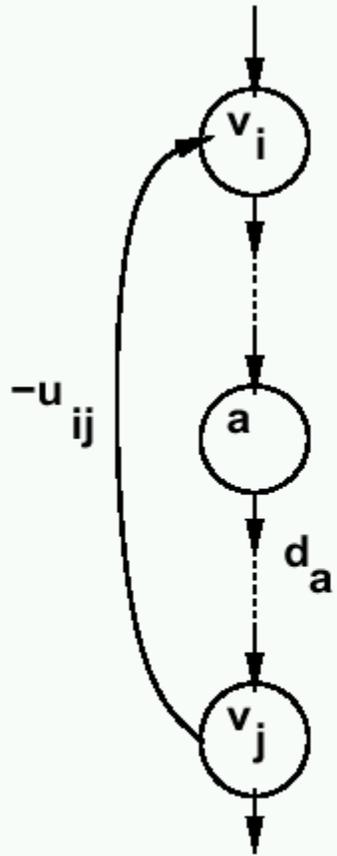
Relative scheduling under timing constraints

- Problem definition:
 - Detailed timing constraints.
 - Unbounded delay operations.
- Solution:
 - May or may not exist.
 - Problem may be ill-specified.

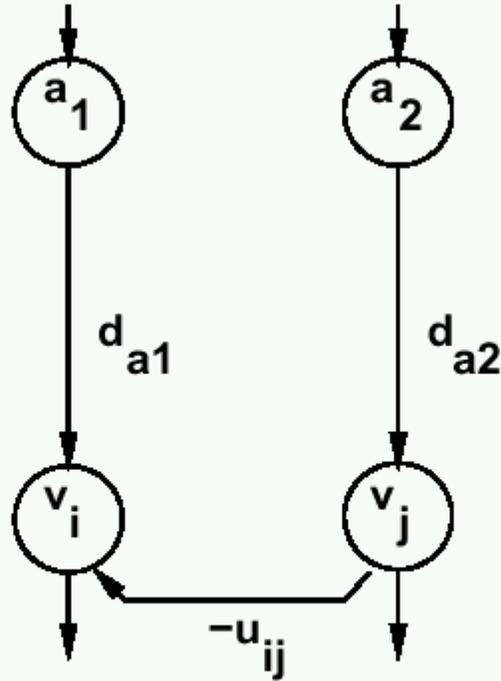
Relative scheduling under timing constraints

- Feasible problem:
 - A solution exists **when unknown delays are zero.**
- Well-posed problem:
 - A solution exists **for any value** of the unknown delays.
- Theorem:
 - A constraint graph can be made **well-posed** *iff* there are **no cycles** with unbounded weights.

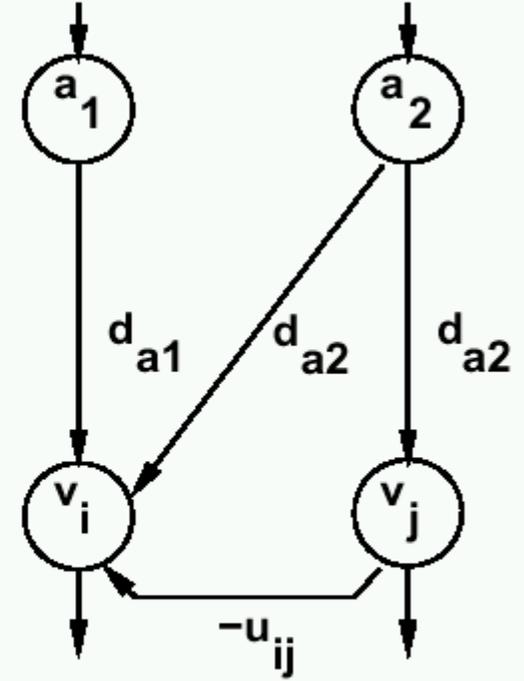
Example of Relative scheduling under timing constraints



(a)



(b)



(c)

- V

- V

- V

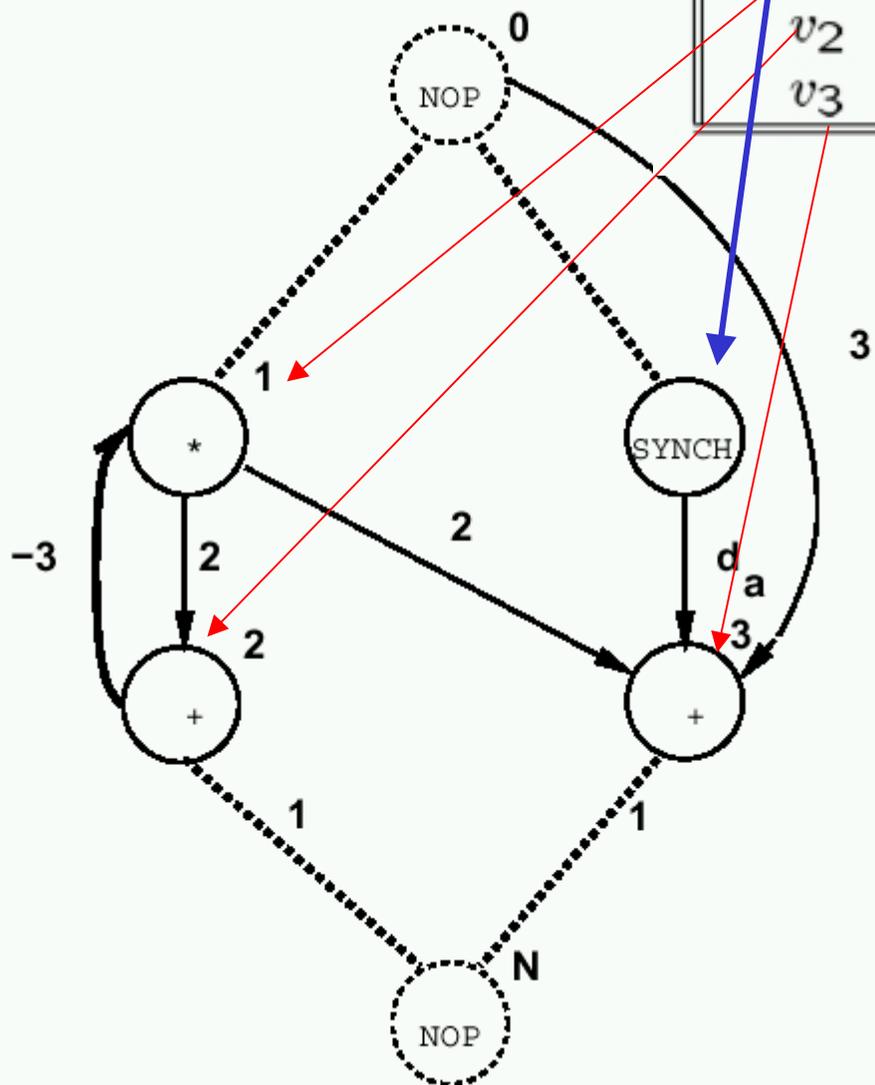
Relative scheduling approach

- Analyze graph:
 - Detect anchors.
 - *Well-posedness* test.
 - Determine dependencies from anchors.
- Schedule ops with respect to relevant anchors:
 - Bellman-Ford, Liao-Wong, Ku algorithms.
- Combine schedules to determine start times:

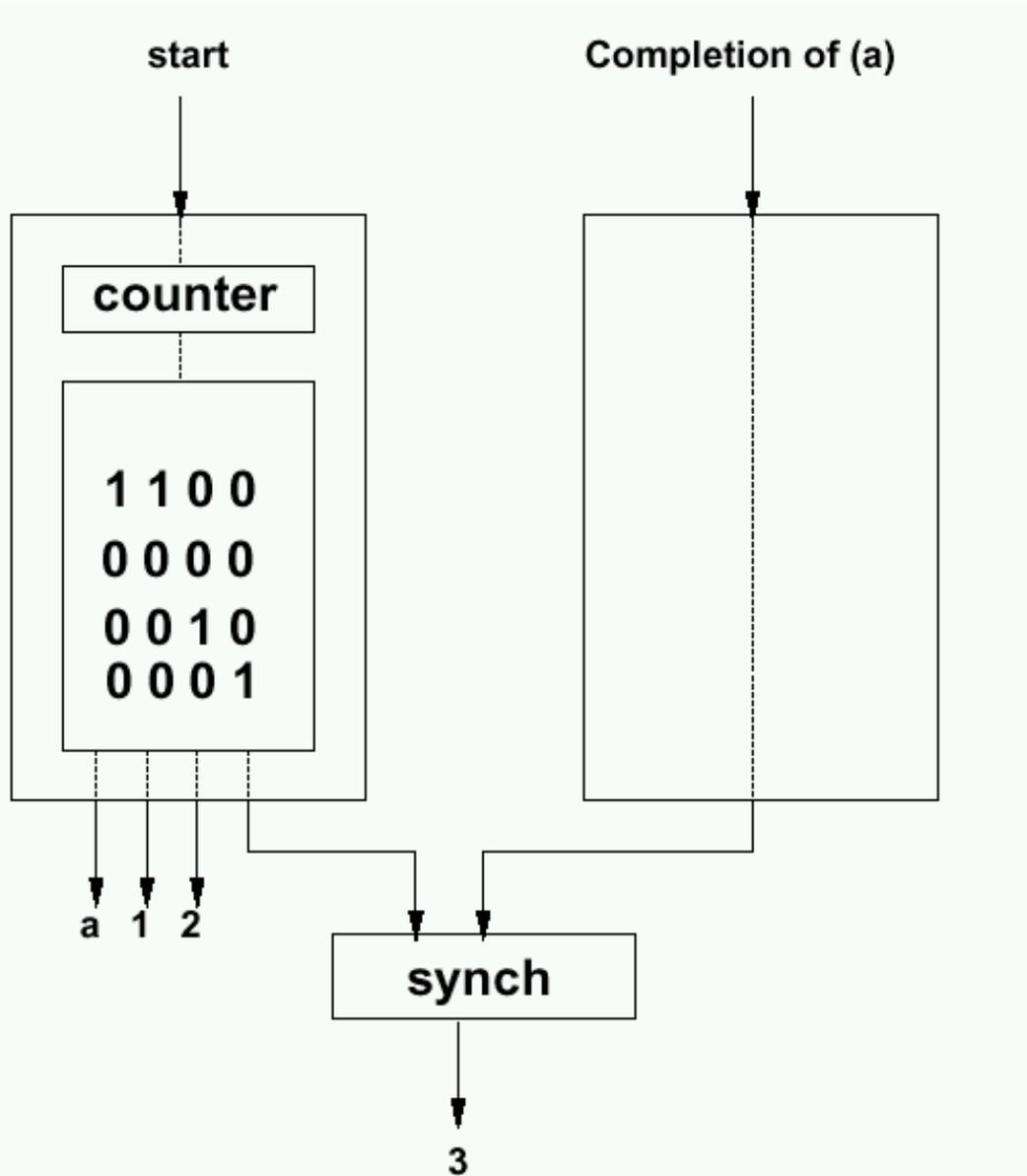
$$t_i = \max_{a \in R(v_i)} \{t_a + d_a + t_i^a\} \quad \forall i$$

Example of Relative scheduling

Vertex v_i	Relevant Anchor Set $R(v_i)$	Offsets	
		t_0	t_a
a	$\{v_0\}$	0	-
v_1	$\{v_0\}$	0	-
v_2	$\{v_0\}$	2	-
v_3	$\{v_0, a\}$	3	0



Example of control-unit synthesized for Relative scheduling



Scheduling under resource constraints

- Classical scheduling problem.
 - Fix **area bound** - minimize **latency**.
- The amount of available resources affects the achievable latency.
- *Dual* problem:
 - Fix **latency bound** - minimize **resources**.
- Assumption:
 - All delays **bounded** and **known**.

Minimum latency resource-constrained scheduling problem

- Given a set of operations V with integer delays D a partial order on the operations E , and upper bounds $\{a_k ; k = 1, 2, \dots, n_{res}\}$:
- Find an integer labeling of the operations $\varphi : V \rightarrow \mathbb{Z}^+$ such that :
 - $t_i = \varphi(v_i)$,
 - $t_i \geq t_j + d_j \forall i, j \text{ s.t. } (v_j, v_i) \in E$,
 - $|\{v_i | \mathcal{T}(v_i) = k \text{ and } t_i \leq l < t_i + d_i\}| \leq a_k$
 $\forall \text{types } k = 1, 2, \dots, n_{res} \text{ and } \forall \text{ steps } l$
 - and t_n is *minimum*.

Scheduling under resource constraints

- Intractable problem.
- Algorithms:
 - **Exact:**
 - **Integer linear** program.
 - **Hu** (restrictive assumptions).
 - **Approximate:**
 - **List** scheduling.
 - **Force-directed** scheduling.

ILP formulation:

- Binary decision variables:

- $X = \{x_{il}; i = 1, 2, \dots, n; l = 1, 2, \dots, \bar{\lambda} + 1\}$.
- x_{il} is TRUE only when operation v_i starts in step l of the schedule (i.e. $l = t_i$).
- $\bar{\lambda}$ is an upper bound on latency.

Start time of operation v_i :

$$\sum_l l \cdot x_{il}$$

ILP formulation constraints

- Operations start only once.

$$\sum_l x_{il} = 1 \quad i = 1, 2, \dots, n$$

- Sequencing relations must be satisfied.

$$t_i \geq t_j + d_j \quad \forall (v_j, v_i) \in E$$

$$\sum_l l \cdot x_{il} - \sum_l l \cdot x_{jl} - d_j \geq 0 \quad \forall (v_j, v_i) \in E$$

ILP formulation constraints (cont)

Resource bounds must be satisfied.

- Simple case (unit delay)

- $$\sum_{i:\mathcal{T}(v_i)=k} x_{il} \leq a_k \quad k = 1, 2, \dots, n_{res}; \quad \forall l$$

ILP Formulation

Sequencing relations must be satisfied.

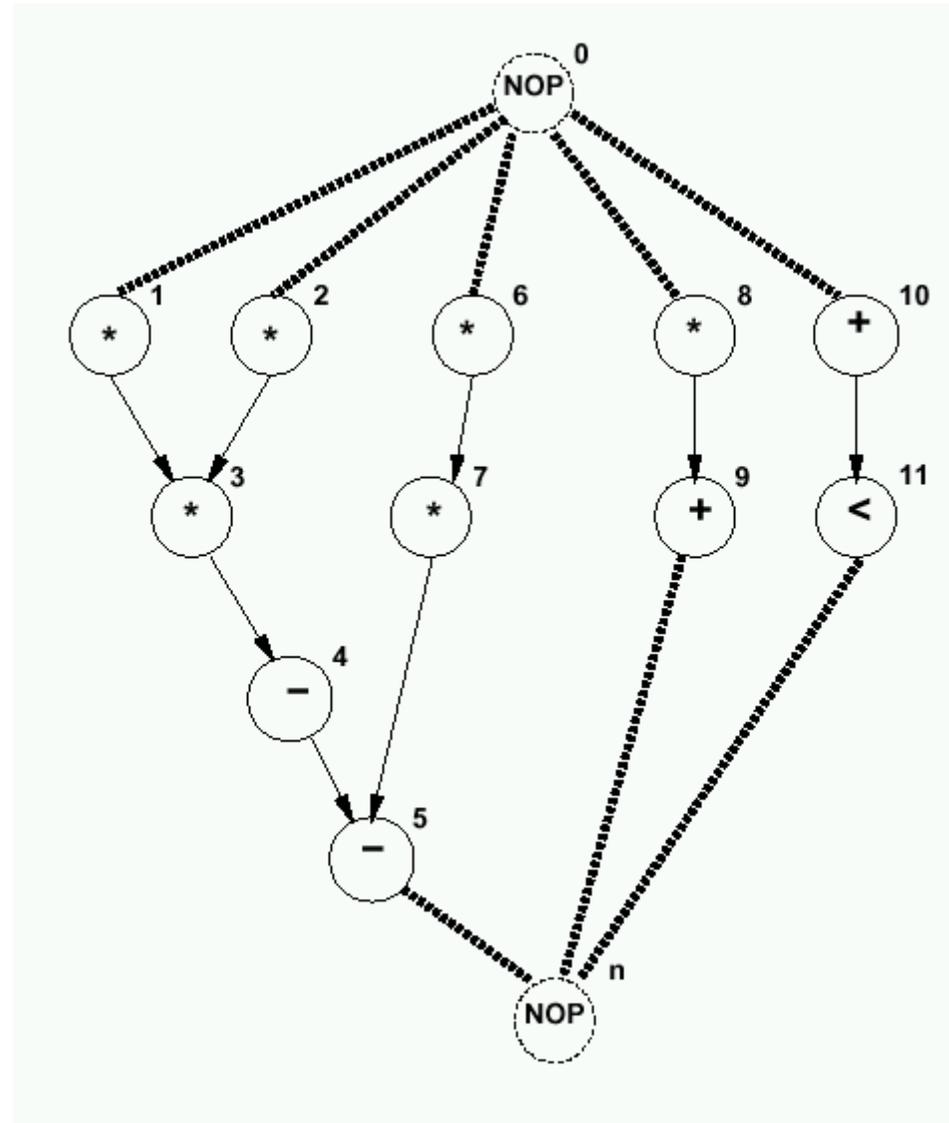
Operations start only once

$$\begin{aligned} & \min \quad \|\mathbf{t}\| \quad \text{such that} \\ & \sum_j x_{ij} = 1 \quad i = 1, 2, \dots, n \\ & \sum_l l \cdot x_{il} - \sum_l l \cdot x_{jl} - d_j \geq 0 \quad i, j = 1, 2, \dots, n, (v_j, v_i) \in E \\ & \sum_{i: \mathcal{T}(v_i)=k} \sum_{m=l-d_i+1}^l x_{im} \leq a_k \quad k = 1, 2, \dots, n_{res}; l = 0, 1, \dots, t_n \end{aligned}$$

Resource bounds must be satisfied

Example of ILP Formulation

- Resource constraints:
 - 2 ALUs, 2 Multipliers.
 - $a_1 = 2, a_2 = 2$.
- Single-cycle operation.
 - $d_i = 1 \forall i$.



Example ILP

Operations start only once.

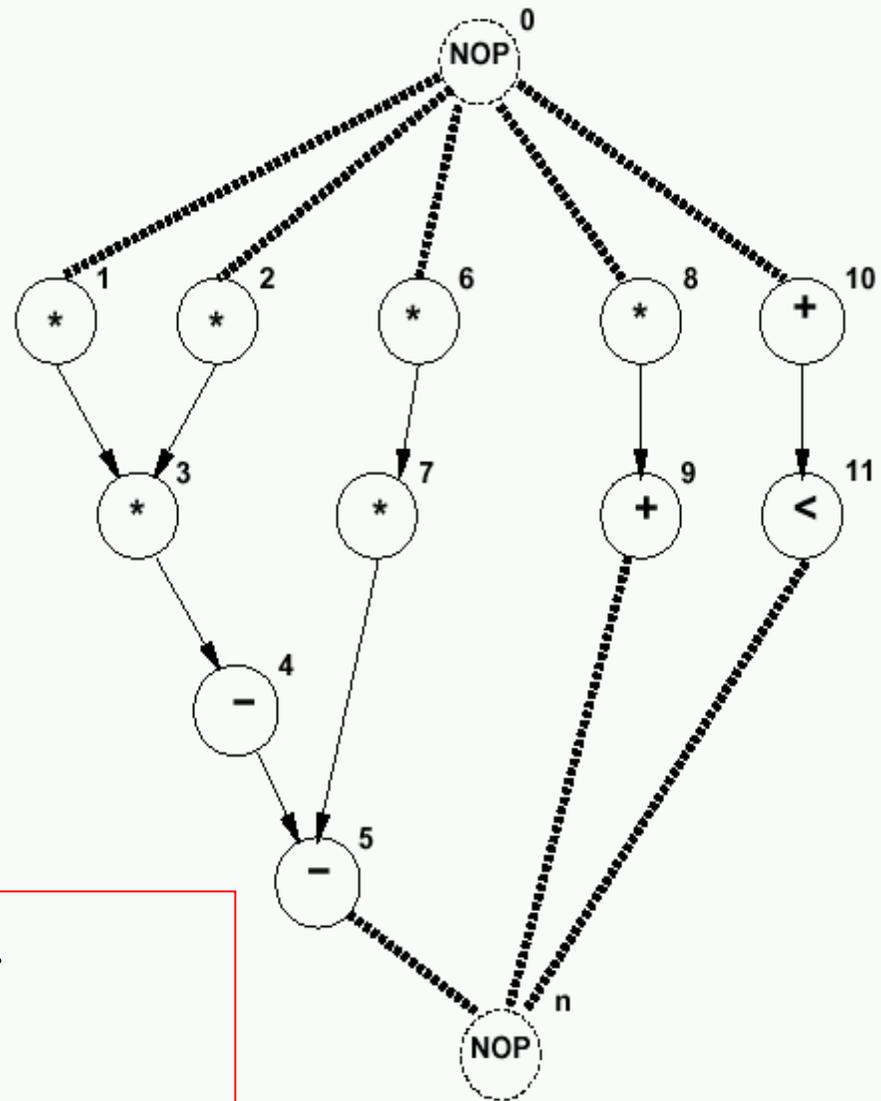
- $x_{11} = 1$
- $x_{61} + x_{62} = 1$
- ...

• **Sequencing relations must be satisfied.**

- $x_{61} + 2x_{62} - 2x_{72} - 3x_{73} + 1 \leq 0$
- $2x_{92} + 3x_{93} + 4x_{94} - 5x_{N5} + 1 \leq 0$
-

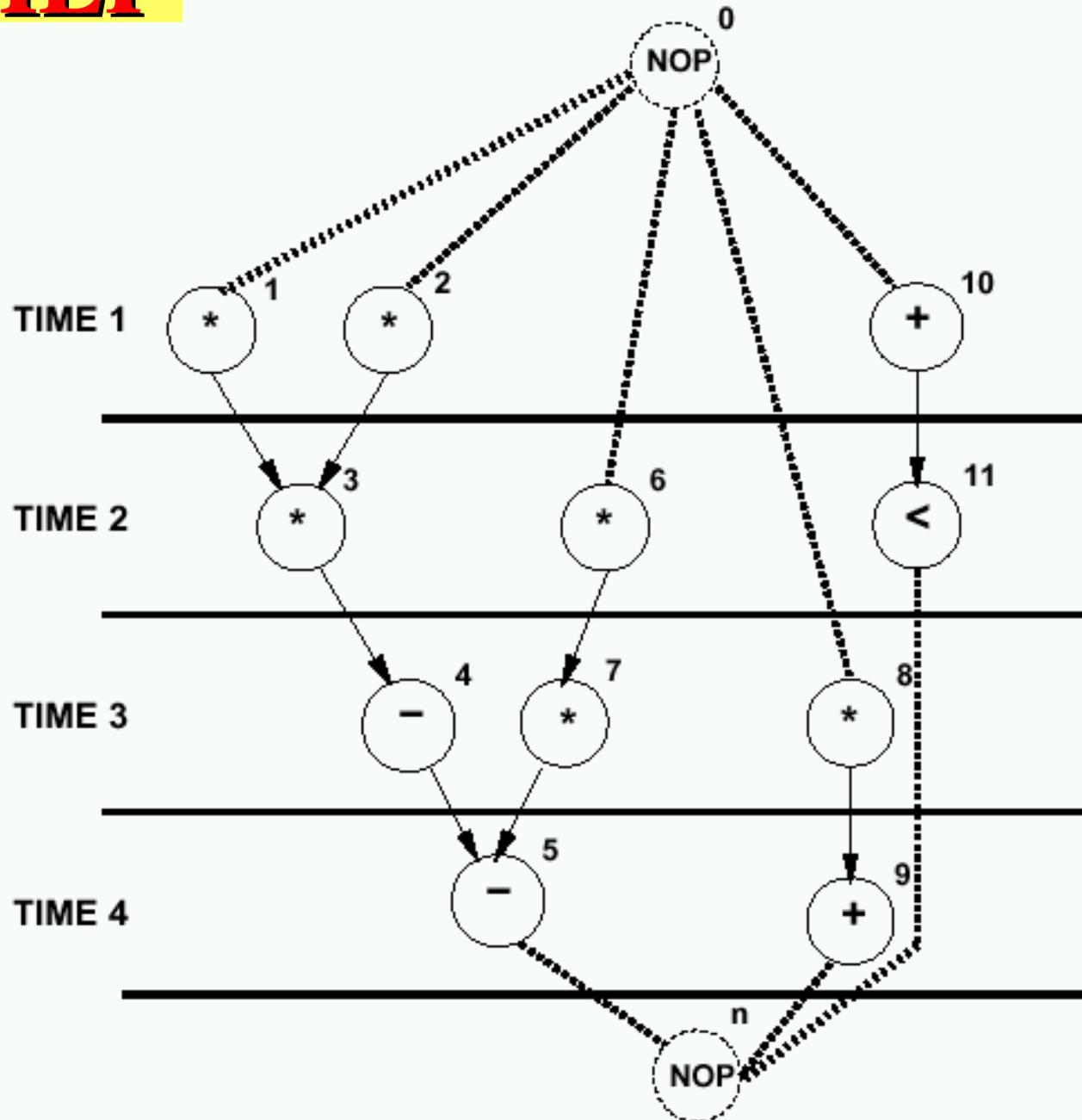
Resource bounds must be satisfied.

- $x_{11} + x_{21} + x_{61} + x_{81} \leq 2$
- $x_{32} + x_{62} + x_{72} + x_{82} \leq 2$
- ...



Example ILP

- Solution
- latency 4
- multipliers =2
- ALU =2



Dual ILP formulation

- Minimize resource usage under latency constraints.
- Additional constraint:
 - Latency bound must be satisfied.

–

$$\sum_l l x_{nl} \leq \bar{\lambda} + 1$$

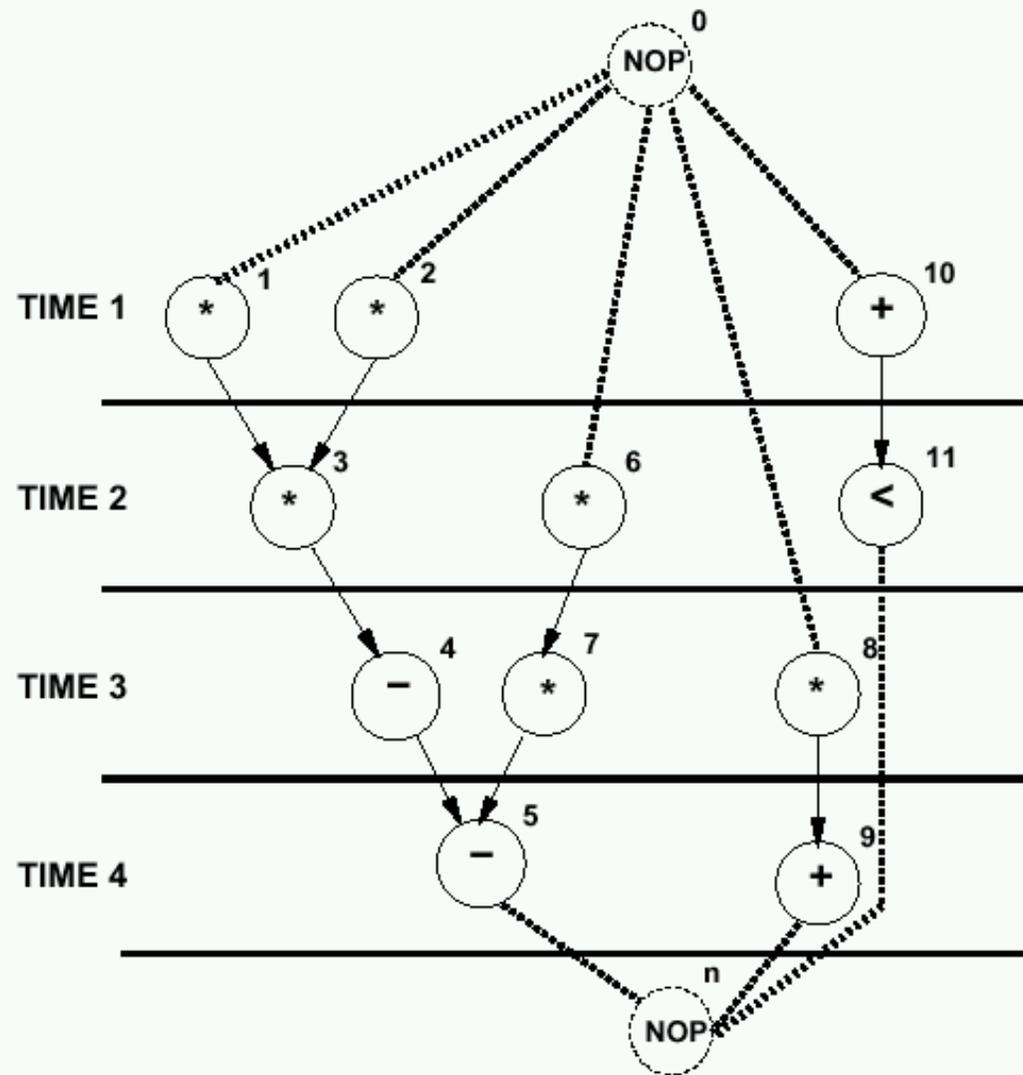
- Resource usage is unknown in the constraints.
- Resource usage is the objective to minimize.

Example

Multipliers = 2

ALUs = 2

$2 * 5 + 2 * 1 = 12 = \text{cost}$
of solution



- Multiplier area = 5. ALU area = 1.
- Objective function: $5a_1 + a_2$.

ILP Solution

- Use **standard ILP** packages.
- Transform into LP (linear programming) problem [Gebotys].
- *Advantages:*
 - Exact method.
 - Others constraints can be incorporated.
- *Disadvantages:*
 - Works well up to few thousand variables.