



Constraint Analysis in Vulcan

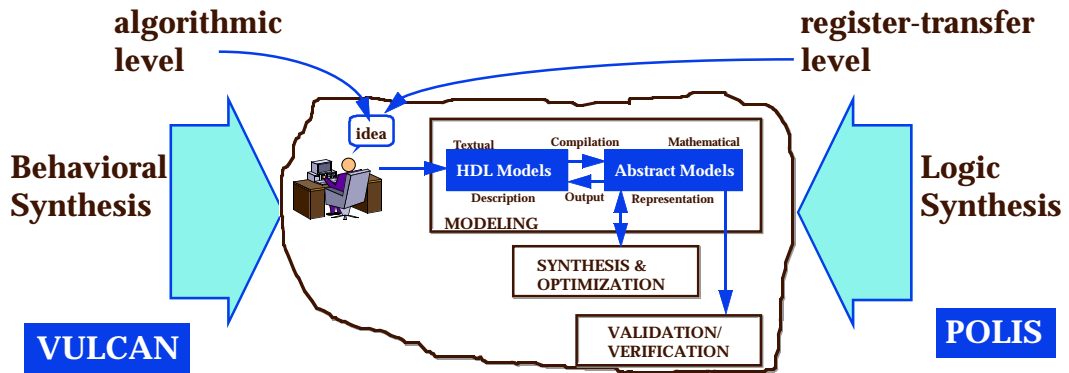
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The Co-Synthesis Approach




The hardware software co-design problem is posed as an evolution of existing synthesis Methods



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Vulcan

- 1 Specification
- 1 Modeling
-  1 Constraint Analysis
- 1 Software and Runtime Environment
- 1 Target Architecture - H/S Interface
- 1 Partitioning
- 1 Co-simulation

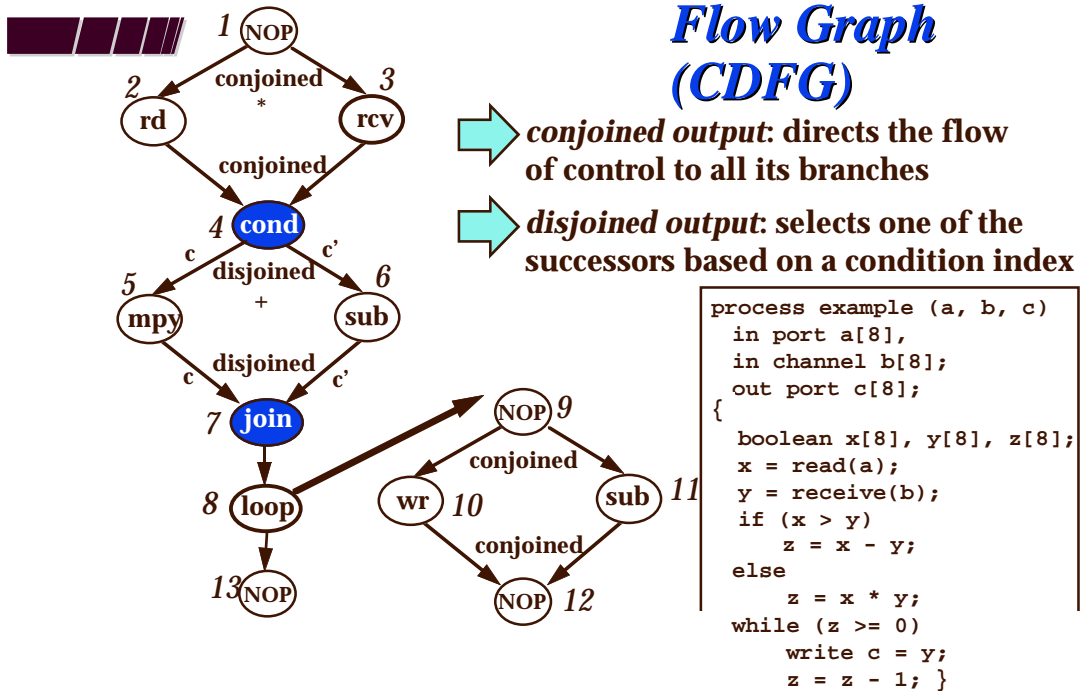
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Model -- Flow Graph

- 1 Hierarchical control/data-flow graph
 - ◆ control flow primitives (iteration and model call modeled though hierarchy)
- 1 Acyclic
 - ◆ models partial order of tasks/operations
 - ◆ iteration is modeled outside the graph
- 1 Polar
 - ◆ source and sink vertices model No-Operatios

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Flow Graph (CDFG)



```

process example (a, b, c)
  in port a[8],
  in channel b[8];
  out port c[8];
  {
    boolean x[8], y[8], z[8];
    x = read(a);
    y = receive(b);
    if (x > y)
      z = x - y;
    else
      z = x * y;
    while (z >= 0)
      write c = y;
      z = z - 1; }
  
```

Operation vertices in a Flow Graph

- 1 **no-op**: no operation
- 1 **cond**: conditional fork
- 1 **join**: conditional join
- 1 **op-logic**: logical operations
- 1 **op-arithmetic**: arithmetic operations
- 1 **op-relational**: relational operations
- 1 **op-io**: I/O operations
- 1 **wait**: wait on a signal variable
- 1 **link**: hierarchical operations
 - ◆ **call**: procedure call (invocation times = 1)
 - ◆ **loop**: iteration (invocation times ≥ 1)

System Model

- 1 Consists of one or more flow graphs that may be hierarchically linked to other flow graphs

→ System Model: $\Phi = \{G_1^*, G_2^*, \dots, G_n^*\}$

where

G_i^* represents the process graph model G_i and all the flow graphs that are hierarchically linked to G_i .

- ★ A flow graph model that is common to two hierarchies of a system model is called a *shared model*

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Flow Graphs: Execution Semantics

- 1 At any time, an operation may be
 - ◆ waiting for execution
 - ◆ presently executing
 - ◆ having completed its execution



The state of a vertex is defined as being one of $\{s_r, s_e, s_d\}$

s_r : reset state ==> waiting for execution

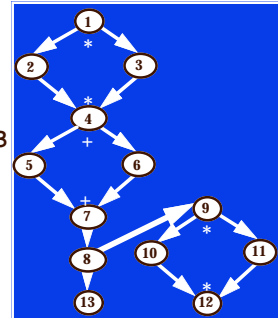
s_e : enable state ==> presently executing

s_d : done state ==> completed execution

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Example

1	2	3	4	5	6	7	8	9	10	11	12	13
-	-	-	-	-	-	-	-	-	-	-	-	-
e	-	-	-	-	-	-	-	-	-	-	-	-
d	e	e	-	-	-	-	-	-	-	-	-	-
-	d	d	e	-	-	-	-	-	-	-	-	-
-	-	-	d	e	-	-	-	-	-	-	-	-
-	-	-	-	d	-	e	-	-	-	-	-	-
-	-	-	-	-	-	d	e	-	-	-	-	-
-	-	-	-	-	-	-	e	e	-	-	-	-
-	-	-	-	-	-	-	e	d	e	-	-	-
-	-	-	-	-	-	d	e	-	d	d	e	-
-	-	-	-	-	-	-	d	-	-	d	e	-
e	-	-	-	-	-	-	-	-	-	-	-	d



- → reset
e → enable
d → done

→ No assumption about timing of the operations is made =>
(consecutive rows can be spaced arbitrarily over the time axis)

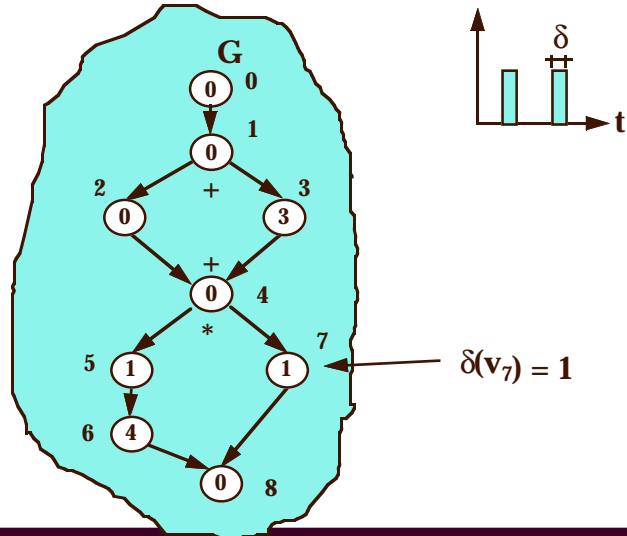
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Timing Properties

- 1 Operation delay
- 1 Graph Latency
- 1 Rate of Execution (operations)
- 1 Rate of Reaction (graphs)

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Operation Delay



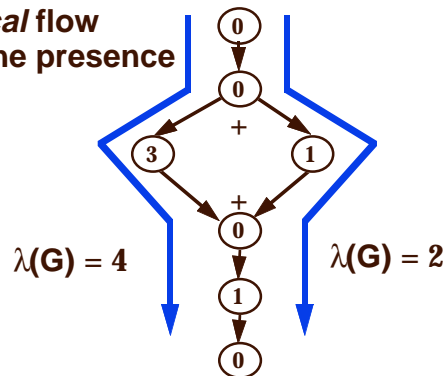
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Latency

- 1 **Latency**, $\lambda(G)$, of a graph model G refers to the execution delay of G

$$\lambda_k(G) = t_k(v_n) - t_k(v_0)$$

- ◆ the latency of a *non-hierarchical* flow graph may be variable due to the presence of conditional paths



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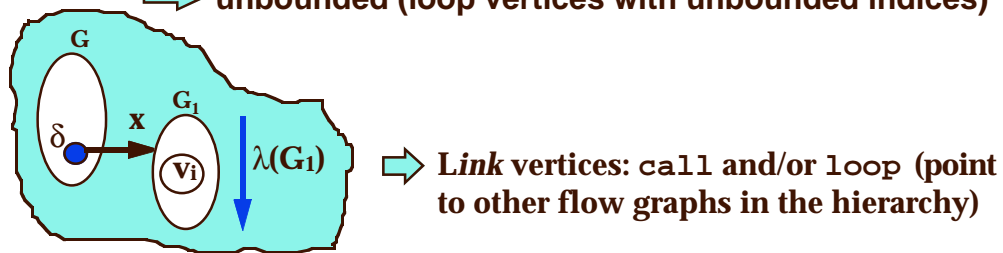
Execution Delay of Link Vertices

1 Given by

- ◆ latency of the corresponding graph model times the number of times the called graph is invoked
- ◆ execution delay of a link vertex can be

⇒ variable

⇒ unbounded (loop vertices with unbounded indices)



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Timing Properties

1 Operation delay

1 Graph Latency

⇒ 1 Rate of Execution (operations)

⇒ 1 Rate of Reaction (graphs)

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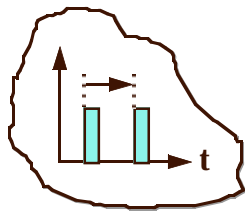
Rate of Execution (operations)

- 1 assuming a *synchronous* execution model with cycle time τ ,

→ the rate of execution at invocation k of operation v_i is given by the time interval between its current and previous execution

$$\rho_i(\mathbf{k}) := \frac{1}{t_k(v_i) - t_{k-1}(v_i)} \quad (\text{sec}^{-1})$$

$$= \frac{\tau}{t_k(v_i) - t_{k-1}(v_i)} \quad (\text{cycle}^{-1})$$



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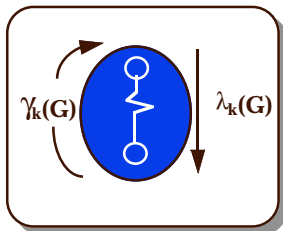
Rate of Reaction (Graphs)

- 1 For a graph model, G , its *rate of reaction* is defined as the rate of execution of its source operation

$$\rho_G(\mathbf{k}) := \rho_0(\mathbf{k})$$

The reaction rate is used to capture the effect on the run-time system of the type of implementation chosen for the graph model

- ♦ e.g., the choice of a non-pipelined implementation leads to



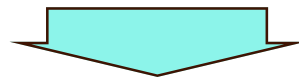
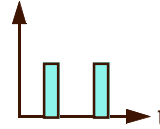
$$\rho_G(\mathbf{k})^{-1} = \lambda_k(G) + \gamma_k(G)$$

where $\gamma_k(G)$ represents the *overhead delay* (delay of reinvocation of G).

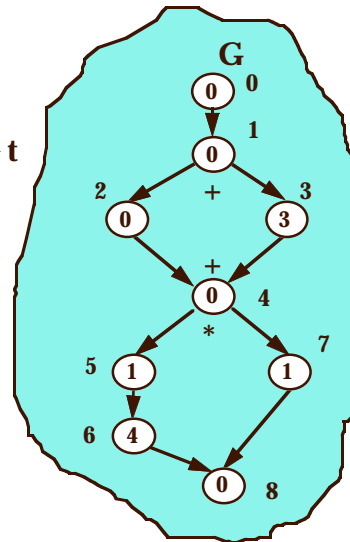
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Timing Properties

- 1 Operation delay
- 1 Graph Latency
- 1 Rate of Execution (operations)
- 1 Rate of Reaction (graphs)



fixed, variable, bounded/unbounded



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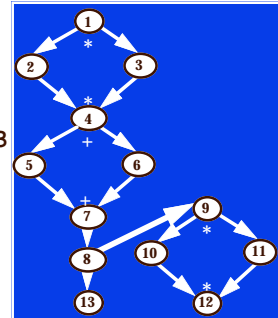
Non-Determinism

- 1 the delay of an operation may be variable, depending on
 - ◆ the *value* of input data: e.g., loops with data dependent iteration counts, call vertices with conditionals
 - ◆ the *timing* of input data: e.g., wait operation
- 1 the latency of a graph may be variable

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//// Data Dependent Delays

1	2	3	4	5	6	7	8	9	10	11	12	13
-	-	-	-	-	-	-	-	-	-	-	-	-
e	-	-	-	-	-	-	-	-	-	-	-	-
d	e	e	-	-	-	-	-	-	-	-	-	-
-	d	d	e	-	-	-	-	-	-	-	-	-
-	-	-	d	e	-	-	-	-	-	-	-	-
-	-	-	-	d	-	-	-	-	-	-	-	-
-	-	-	-	-	-	e	-	-	-	-	-	-
-	-	-	-	-	-	-	e	-	-	-	-	-
-	-	-	-	-	-	-	-	e	-	-	-	-
-	-	-	-	-	-	-	-	-	e	-	-	-
-	-	-	-	-	-	-	-	-	-	e	-	-
e	-	-	-	-	-	-	-	-	-	-	-	d



- → reset
e → enable
d → done

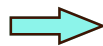
alternative paths

execution delay of "link" vertex

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//// Non-determinism and Execution Rate

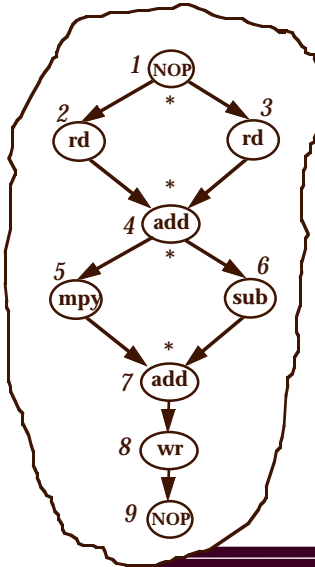
- 1 **Data-dependent loop and synchronization operations introduce uncertainty over**
 - ◆ the precise *execution delay* of the model
 - ◆ the *order of execution* of the operations in the model



Operations with *variable delays* are termed *non-deterministic delay* or *ND operations*.

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////// A Single Rate Model



On each execution of the flow graph, each operation executes once

☀ In this case, the reaction rate of the graph G is:

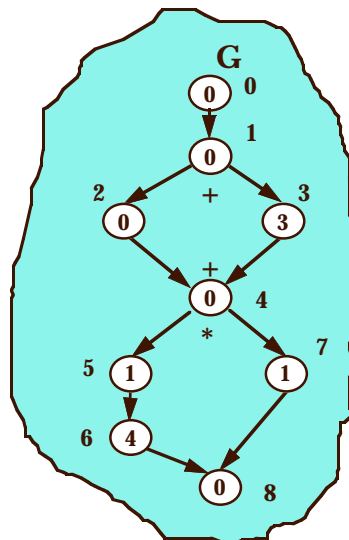
$$\rho_G(\mathbf{k}) := \rho_0(\mathbf{k}) = \rho_{v_i}(\mathbf{k}),$$

for all $v_i \in V(G)$ and for all $k \geq 0$

The execution of G proceeds at *single rate*.

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////// A Multi Rate Model



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Timing Constraints and Constraint Analysis

- ➔ 1 Timing Constraints
- 1 Scheduling
- 1 Constraint Satisfiability

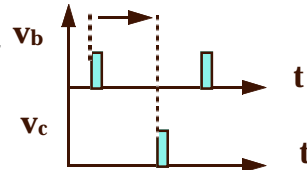
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Timing Constraints

- ➔ 1 *Operation delay* constraints
 - ◆ *unary*: bounds on the delay of an operation
 - ◆ *binary*: bounds on the delay between the starting time of two operations
- 1 Execution rate constraints

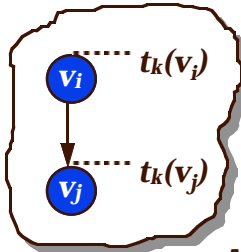
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Binary Delay Constraints



- 1 **Minimum timing constraint**, $l_{ij} \geq 0$ from operation vertex v_i to v_j is defined as

$$t_k(v_j) \geq t_k(v_i) + l_{ij}$$



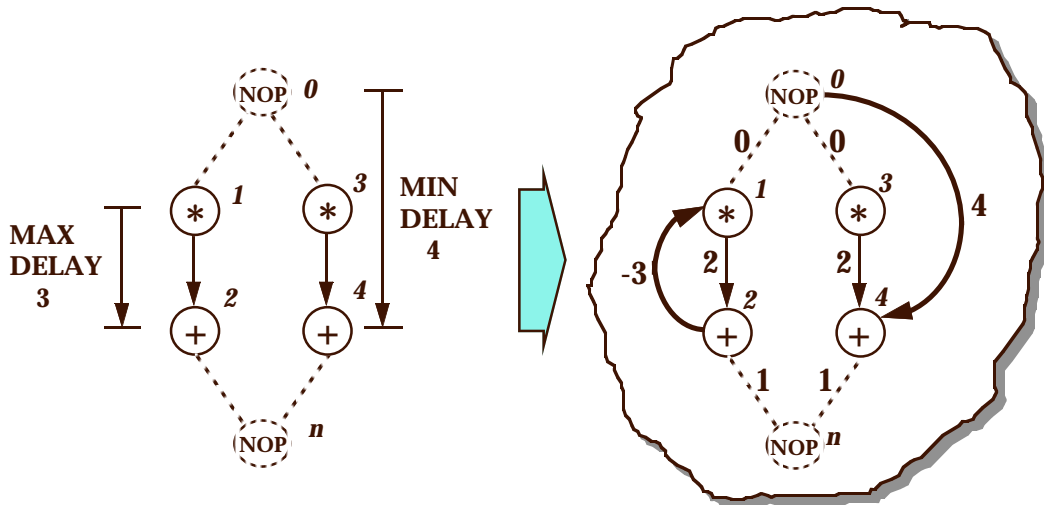
- by default, any sequencing dependency between two operations induces a minimum timing constraint

- 1 **Maximum timing constraint**, $u_{ij} \geq 0$ from operation vertex v_i to v_j is defined as

$$t_k(v_j) \leq t_k(v_i) + u_{ij}$$

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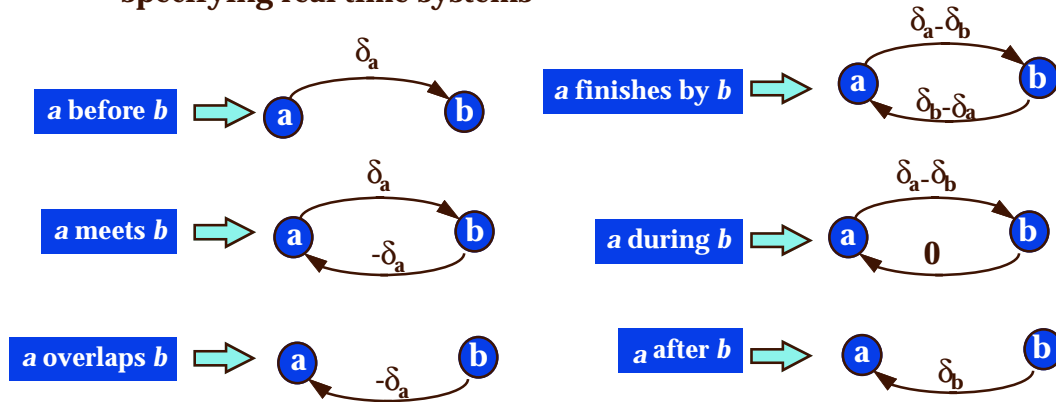
Example



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Operation Delay Constraints

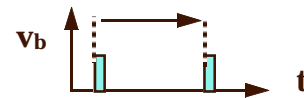
→ Can capture durational and deadline constraints in specifying real time systems



etc...

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Timing Constraints



1 Operation delay constraints

→ 1 Execution rate constraints

- ◆ refer to constraints on the interval of time between successive executions of an operation
 - » rate constraints on *input (output)* operations refer to the rates at which the data is required to be *produced (consumed)*

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//// Data Rate Constraints

- 1 **Minimum data rate constraint, r_i (cycles⁻¹),** on an input/output operation v_i defines a lower bound on the execution rate of the operation

$$\rho_{v_i}(k) \geq r_i \quad \forall k > 0 \quad [\text{min rate}]$$

$$\Rightarrow t_k(v_i) - t_{k-1}(v_i) \leq \tau \cdot r_i^{-1} \quad \forall k > 0$$

- 1 **Maximum data rate constraint, R_i (cycles⁻¹),** on an I/O operation v_i defines an upper bound on the execution rate of the operation

$$\rho_{v_i}(k) \leq R_i \quad \forall k > 0 \quad [\text{max rate}]$$

$$\Rightarrow t_k(v_i) - t_{k-1}(v_i) \geq \tau \cdot R_i^{-1} \quad \forall k > 0$$

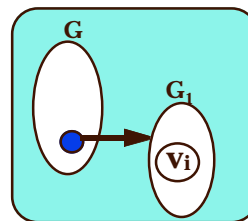
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//// Ex.: Specification of Data Rate Constraints

```

process example (a,b,c)
  in port a[8],b[8];
  out port c[8];
{
  boolean x[8],y[8],z[8],w[8];
  tag A;
  x = read(a);
  y = read(b);
  z = x * y;
  w = x + y;
  while(z >= 0) {
    while(w >= 0) {
      A: write c = y;
          w = w - 1; }
      z = z - w;
      write c = z; }
}

```



relative min constraints -- indexed by the corresponding loops

r = 0.01 per cycle

```

attribute "constraint minrate of A = 100 cycles/sample"
attribute "constraint minrate 0 of A = 1 cycles/sample"
attribute "constraint minrate 1 of A = 10 cycles/sample"

```

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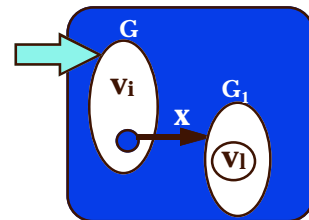
Timing Constraints and Constraint Analysis

- 1 Timing Constraints
- ➔ 1 Scheduling
- 1 Constraint Satisfiability

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Scheduling

- 1 For each invocation of a flow graph model, an operation is invoked zero, one, or many times depending upon its *position on the hierarchy* of the flow model



➔ The execution times $t_k(v)$ of an operation v are determined by two separate mechanisms

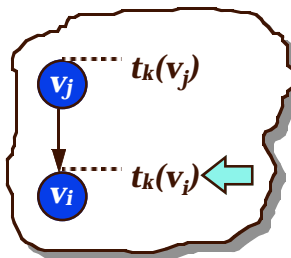
- ☀ The runtime scheduler, γ
 - ➔ determines the invocation time of flow graphs

- ☀ The operation scheduler, Ω

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Timing Constraints and Scheduling

- Given a scheduling function, a timing constraint is considered *satisfied* if
 - the operation starting times determined by the scheduling function satisfy the inequalities



$$t_k(v_j) \geq t_k(v_i) + l_{ij} \quad [\text{min delay}]$$

$$t_k(v_j) \leq t_k(v_i) + u_{ij} \quad [\text{max delay}]$$

$$\rho_{vi}(k) \leq R_i \quad [\text{max rate}]$$

$$\rho_{vi}(k) \geq r_i \quad [\text{min rate}]$$

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Relative Scheduler

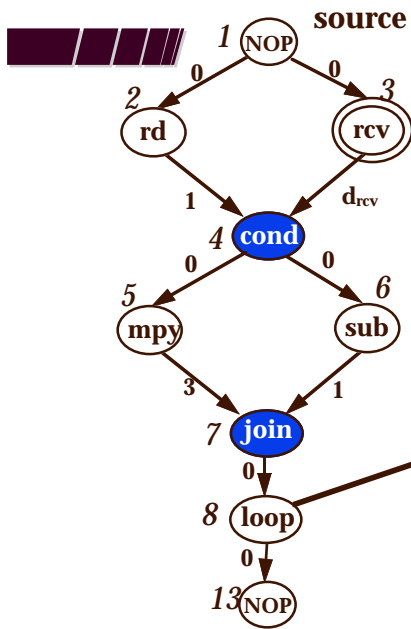
- For a given vertex v_i a set $A(v_i)$ of *anchor* vertices is defined as the set of conditional (CD) and loop, wait (ND) vertices that have a path to v_i

$$A(v_i) = \{v_j \in V : v_j \succ^* v_i, v_j \text{ is ND or CD}\}$$

- A *relative schedule function* Ω_r is defined as a set of offsets for each operation such that the operation start time satisfies

$$t_k(v_i) \geq \max_{a \in A(v_i)} [t_k(a) + \delta(a) + \theta_a(v_i)]$$

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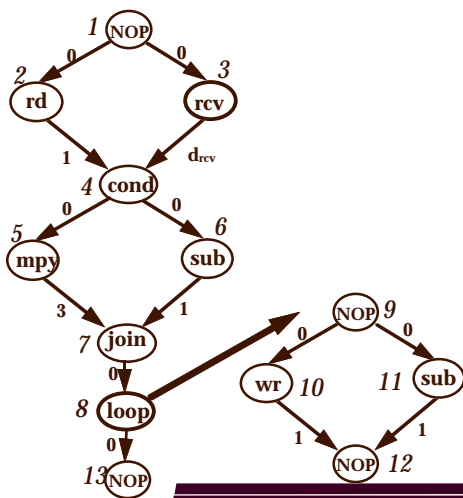


Constraint Graph

anchors??

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Modified Relative Schedule

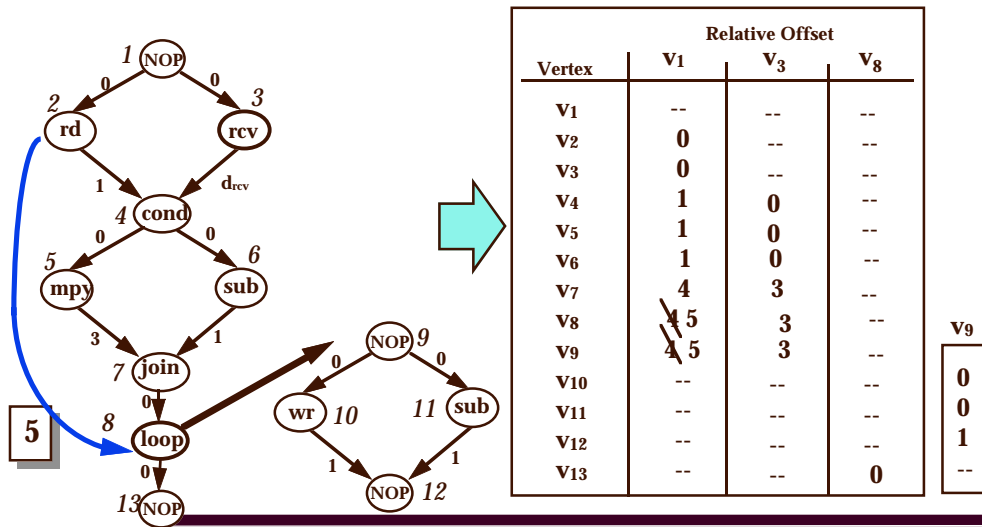


Vertex	Relative Offset		
	v ₁	v ₃	v ₈
v ₁	--	--	--
v ₂	0	--	--
v ₃	0	--	--
v ₄	1	0	--
v ₅	1	0	--
v ₆	1	0	--
v ₇	(2,4)	(1,3)	--
v ₈	(2,4)	(1,3)	--
v ₉	(2,4)	(1,3)	--
v ₁₀	--	--	--
v ₁₁	--	--	--
v ₁₂	--	--	--
v ₁₃	--	--	0

v ₉
0
0
1
--

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Modified Relative Schedule



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Constraint Satisfiability

- For constraint analysis purposes, it is not necessary to determine a schedule of the operations, but only to *verify the existence* of a schedule

Constraint satisfiability \Rightarrow identifying conditions under which no solution (i.e., schedule) exists

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ND operations

1 In the presence of *ND* operations

- ♦ satisfiability analysis attempts to determine the existence of a schedule of operations for all possible (and conceivable) values of the delay of the *ND* operations

☀ Modified relative scheduling - Min/Max Delay Constraints

⇒ A minimum delay constraint is always satisfiable

$$t_k(v_i) \geq \max_{a \in A_b(v_i)} [t_a + \delta(a) + |\theta_a(v_i)|_\infty]$$

For each constraint l_{ij} solution can be constructed such that

$$\theta_{v_j}(v_i) \geq \max(l(v_j, v_i), l_{ij})$$

⇒ A maximum delay constraint may not always be satisfiable

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Satisfiability - Delay Constraints

Feasibility:

⇒ A constraint graph is considered feasible if it contains no positive cycle when the delay of the *ND* operations is assigned to zero.

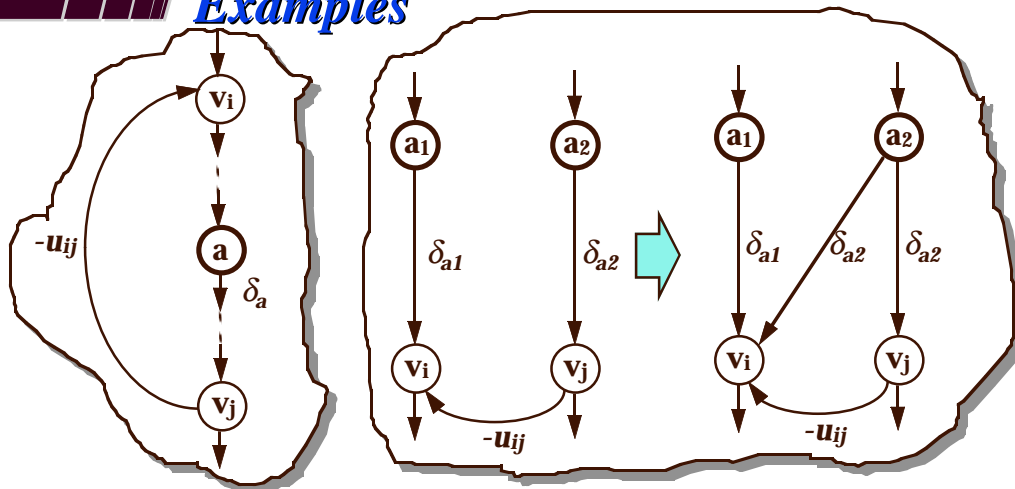
Condition necessary and sufficient to determine the satisfiability of constraints in the presence of *ND* operations:

☀ Operation delay constraints are satisfiable if and only if

→ the constraint graph is feasible
there exists no cycles with *ND* operations

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Examples



Constraints are not satisfiable
(maybe feasible)

can be modified
such that...

Constraints are satisfiable

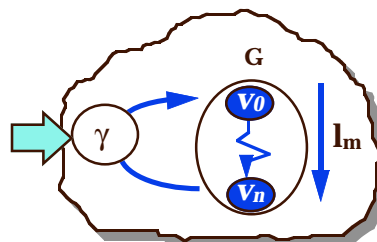
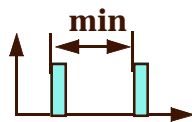
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Max Rate Constraints

A max-rate constraint, R_i , in G is satisfied if

$$I_m(G) \geq R_i^{-1}$$

As with minimum delay constraints, maximum rate constraints are always satisfiable

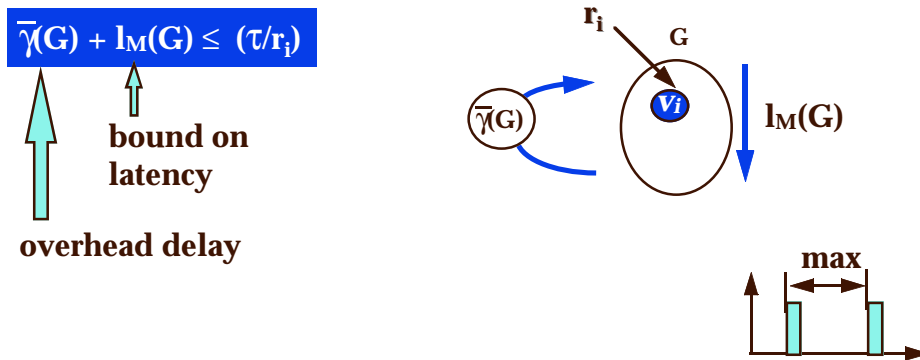


when the lower bound $I_m(G) \leq R_i^{-1}$ the max-rate constraint can still be satisfied by an appropriate choice of *overhead delay* that is applicable to every execution of G

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Min Rate Constraint

→ A minimum rate constraint r_i on an operation $v_i \in V(G)$, where G contains no ND operations is satisfiable if



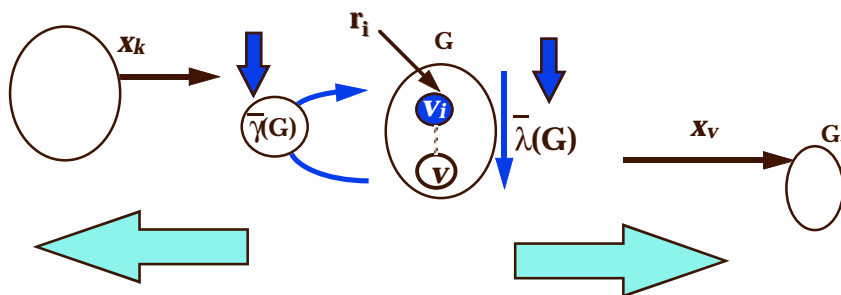
- 1 A minimum rate constraint places an upper bound on the interval of successive executions of an operation

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Min Rate Constraints

→ General case: involves two bounds

$$\bar{\gamma}(G) + \bar{\lambda}(G) \leq (\tau/r_i)$$

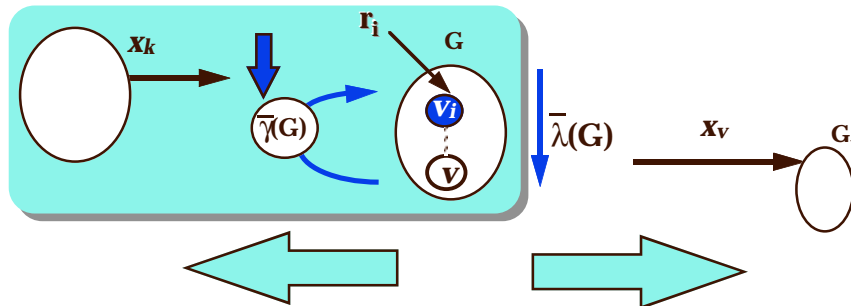


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Min Rate Constraints

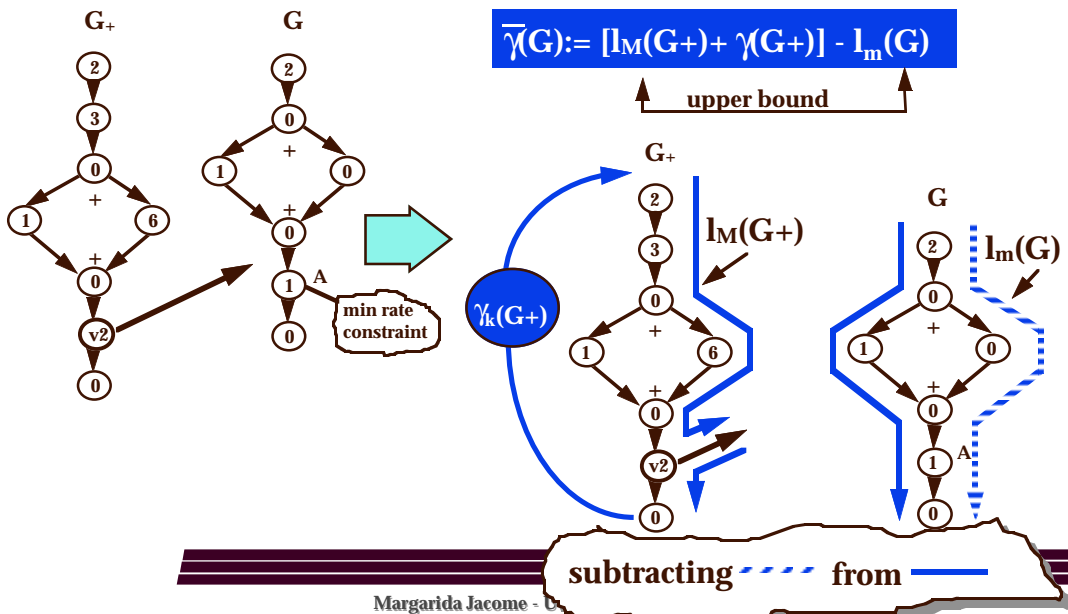
→ General case: involves two bounds

$$\bar{\gamma}(G) + \bar{\lambda}(G) \leq (\tau/r_i)$$



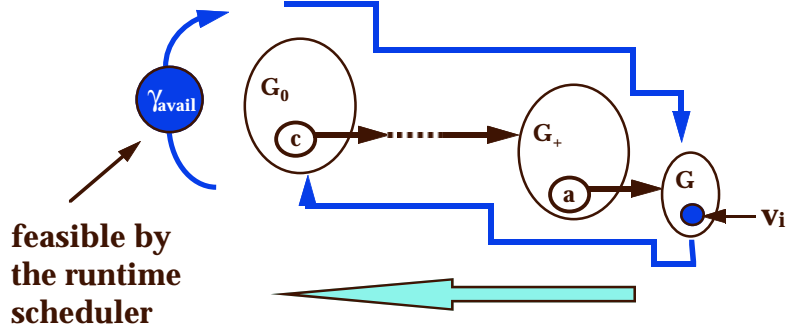
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Upper Bound on Overhead Delay



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Min Rate: satisfiability



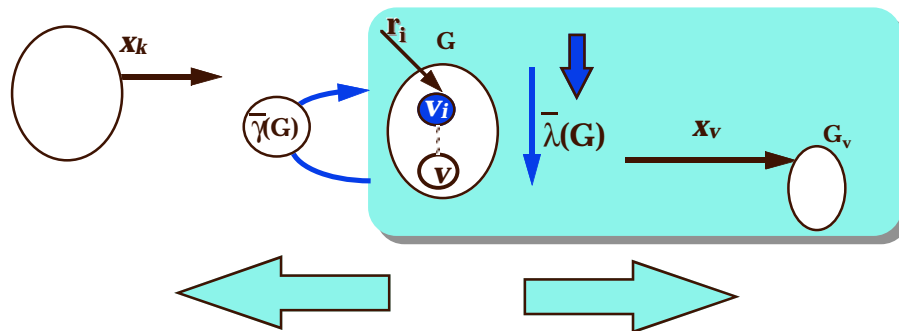
Max delay (min rate) between two executions of v_i occurs when the entire hierarchy is traversed with just one execution of the link operations that lead to v_i .

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Min Rate Constraints

General case: involves two bounds

$$\bar{\gamma}(G) + \bar{\lambda}(G) \leq (\tau/r_i)$$



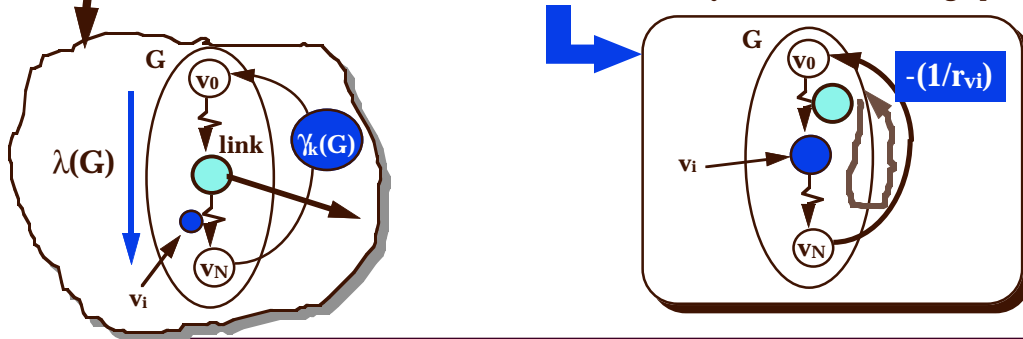
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Min Rate Constraints (with ND operations)

In the presence of ND operations in G:

→ The latency $\bar{\lambda}(G)$ needs to be bounded

- relative rate constraints -- represented as a backward edge (i.e., max delay constraint) from G's sink to source vertices => ND cycle in the constraint graph



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ND Operations: Data-dependent Loops

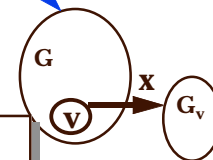
→ The *loop index* determines the number of times the loop body is invoked for each invocation of the loop link operation

=> delay of the loop operation is its loop index times the latency of the loop body

- ⊛ If the constrained graph (G) contains at most one loop operation, v, on a path from source to sink

→ The *minimum rate constraint* can be seen as a bound on the number of times the loop body (G) is invoked.

→ bound on loop index, \bar{x}



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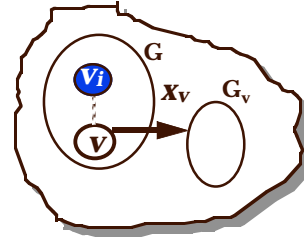
////// Satisfiability of Min Rate Constraints

Consider a flow graph G with an ND operation v representing a loop in the flow graph

➔ A minimum rate constraint r_i on operation $v_i \in V(G)$ and $v_i \neq v$ is satisfiable if the loop index, x_v indicating the number of times G_v is invoked for each execution of v is less than the bound

3 ➔

$$\bar{x}_v := \left\lfloor \frac{\tau \cdot r_i^{-1} - \bar{\gamma}(G) - I_M(G) + \mu(v)}{I_M(G_v)} \right\rfloor + 1$$

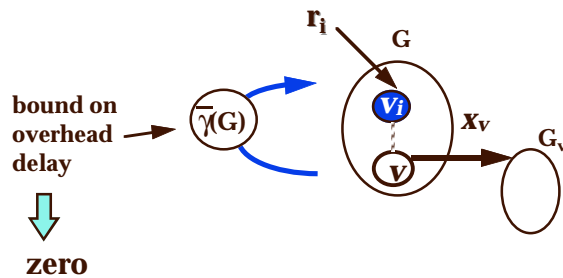


$\mu(v)$ ➔ mobility of operation v
defined as the difference between the longest path that goes through v and I_M

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////// Relative Min Rate Constraints

➔ Relative min rate constraint relative to G ==> applied when G is enabled and executing



$$\bar{x}_v := \left\lfloor \frac{\tau \cdot r_i^{-1} - \bar{\gamma}(G) - I_M(G) + \mu(v)}{I_M(G_v)} \right\rfloor + 1$$

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Constraint Satisfiability in Vulcan

1. Construct the Constraint Graph
 - ◆ add forward edges for *minimum delay* and *maximum rate* constraints
 - ◆ add backward edges for *maximum delay* and (relative) *minimum rate* constraints
2. Identify *cycles* by path enumeration for each of the *backward edges* in the constraint graph
 - ⇒ check for constraint satisfiability, bound delays, etc.
3. Propagate *minimum rate* constraints up the graph hierarchy