

Dataflow Design

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Outline

- * Dataflow basics
 - Pipelining primitives
- * Performance estimation
 - “Canopy Graph” analysis

Dataflow basics

What is Dataflow?

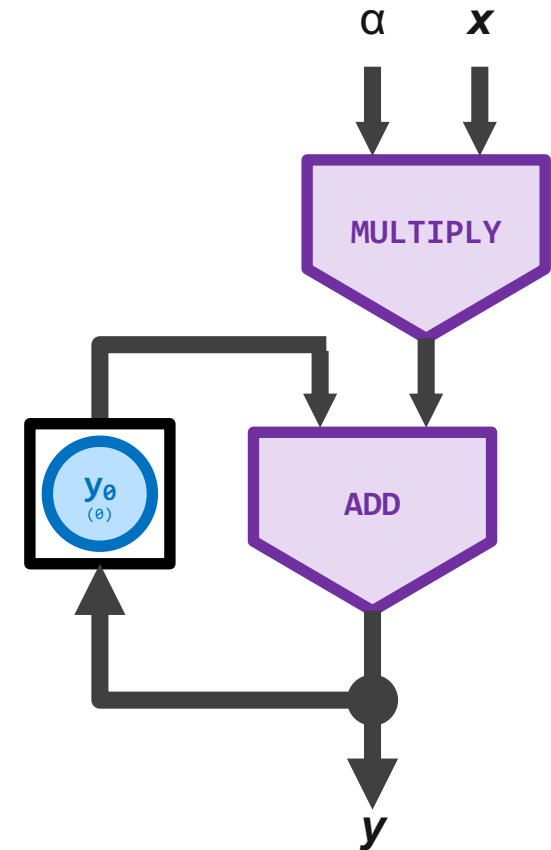
- * Graphical description of operations in a computation
- * Sequencing is determined by data dependencies
 - inputs trigger a function
 - ... instead of an overall control structure
- * Intuitive, natural representation for:
 - data-driven algorithms, e.g. DSPs
 - stream processing
- * Implementation is not necessarily asynchronous
 - but async is often a natural match

Example: multiply-accumulate

Motivation: linear algebra core operation

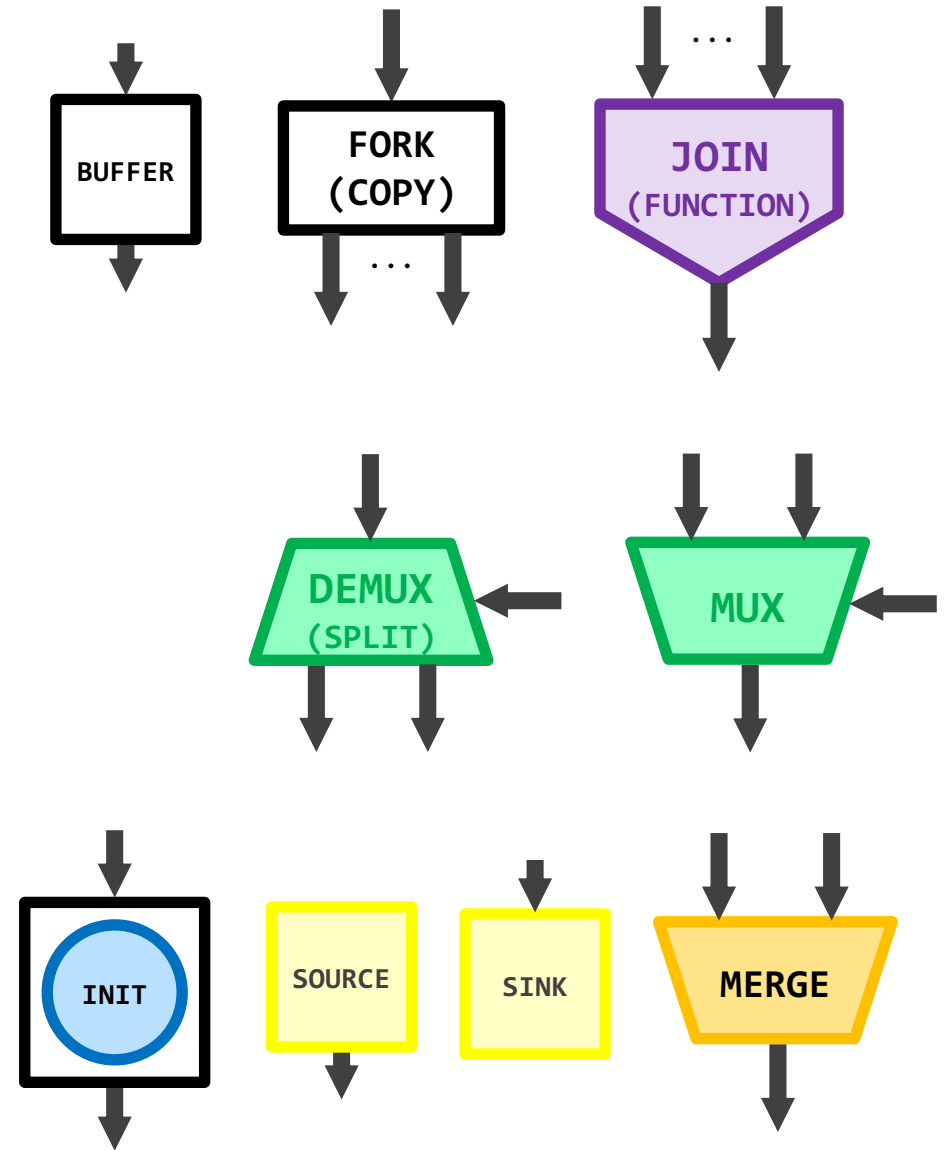
$$\mathbf{y} \leftarrow \alpha \mathbf{x} + \mathbf{y} \quad (\text{SAXPY})$$

If you care about DSP, HPC, AI/deep learning... this is a useful kernel to implement



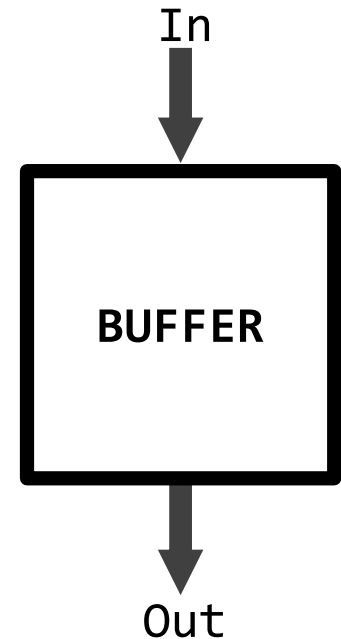
Dataflow primitives

- * Reading from all input channels, writing to all output channels
- * Reading from 1, writing to one-of- N (demux)
- * Reading from one-of- N , writing to 1 (mux / conditional merge)
- * Other misc useful blocks:
 - initialization
 - source/sink
 - merging/arbitration



BUFFER

- * Transmit token from input to output with storage and handshaking flow control
 - one pipeline stage (FIFO stage)
 - latch + handshake control

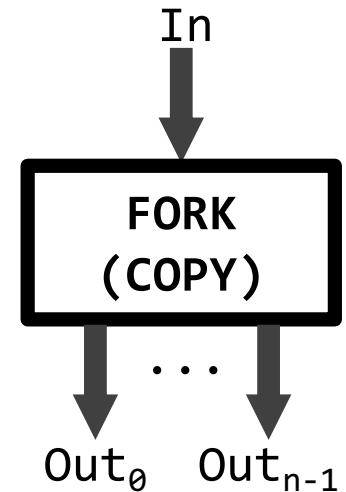


Also known as: slack buffer, one-place FIFO, latch

*[In?x; Out!x]

FORK / COPY

Copy input token to multiple destinations



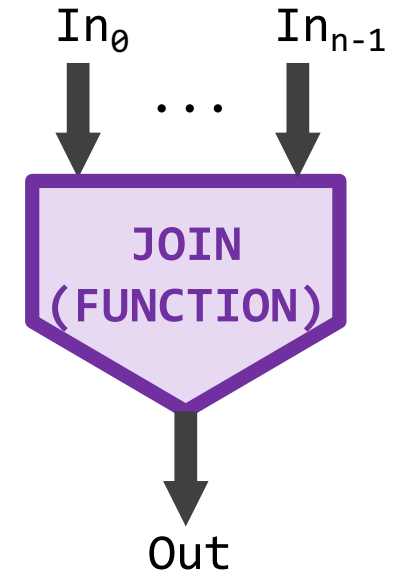
Also known as: n-way link

$*[In?x; Out_0!x, \dots, Out_{n-1}!x]$

JOIN / FUNCTION

Read values from all inputs,
compute result and send on
output

Example functions: arithmetic,
logic, decoding, etc.



```
*[ In0?arg0, In1?arg1, ... , Inn-1?argn-1;  
  Out!func(arg0, arg1, ..., argn-1)  
]
```

Also known as: OPERATOR

Multiplexer (MUX)

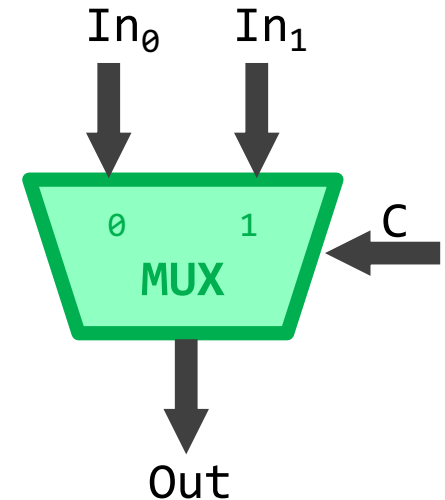
Select one input to send to output based on control signal

- ignore other input (do not consume)
- generalizable to N inputs

Not to be confused with combinational MUX:

- same basic behavior, but this is a dataflow operator
- unused input channel is not consumed; its data is still available

Also known as: controlled merge, conditional join

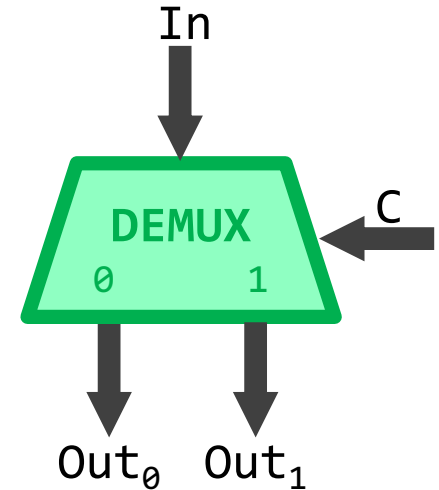


```
*[C?c;  
  [ c=0 -> In_0?x  
  [ ] c=1 -> In_1?x  
  ];  
  Out!x  
]
```

DEMUX

Steer/route input to one of two outputs

- based on value of control signal
- generalizable to N outputs

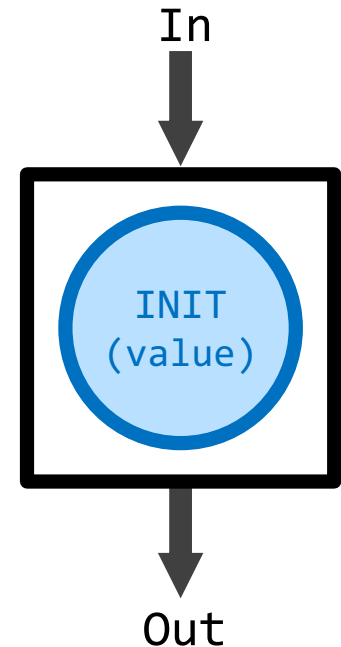


Also known as: SPLIT

```
*[In?x, C?c;  
  [ c=0 -> Out0!x  
  [ ] c=1 -> Out1!x  
  ]  
]
```

Initial token buffer

Send one initial value token, then
behave as a normal buffer

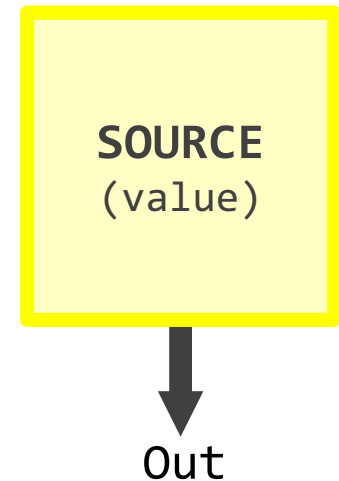


Also known as: INITIALIZER

```
Out!value; *[In?x; Out!x]
```


SOURCE

Repeatedly send tokens with same constant value



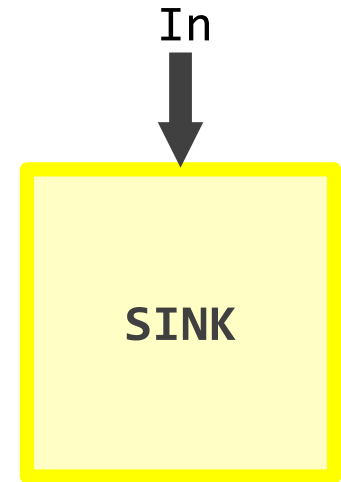
Also known as: bit/token generator

*[Out!value]

SINK

Consume and discard input token

- Not particularly useful by itself, but in combination with other dataflow primitives



Also known as: (bit) bucket

*[In?value]

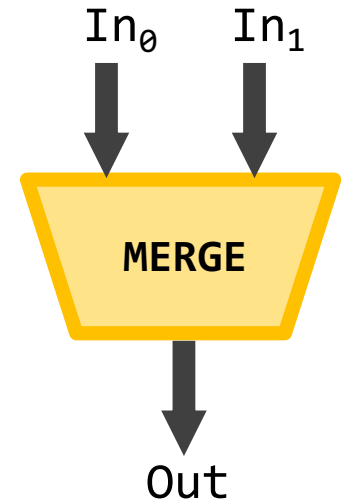
Uncontrolled merge

Combine two input streams to one output

Depending on system design, selection is either:

- **deterministic** – only one input will arrive at a time (ensured by environment)
- **non-deterministic** – requires arbitration to choose if both inputs can arrive close together

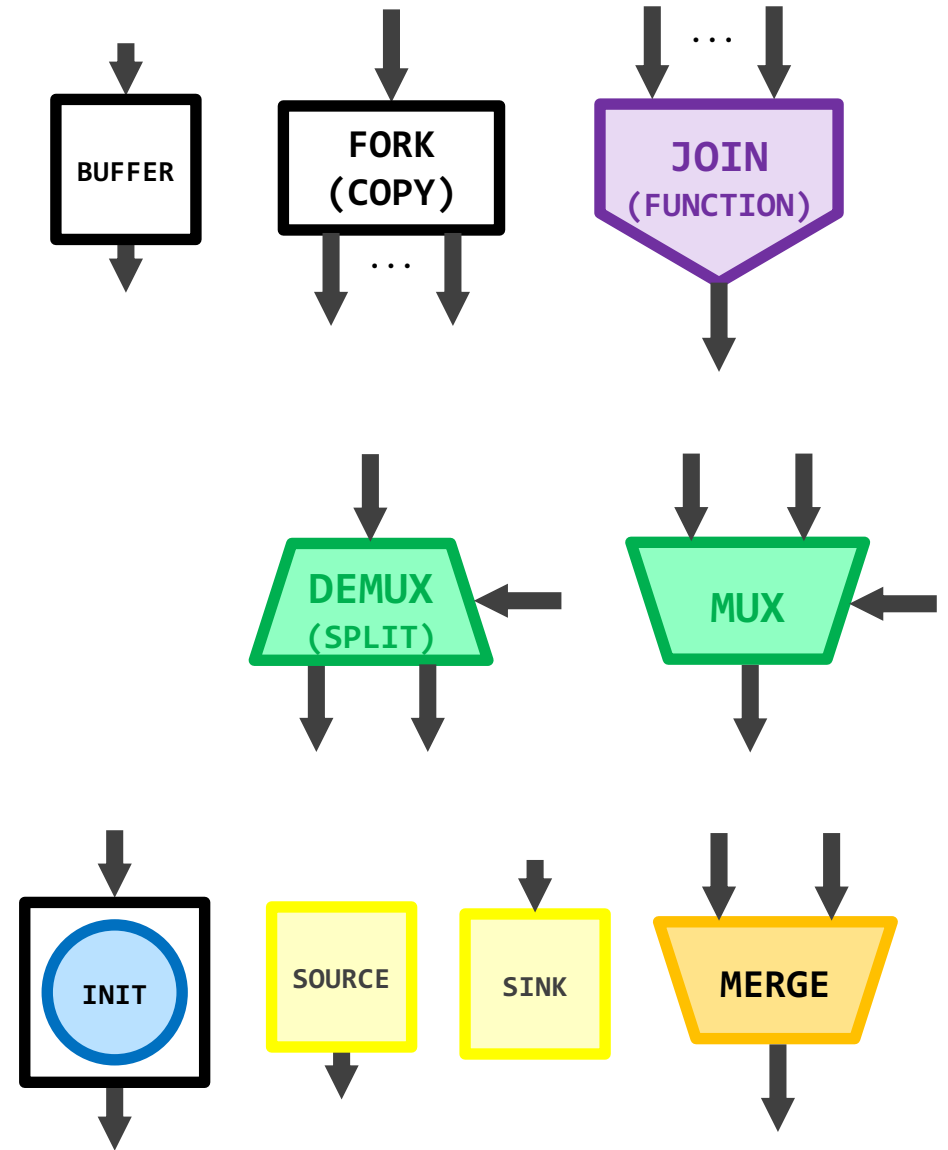
Also known as: MIXER, JOIN



```
*[ [ #In_0 -> In_0?x  
  [] #In_1 -> In_1?x  
  ];  
  Out!x  
]
```

Recap: Dataflow primitives

- * Reading from all input channels, writing to all output channels
- * Reading from 1, writing to one-of- N (demux)
- * Reading from one-of- N , writing to 1 (mux / conditional merge)
- * Other misc useful blocks:
 - initialization
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Some useful design patterns

Wagging or Multithreading

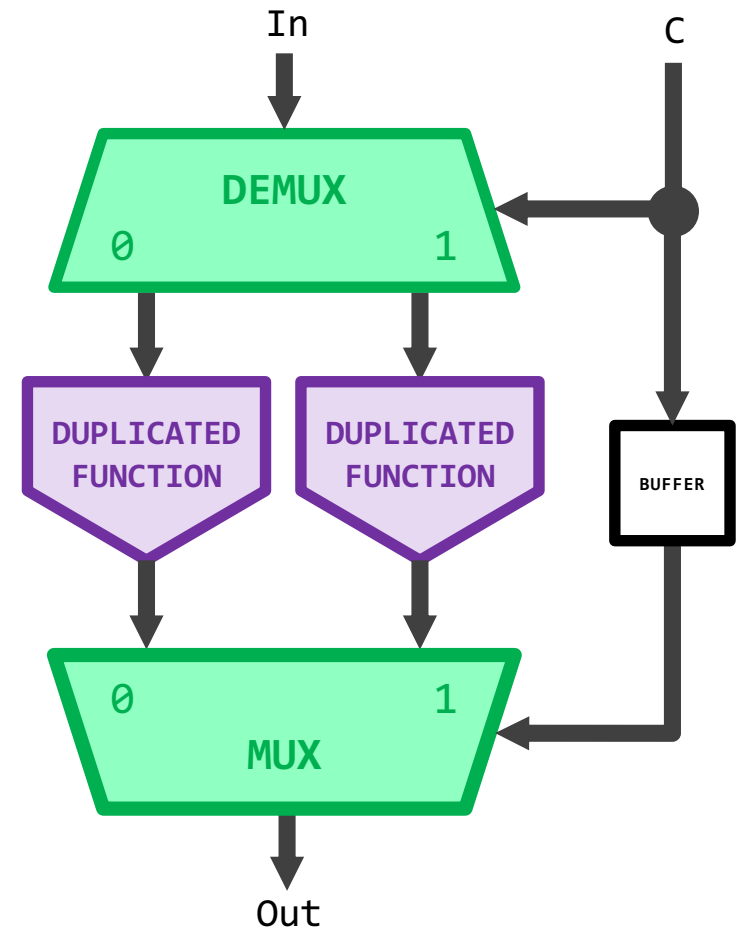
Problem: Slow function block

Solution: Duplicate function block and interleave data between them

→ Improves throughput at the cost of area

Example: large arithmetic block where it is difficult to add internal pipelining

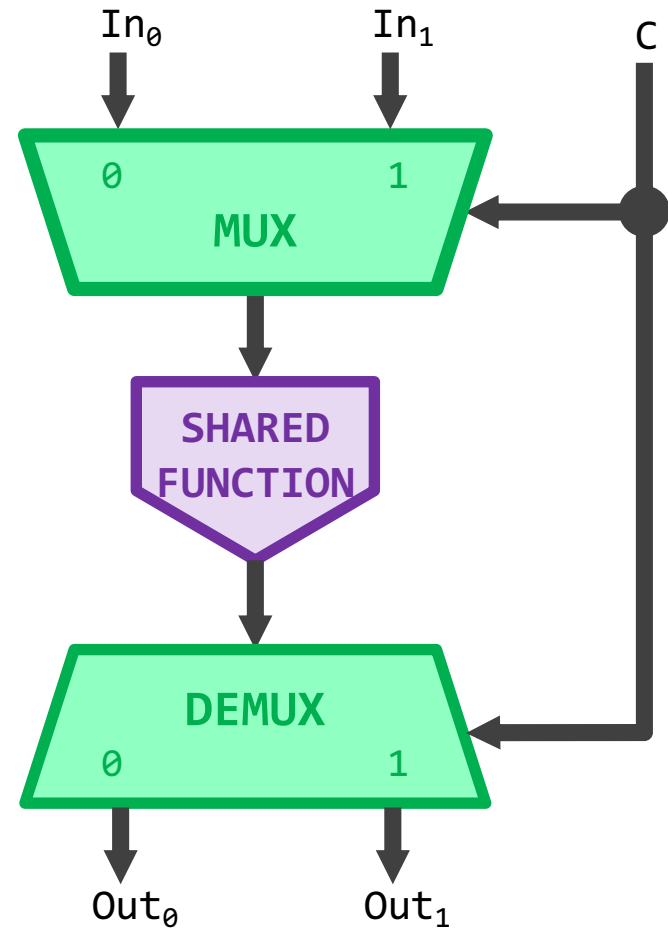
Not just for compute, could also be storage (e.g. tree FIFO)



Resource sharing

Idea: share one expensive or unique resource between multiple users

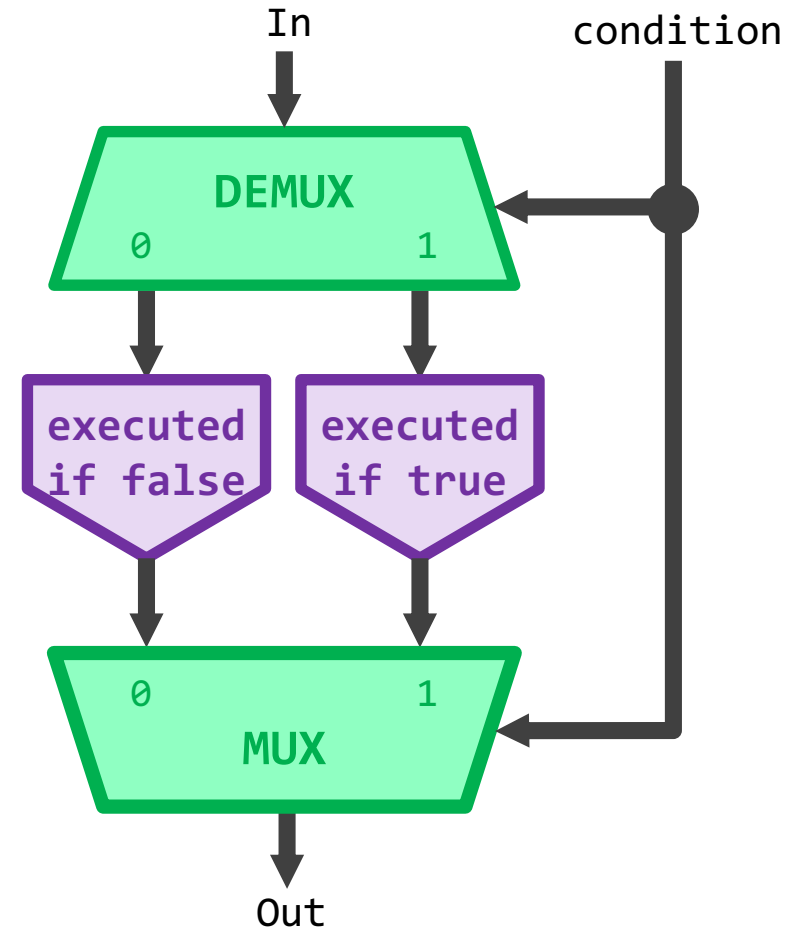
Improves area at the cost of throughput



IF statement

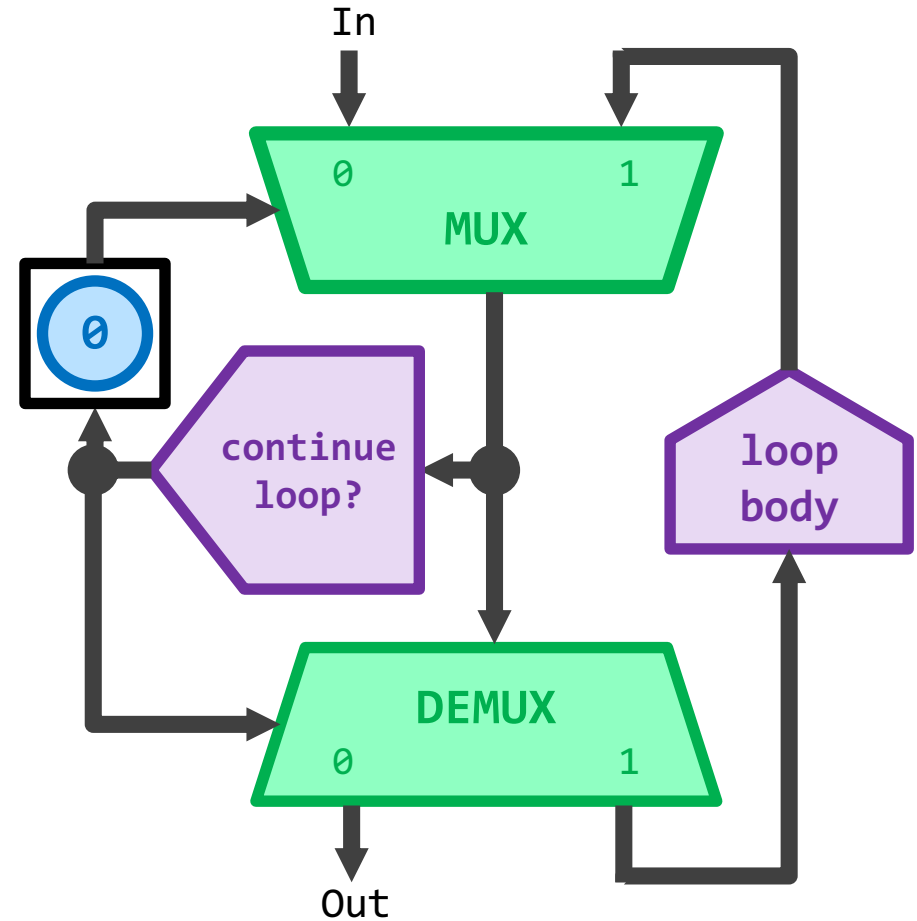
Useful for high-level synthesis

Shown with FUNCTION blocks but can also be other dataflow graphs (e.g. nested IF statements)



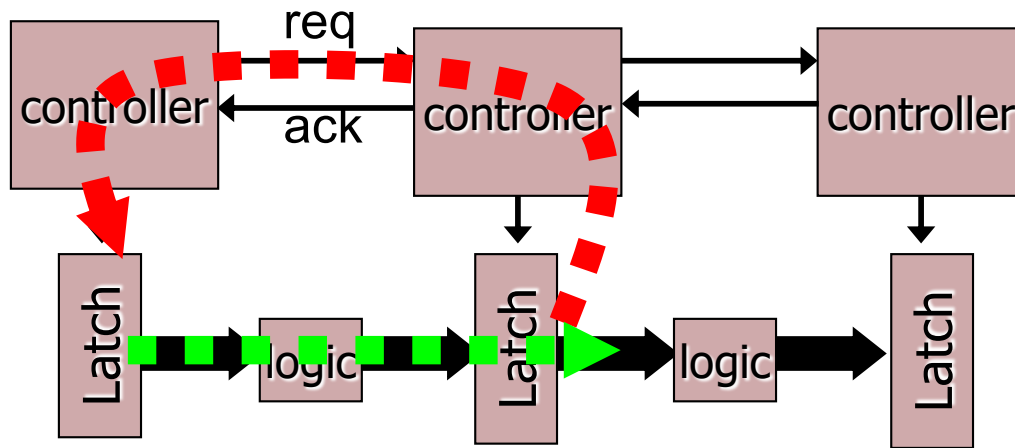
WHILE loop

Can also implement other loop constructs with a similar pattern

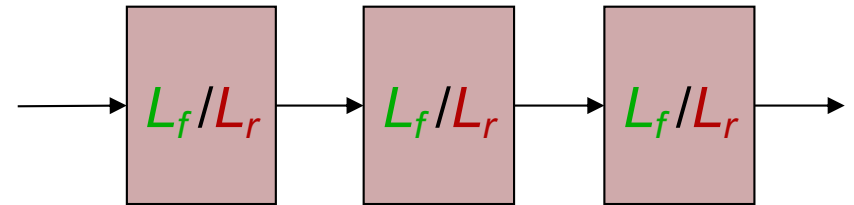


Performance Estimation

Performance Basics: pipeline stages



Cycle time in an asynchronous pipeline



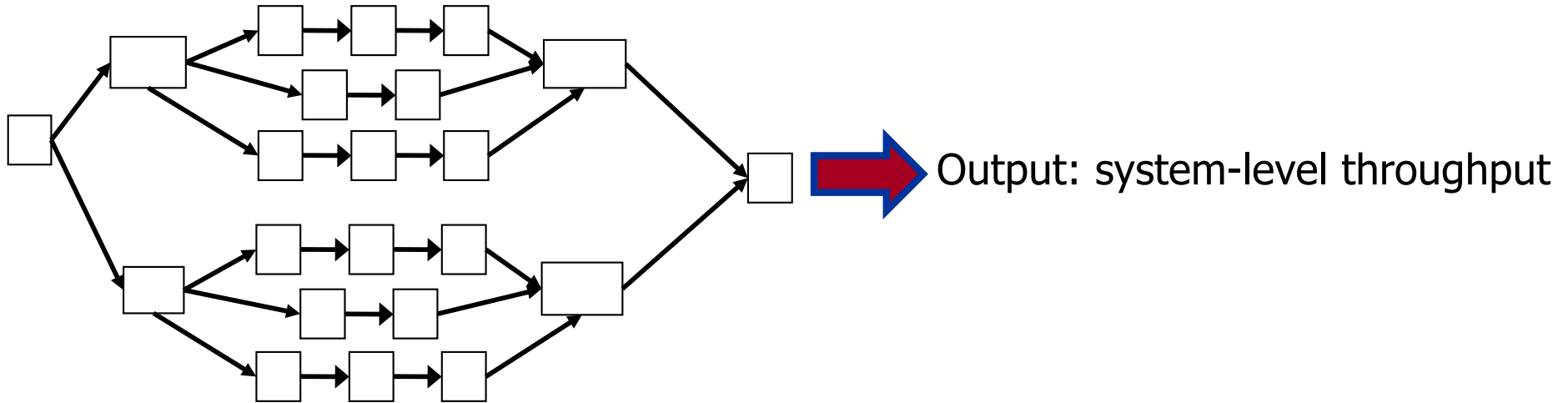
An abstracted view of the pipeline

Each stage characterized by three delays:

- Forward latency, L_f
 - Time for data to propagate forward
- Reverse latency, L_r
 - Time for a stage to receive and process ack
 - Time for a 'hole' to travel backward
- Cycle time, $T = L_f + L_r$
 - Throughput, $tpt = 1 / \text{cycle time}$

Goal

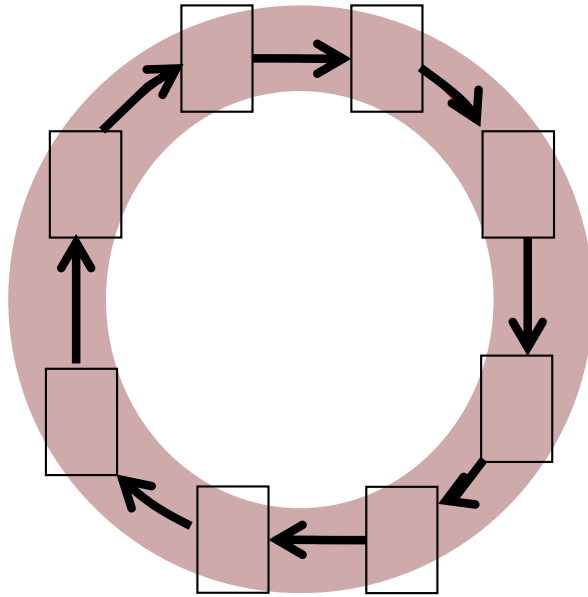
Input: pipelined system-level implementation



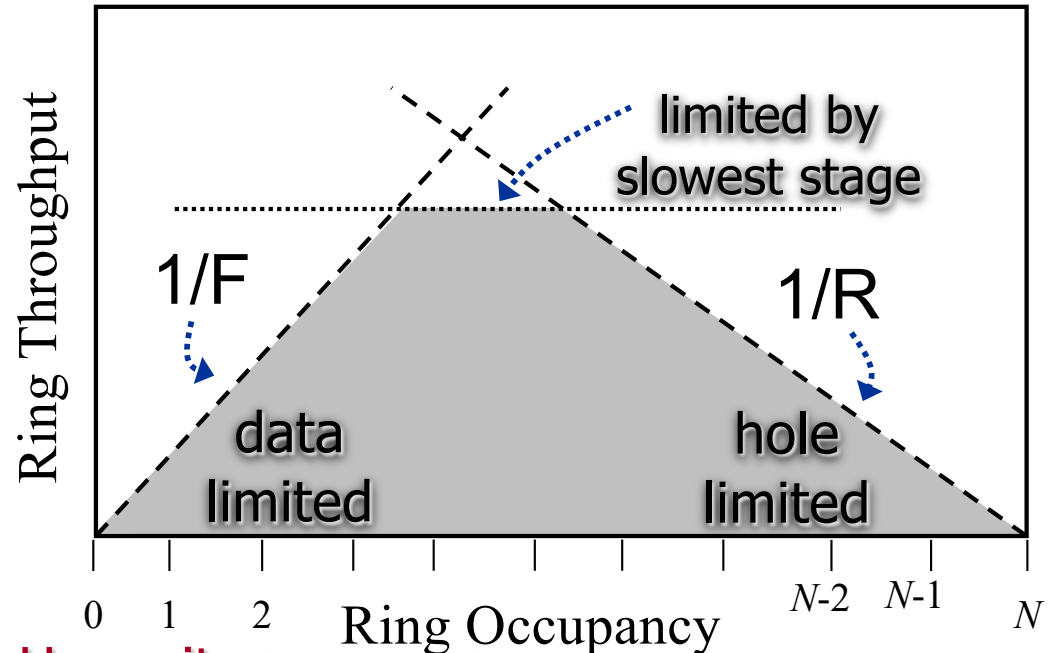
* Motivation: crucial part of an optimizing design flow

- Used repeatedly in an optimization loop
- Requires low runtime and good accuracy

Early work: Pipeline Rings



“Canopy Graph”

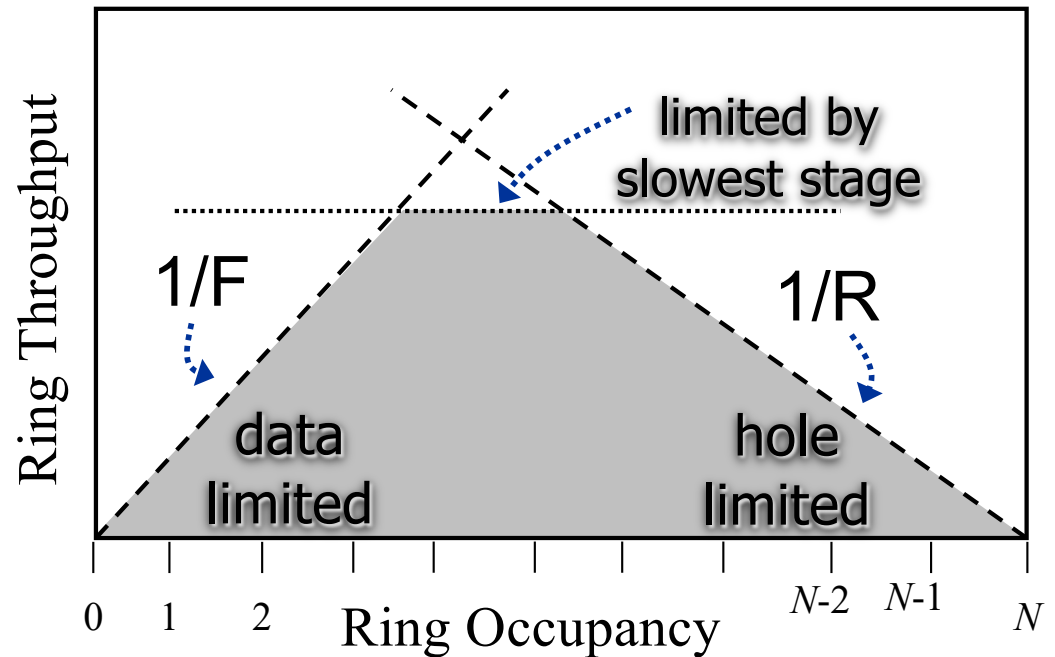
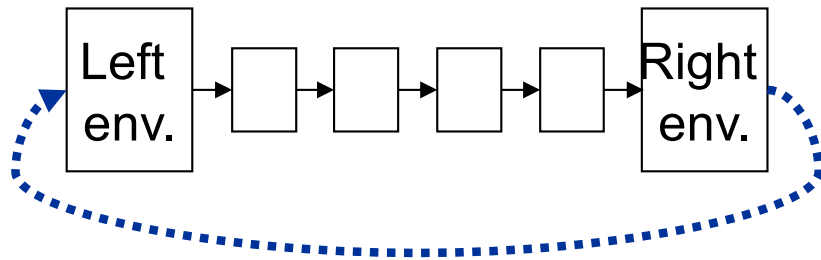


Classic work by T. Williams and M. Horowitz [ISSCC-91]

Ring throughput depends on its occupancy (#items)

- For small number of items: under-utilization limits throughput
- For small number of holes: congestion limits throughput
- Throughput also limited by the **slowest stage**
- Graph is a convex shape: *“Canopy Graph”*
 - [term coined by Singh et al. ASYNC-02 and Gill/Singh ICCAD-08]

Canopy Graphs for linear pipeline



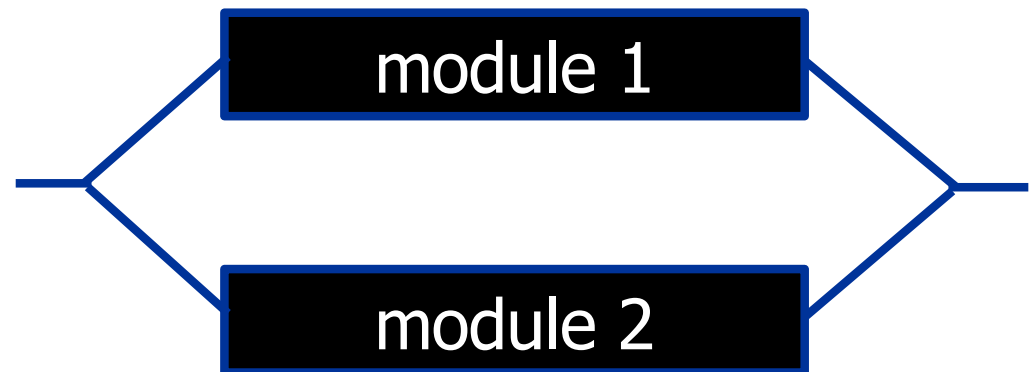
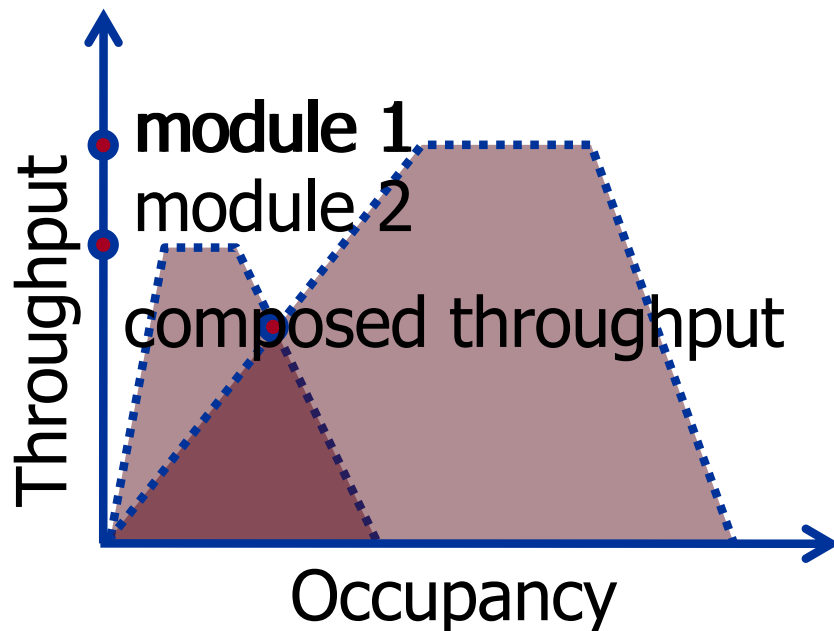
* Canopy graph: also useful approximation for *linear* pipelines

- In steady state: linear pipeline can be modeled as ring

- Rate at which data enters and leaves is identical
- *i.e.* one token leaves → one token enters

Key Idea: Generalize Canopy Graphs

- * **Goal:** Find the system-level throughput for an async dataflow system
 - Use a modular, “divide-and-conquer” method
- * **Challenge:** Throughput is not composable
 - Complex interdependencies dictate throughput
- * **Take problem to higher dimension to make decomposable**
 - One-dimensional throughput is not composable
 - Two-dimensional throughput-occupancy pairs are



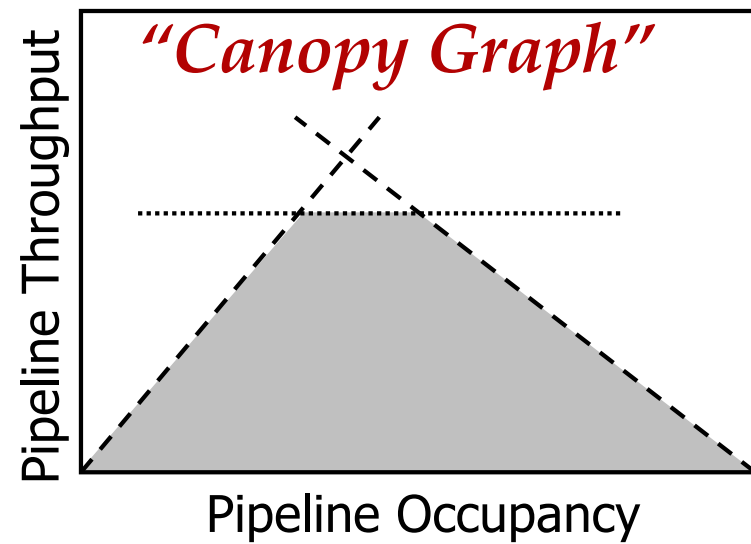
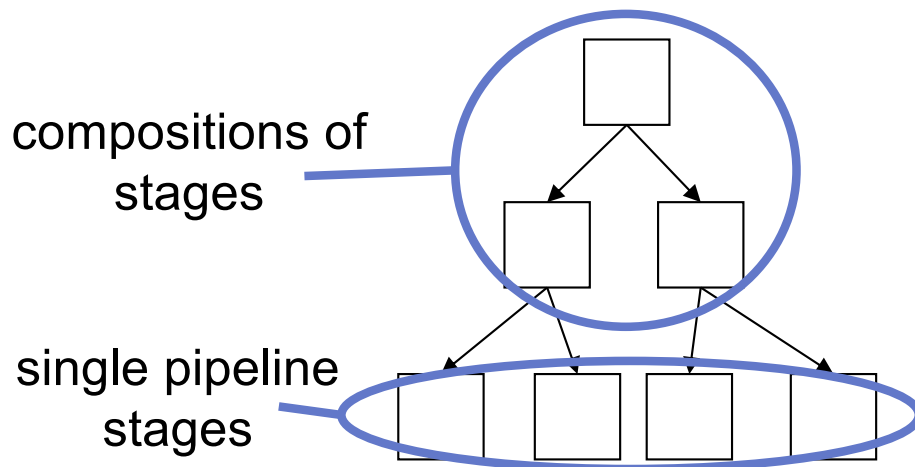
Performance Analysis: Method

* Modular method for performance analysis

- Exploits system hierarchy with “divide-and-conquer” method
- First: calculate canopy graph at each leaf node
 - Each leaf node is a single stage
- Next: compose canopy graphs at each level of the hierarchy
- Finally: canopy graph for root node gives system-level performance

* Requires composition algorithm for common circuit structures

- **Parallel, sequential, conditional, and iterative**



1) Parallel Composition

Parallel Composition of A and B



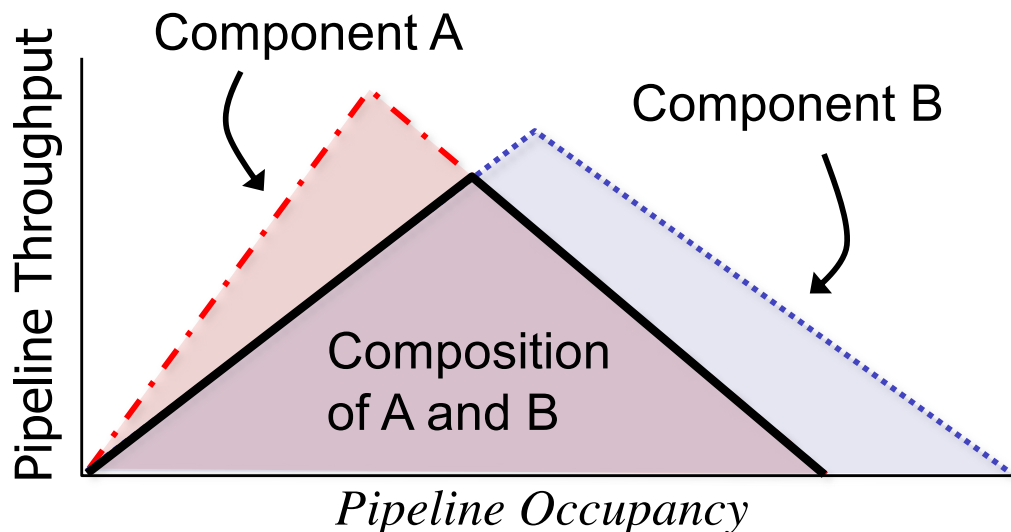
Parallel structures [Lines98]

- Data copied at fork
- A and B compute in parallel
- Results recombined at join

Operation invariants under composition:

- 1) # of items in each branch equal
- 2) Branches have same throughput

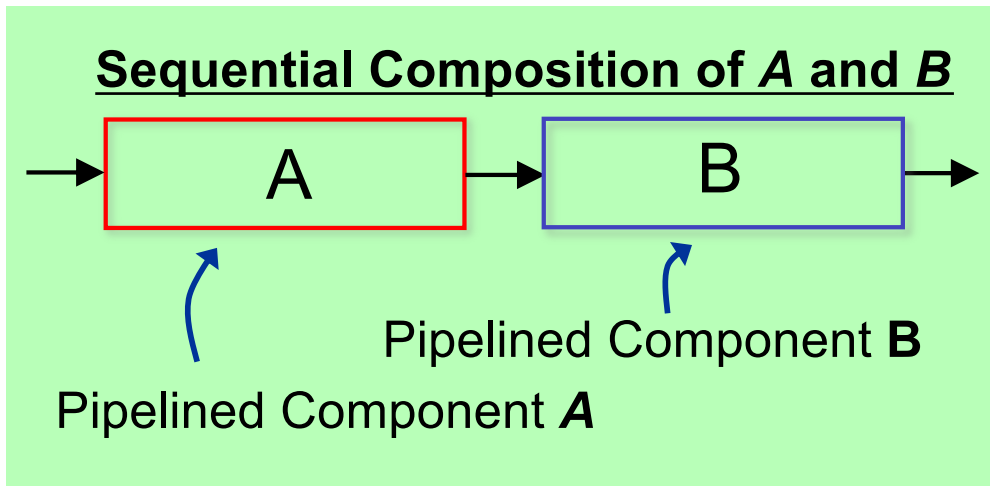
Canopy Graph of Composed Structure



Throughput of structure:

- Intuition: at each occupancy, throughput limited by slower branch
- ⇒ Intersection of canopy graphs of A and B

2) Sequential Composition

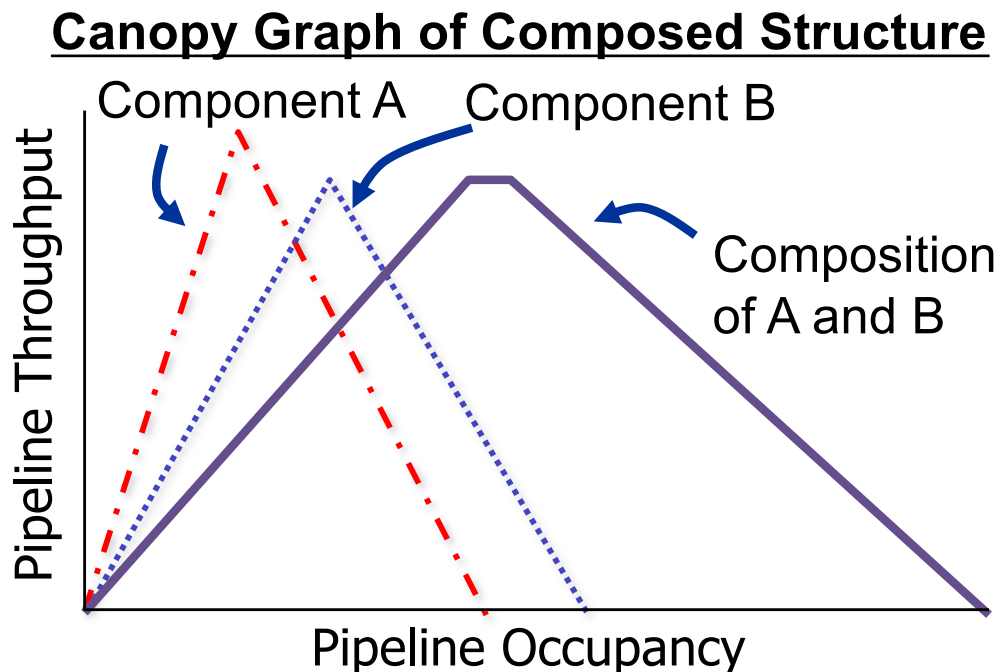


Sequential Structures [Lines98]

- Data transmitted through A, then through B

Operation invariants under composition:

- 1) Find total # items: sum of # items in both pipes
- 2) Throughput A = Throughput B
- 3) Max throughput: limited by slower pipeline

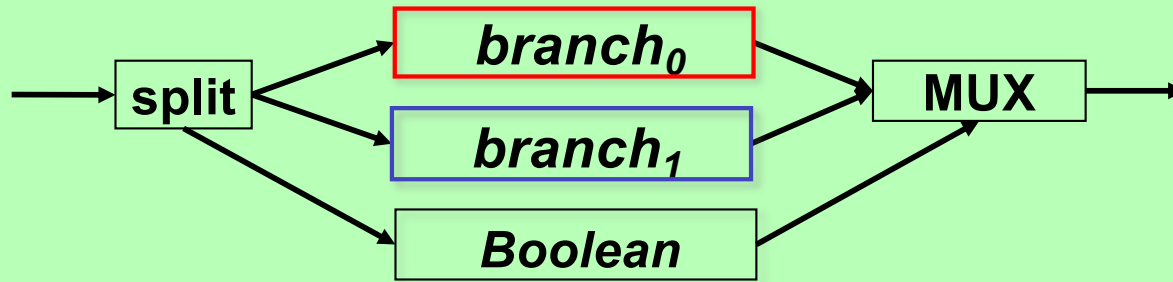


Throughput of structure:

- ⇒ "horizontal sum" of canopy graphs of A and B
- At each throughput, add the occupancies of the two pipelines

3) Conditional Composition

Conditional Composition of $branch_0$ and $branch_1$



Example:

$p_0 = 2/3$ and $p_1 = 1/3$

2 items enter $branch_0$



1 item enters $branch_1$

* Operation invariants under composition:

- Ratio of # items in each branch = ratio of probabilities

$$\frac{Occupancy_0}{p_0} = \frac{Occupancy_1}{p_1}$$

- Ratio of throughput of each branch = ratio of probabilities

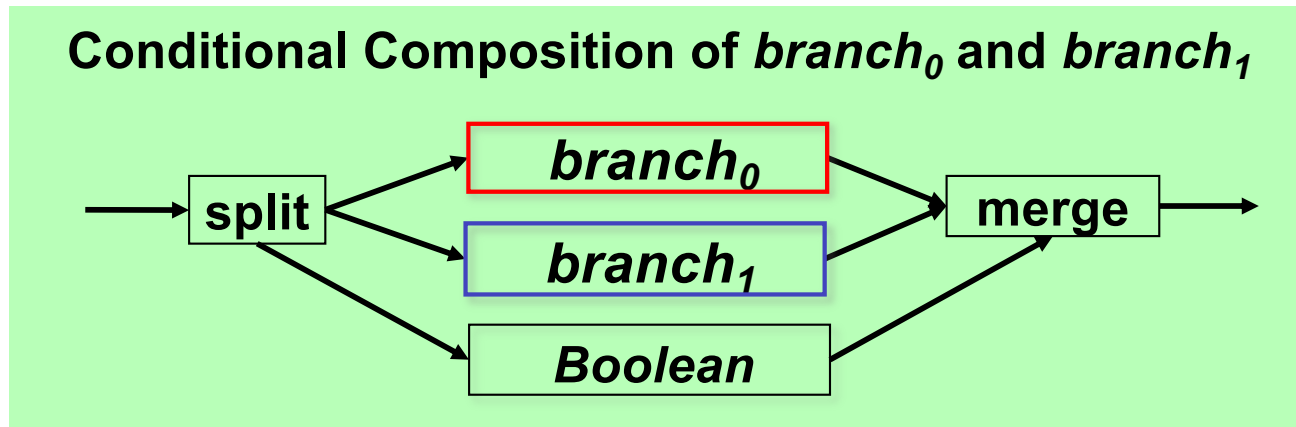
$$\frac{TPT_0}{p_0} = \frac{TPT_1}{p_1}$$

* Throughput of conditional structure:

- Divide each branch's canopy graph by its probability p_i
- Compute intersection of scaled canopy graphs

* "Bursty" inputs cause additional bottlenecks (see ICCAD-08 paper for details)

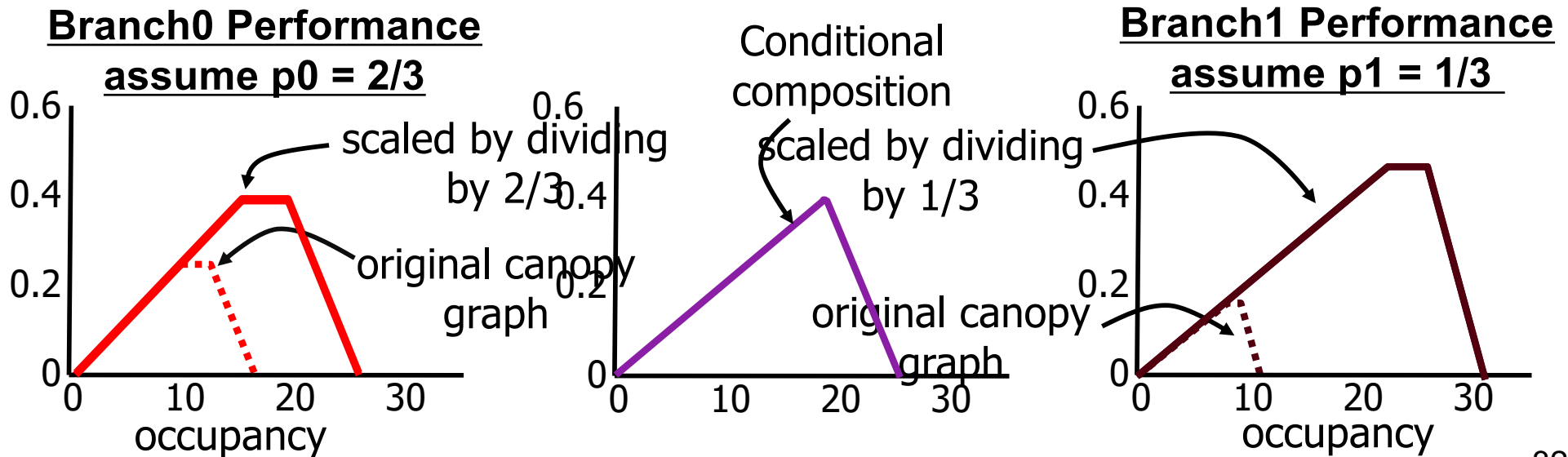
3) Conditional Composition (cont'd)



➔ **Step 1) uniform scaling:** enlarge each branch's canopy graph

Example: $p_0 = 2/3$ and $p_1 = 1 - p_0 = 1/3$

➔ **Step 2) intersection:** finds system-level performance



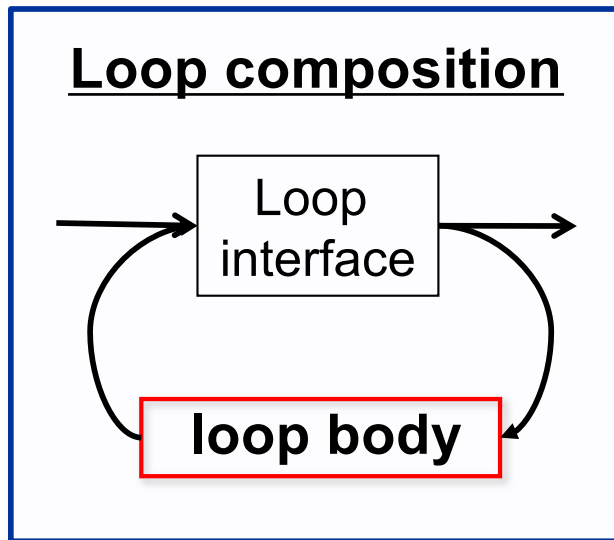
4) Iterative Loop Composition

* Operation invariants under composition:

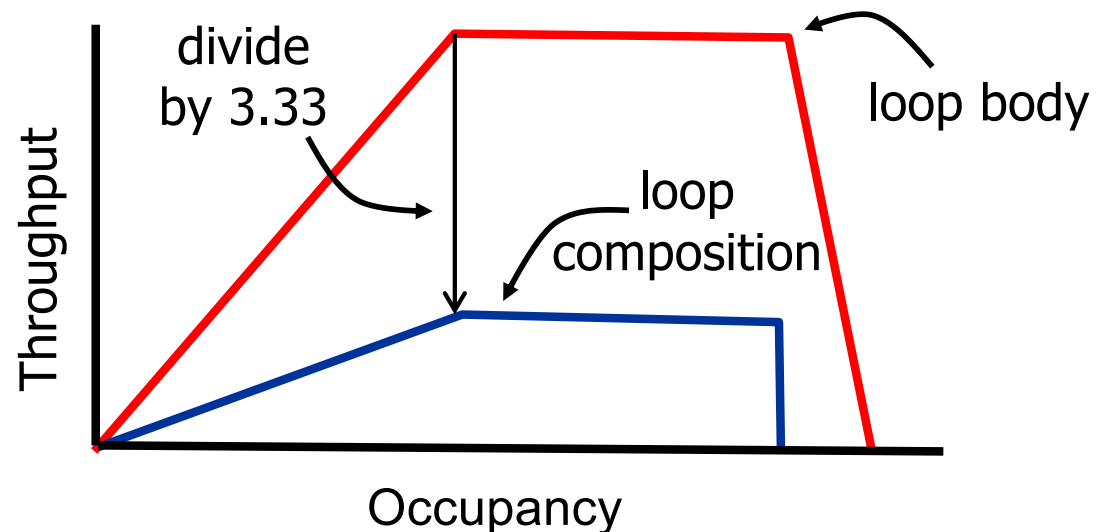
- Each item passes through the loop multiple times
- Loop can handle multiple items simultaneously

* Throughput of composition

- # data items processed decreases as iteration count increases
- ⇒ Scale down based on expected number of iterations



Throughput vs. Occupancy with Expected Iterations = 3.33



Analysis: Benchmark Examples

- * **Analysis algorithm demonstrated on 8 benchmarks**
 - Chosen to represent a variety of circuit constructs

Example	Composition Type			
	Parallel	Sequential	Conditional	Iteration
CORDIC	✓	✓	✓	
CRC		✓	✓	
DIFFEQ	✓	✓		✓
GCD		✓		✓
Ray-tracing		✓	✓	✓
MULT	✓	✓		
JPEG	✓	✓	✓	✓

- * **Evaluated several circuit implementations of some**
 - Naive implementation vs. hand-optimized version
 - Different choice models: uniform random vs. correlated

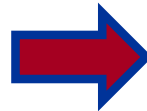
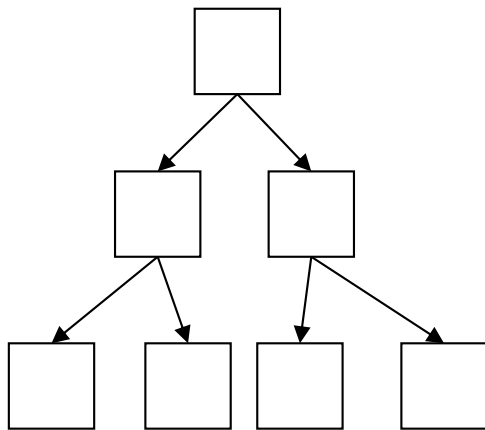
Performance Analysis: Results

- * Total of 12 different circuit examples tested
 - Error < 4% for all examples, runtime ≤ 10 ms for all examples

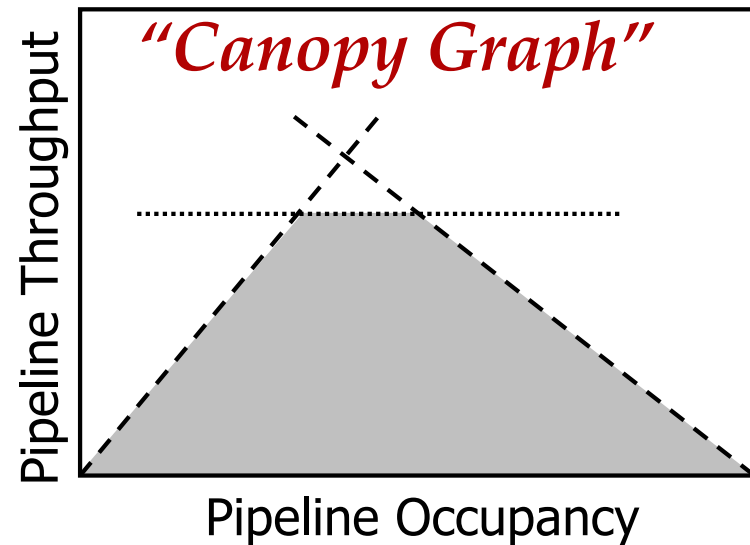
Example	Version	Size # stages	Throughput		Error (%)	Runtime (ms)
			simulated	predicted		
CORDIC	<i>original</i>	31	90.9	90.9	0.00	~10
	<i>optimized</i>	44	167	167	0.00	~10
	<i>bursty inputs</i>	44	83	83	0.00	~10
CRC	<i>original</i>	23	292	286	2.05	~10
	<i>optimized</i>	27	352	357	1.42	~10
	<i>bursty inputs</i>	27	305	300	1.64	~10
DIFFEQ	<i>original</i>	10	18.3	18.2	0.55	<10
GCD	<i>original</i>	21	49	50	2.04	~10
Ray-tracing	<i>original</i>	21	161	167	3.73	~10
	<i>optimized</i>	166	222	222	0.00	~10
MULT	<i>original</i>	13	38.7	38.4	0.78	<10
	<i>optimized</i>	21	167	167	0.00	<10

Performance Analysis: Summary

circuit hierarchy



system-level performance



- * **Fast:** restriction to hierarchical systems yielded fast runtimes
 - Divide-and-conquer approach with linear runtime
 - Modular canopy graph analysis for many constructs
 - Sequential, parallel, conditional, and loop
 - Expressive subset: modeled real-world applications
 - CORDIC, CRC, ray intersection algorithm, etc.
- * **Accurate:** tested on several many non-trivial examples
 - Throughput estimates within 4% of simulation results

References

- * Gennette Gill. Analysis and Optimization for Pipelined Asynchronous Systems. PhD thesis. UNC Chapel Hill. 2010.
- * Gennette Gill and Montek Singh. *"Performance Estimation and Slack Matching for Pipelined Asynchronous Architectures with Choice."* ICCAD 2008.
- * Montek Singh and Steven Nowick. *"MOUSETRAP: Ultra-High-Speed Transition-Signaling Asynchronous Pipelines."* ICCD 2001.
- * Montek Singh and Steven Nowick. *"MOUSETRAP: High-Speed Transition-Signaling Asynchronous Pipelines."* TVLSI 2007.
- * Gennette Gill, J. Hansen, A. Agiwal, L. Vicci and M, Singh. *"A High-Speed GCD Chip: A Case Study in Asynchronous Design."* ISVLSI 2009.