

A Stream Compiler for Communication-Exposed Architectures

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The Streaming Domain

- Widely applicable and increasingly prevalent
 - Embedded systems
 - Cell phones, handheld computers, DSP's
 - Desktop applications
 - Streaming media
 - Software radio
 - Real-time encryption
 - Graphics packages
 - High-performance servers
 - Software routers (Example: Click)
 - Cell phone base stations
 - HDTV editing consoles
- Based on audio, video, or data stream
 - Predominant data types in the current data explosion

Properties of Stream Programs

- A large (possibly infinite) amount of data
 - Limited lifetime of each data item
 - Little processing of each data item
- Computation: apply multiple filters to data
 - Each filter takes an input stream, does some processing, and produces an output stream
 - Filters are independent and self-contained
- A regular, static computation pattern
 - Filter graph is relatively constant
 - A lot of opportunities for compiler optimizations

StreamIt: A spatially-aware Language & Compiler

- A language for streaming applications
 - Provides high-level stream abstraction
- Breaks the Von Neumann language barrier
 - Each filter has its own control-flow
 - Each filter has its own address space
 - No global time
 - Explicit data movement between filters
 - Compiler is free to reorganize the computation
- Spatially-aware Compiler
 - Intermediate representation with stream constructs
 - Provides a host of stream analyses and optimizations

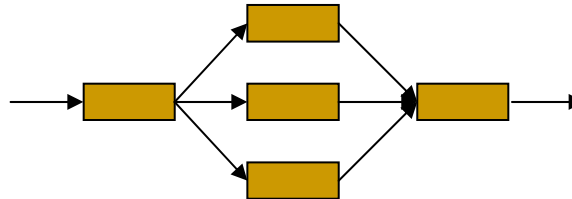
Structured Streams

- Hierarchical structures:

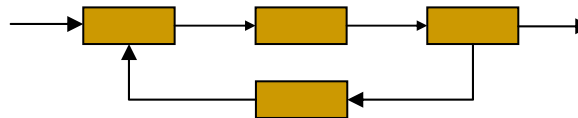
- Pipeline



- SplitJoin



- Feedback Loop

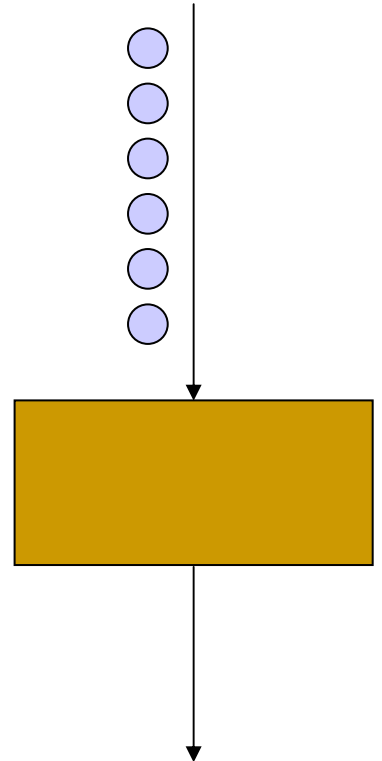


- Basic programmable unit: Filter



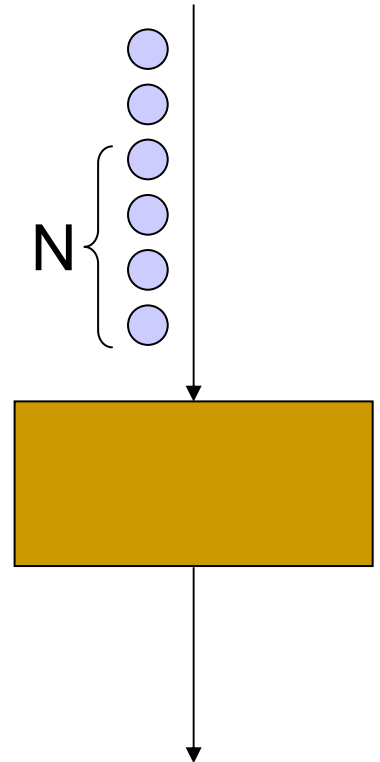
Filter Example: LowPassFilter

```
float->float filter LowPassFilter(int N) {  
    float[N] weights;  
  
    init {  
        for (int i=0; i<N; i++)  
            weights[i] = calcWeights(i);  
    }  
  
    work push 1 pop 1 peek N {  
        float result = 0;  
        for (int i=0; i<N; i++)  
            result += weights[i] * peek(i);  
        push(result);  
        pop();  
    }  
}
```



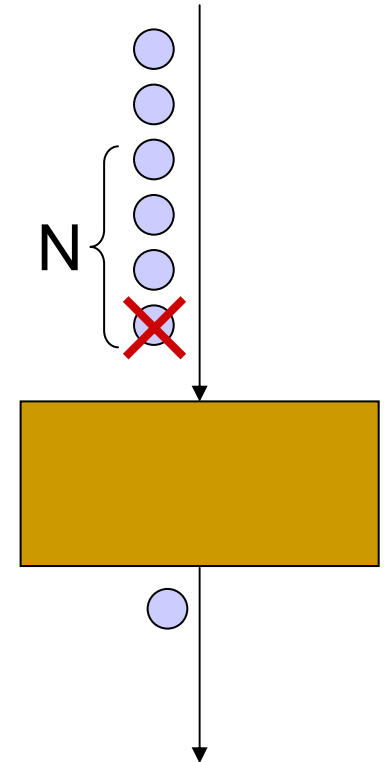
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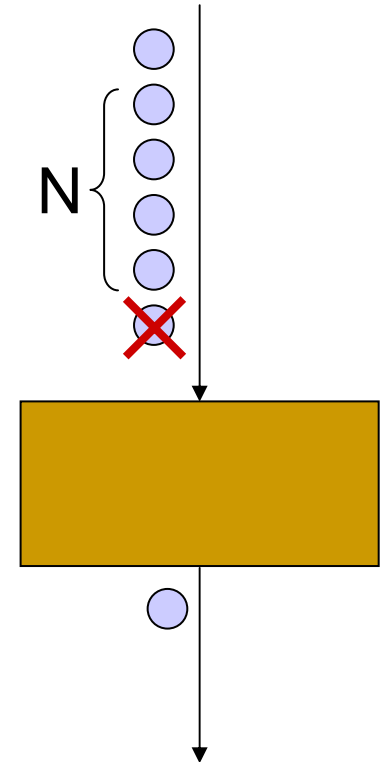
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```



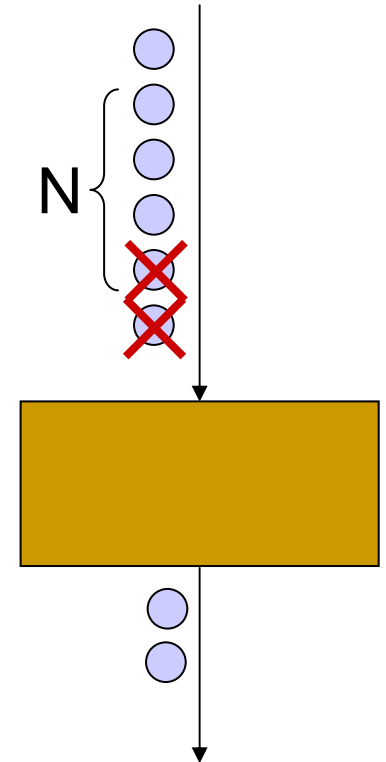
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Filter Example: LowPassFilter

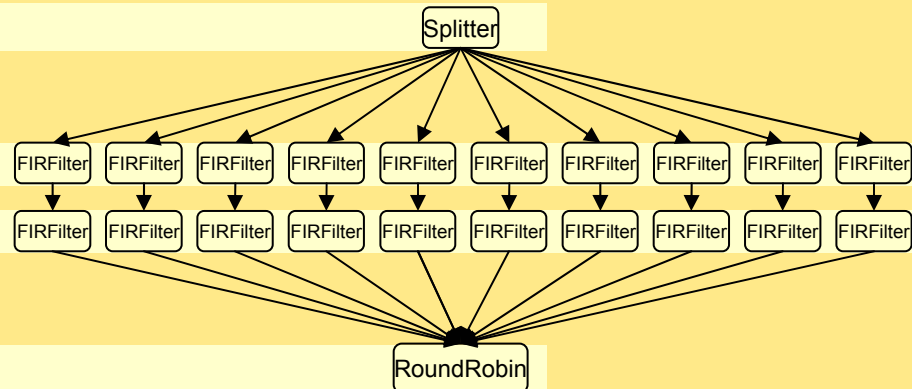
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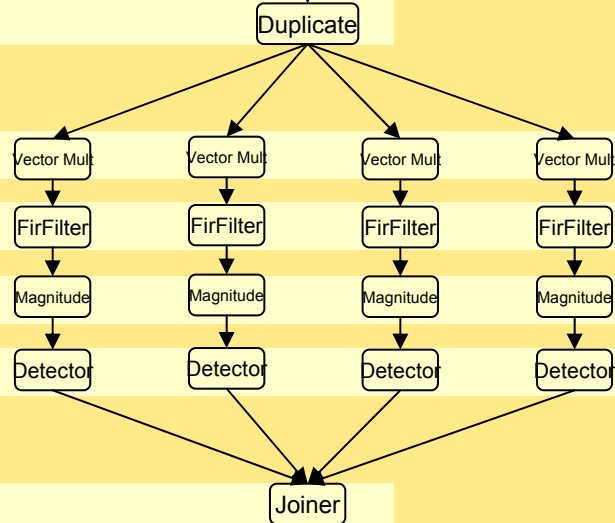
Example: Radar Array Front End

```
complex->void pipeline BeamFormer(int numChannels, int numBeams)
```

```
add splitjoin {  
  split duplicate;  
  for (int i=0; i<numChannels; i++) {  
    add pipeline {  
      add FIR1(N1);  
      add FIR2(N2);  
    };  
  };  
  join roundrobin;  
};
```



```
add splitjoin {  
  split duplicate;  
  for (int i=0; i<numBeams; i++) {  
    add pipeline {  
      add VectorMult();  
      add FIR3(N3);  
      add Magnitude();  
      add Detect();  
    };  
  };  
  join roundrobin(0);  
};
```

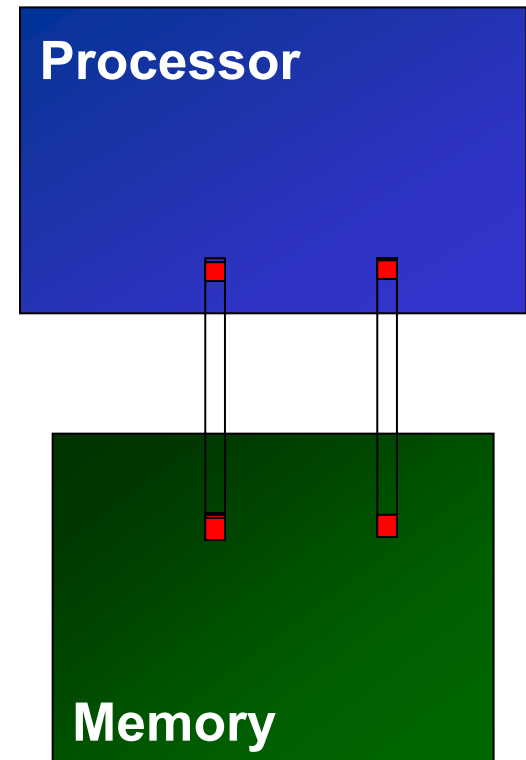


```
}
```

How to execute a Stream Graph?

Method 1: Time Multiplexing

- Run one filter at a time
- Pros:
 - Scheduling is easy
 - Synchronization from Memory
- Cons:
 - If a filter run is too short
 - Filter load overhead is high
 - If a filter run is too long
 - Data spills down the cache hierarchy
 - Long latency
 - Lots of memory traffic
 - Bad cache effects
 - Does not scale with spatially-aware architectures



How to execute a Stream Graph?

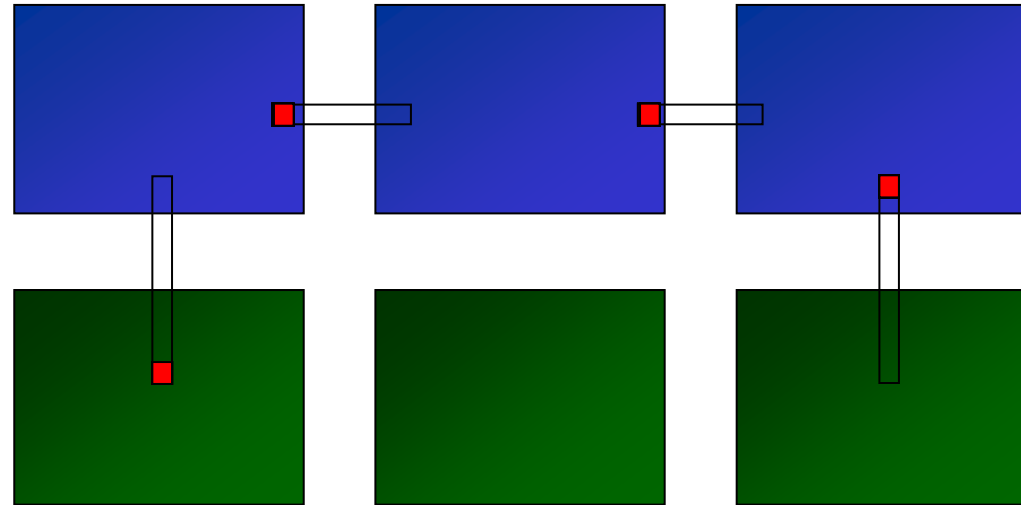
Method 2: Space Multiplexing

- Map filter per tile and run forever



- Pros:

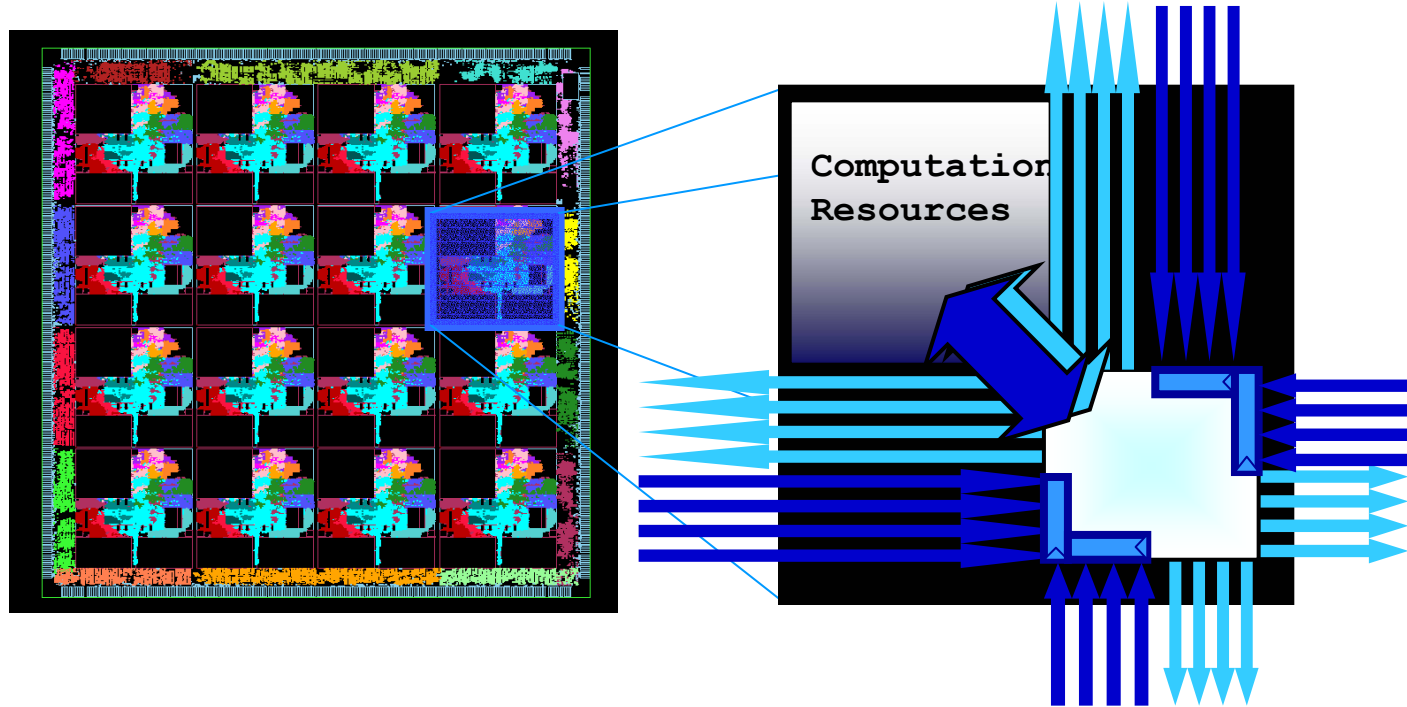
- No filter swapping overhead
- Exploits spatially-aware architectures
 - Scales well
- Reduced memory traffic
- Localized communication
- Tighter latencies
- Smaller live data set



- Cons:

- Load balancing is critical
- Not good for dynamic behavior
- Requires $\# \text{ filters} \leq \# \text{ processing elements}$

The MIT RAW Machine

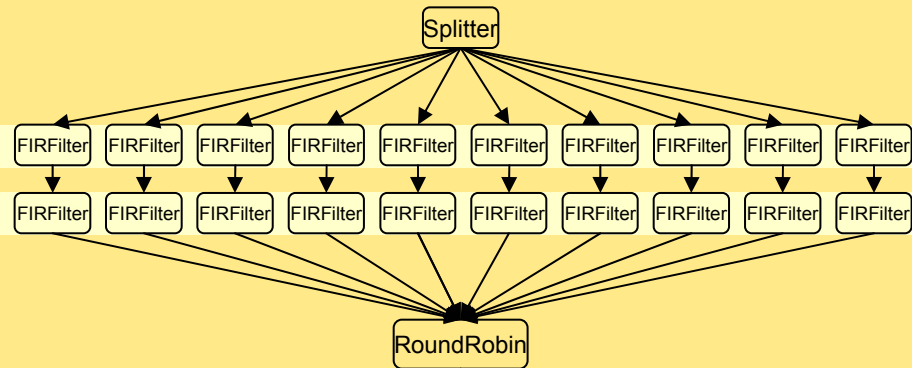


- A scalable computation fabric
 - 4 x 4 mesh of tiles, each tile is a simple microprocessor
- Ultra fast interconnect network
 - Exposes the wires to the compiler
 - Compiler orchestrate the communication

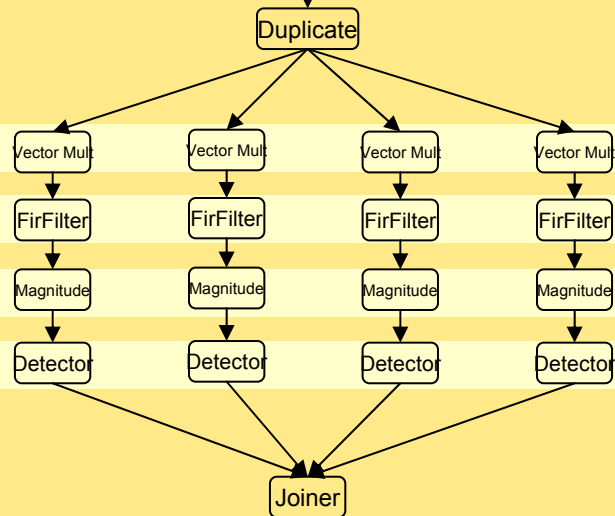
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        add FIR2(N2);  
      };  
    };  
    join roundrobin;  
  };  
};
```

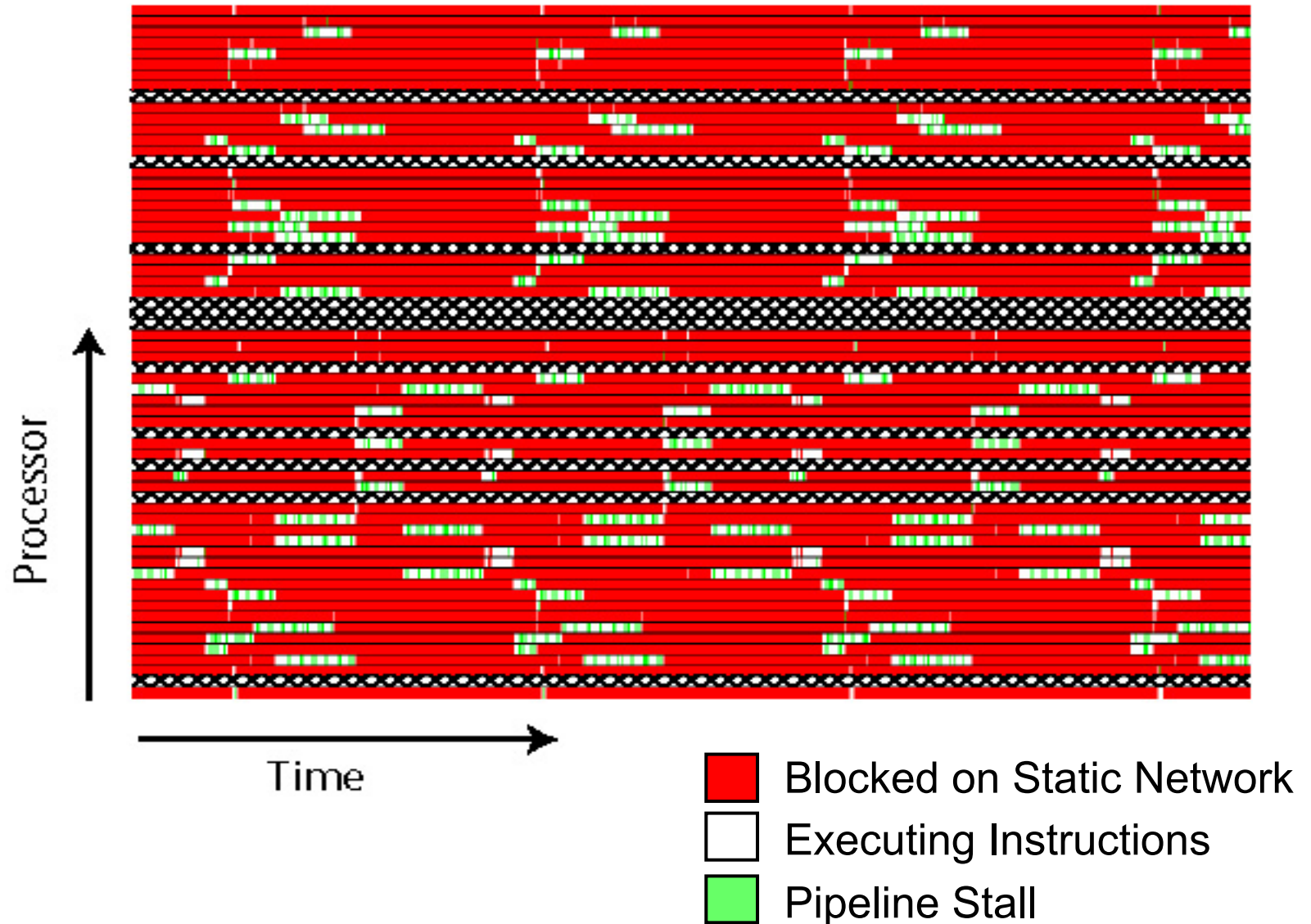


```
  add splitjoin {  
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    for (int i=0; i<numBeams; i++) {  
      add pipeline {  
        add VectorMult();  
        add FIR3(N3);  
        add Magnitude();  
        add Detect();  
      };  
    };  
    join roundrobin(0);  
  };  
};
```



```
}
```

Radar Array Front End on Raw

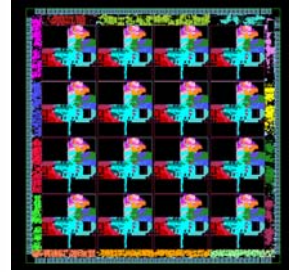
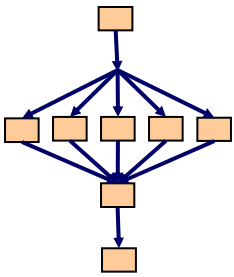


Bridging the Abstraction layers



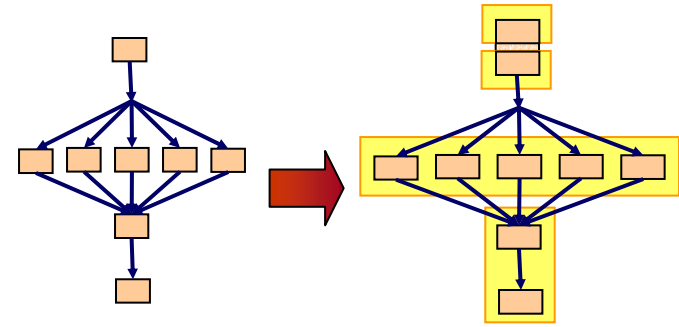
- StreamIt language exposes the data movement
 - Graph structure is architecture independent
- Each architecture is different in granularity and topology
 - Communication is exposed to the compiler
- The compiler needs to efficiently bridge the abstraction
 - Map the computation and communication pattern of the program to the PE's, memory and the communication substrate

Bridging the Abstraction layers



- StreamIt language exposes the data movement
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 - Map the computation and communication pattern of the program to the PE's, memory and the communication substrate
- The StreamIt Compiler
 - Partitioning
 - Placement
 - Scheduling
 - Code generation

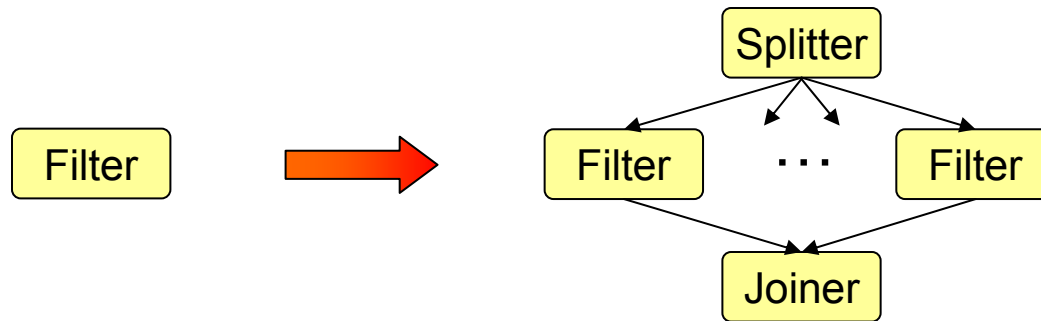
Partitioning: Choosing the Granularity



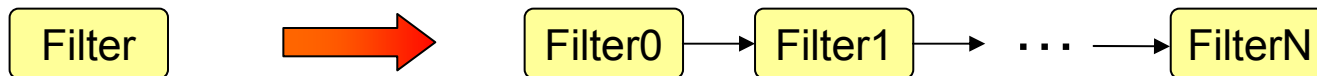
- Mapping filters to tiles
 - # filters should equal (or a few less than) # of tiles
 - Each filter should have similar amount of work
 - Throughput determined by the filter with most work
- Compiler Algorithm
 - Two primary transformations
 - Filter fission
 - Filter fusion
 - Uses a greedy heuristic

Partitioning - Fission

- Fission - splitting streams
 - Duplicate a filter, placing the duplicates in a SplitJoin to expose parallelism.

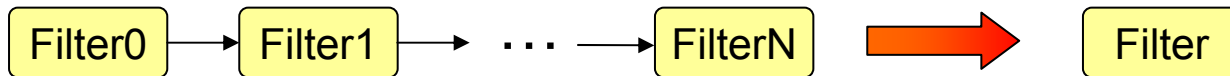
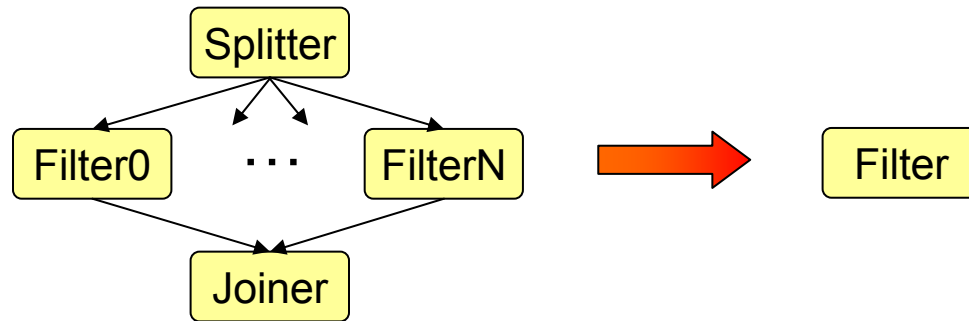


-Split a filter into a pipeline for load balancing

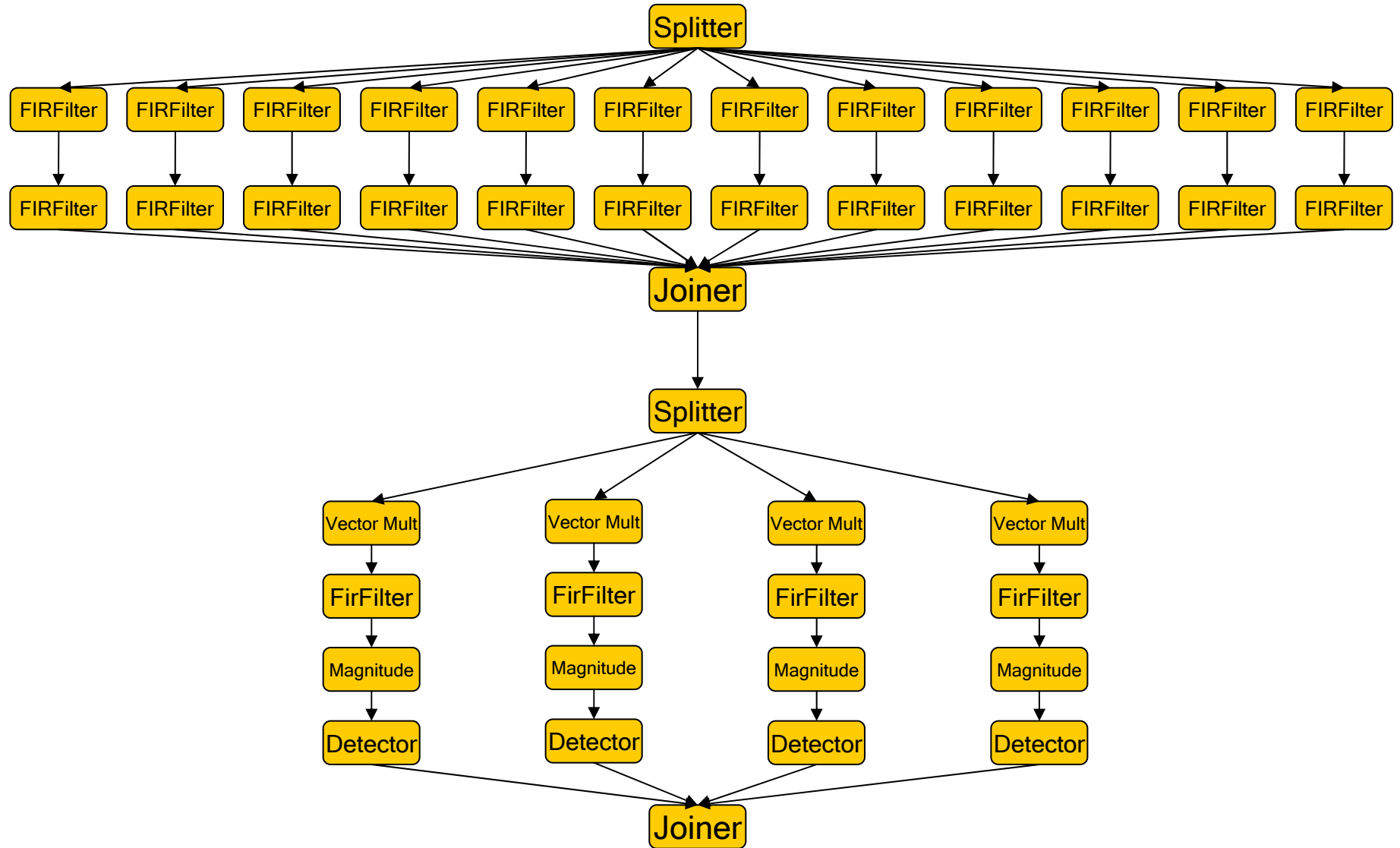


Partitioning - Fusion

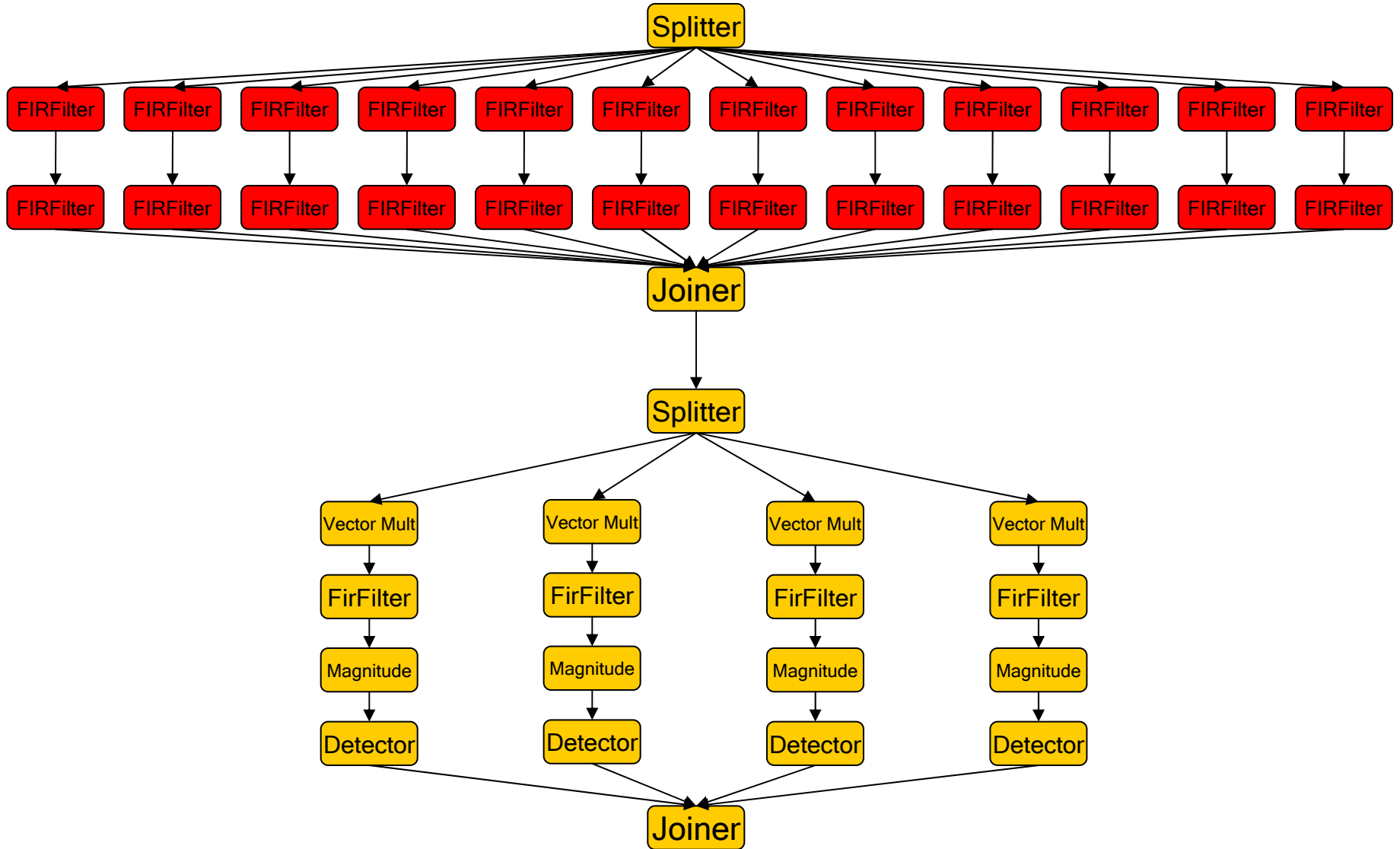
- Fusion - merging streams
 - Merge filters into one filter for load balancing and synchronization removal



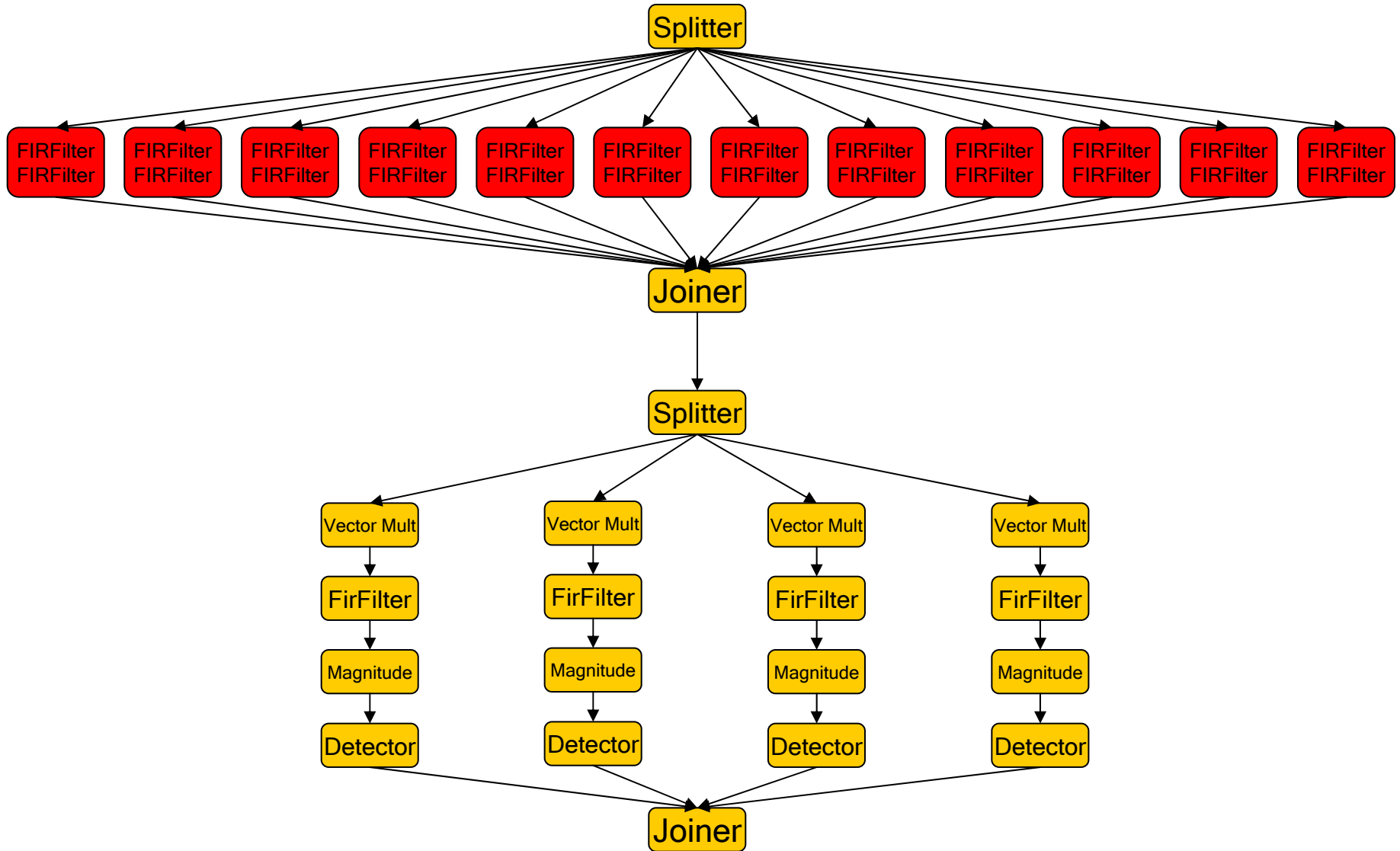
Example: Radar Array Front End (original)



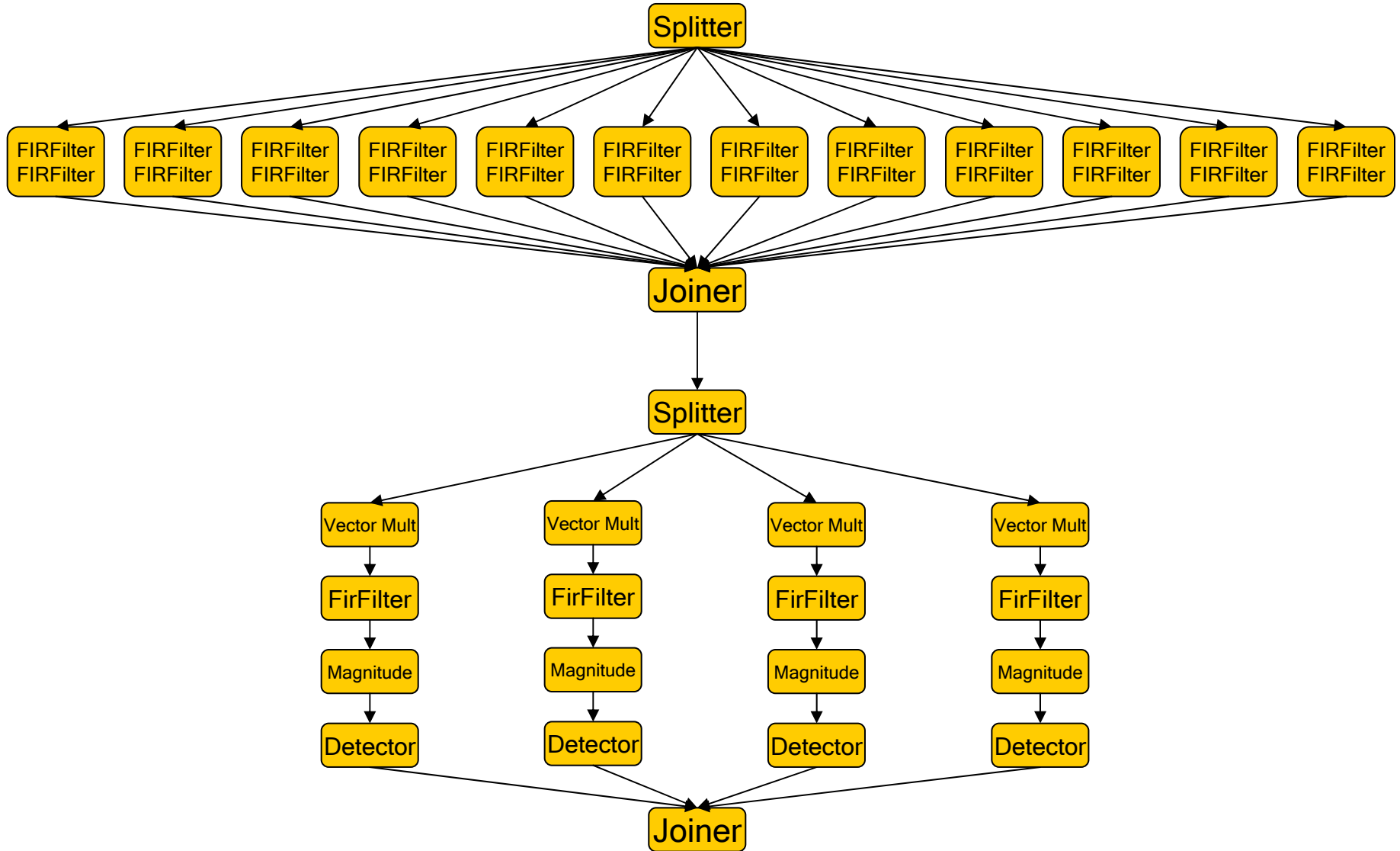
Example: Radar Array Front End



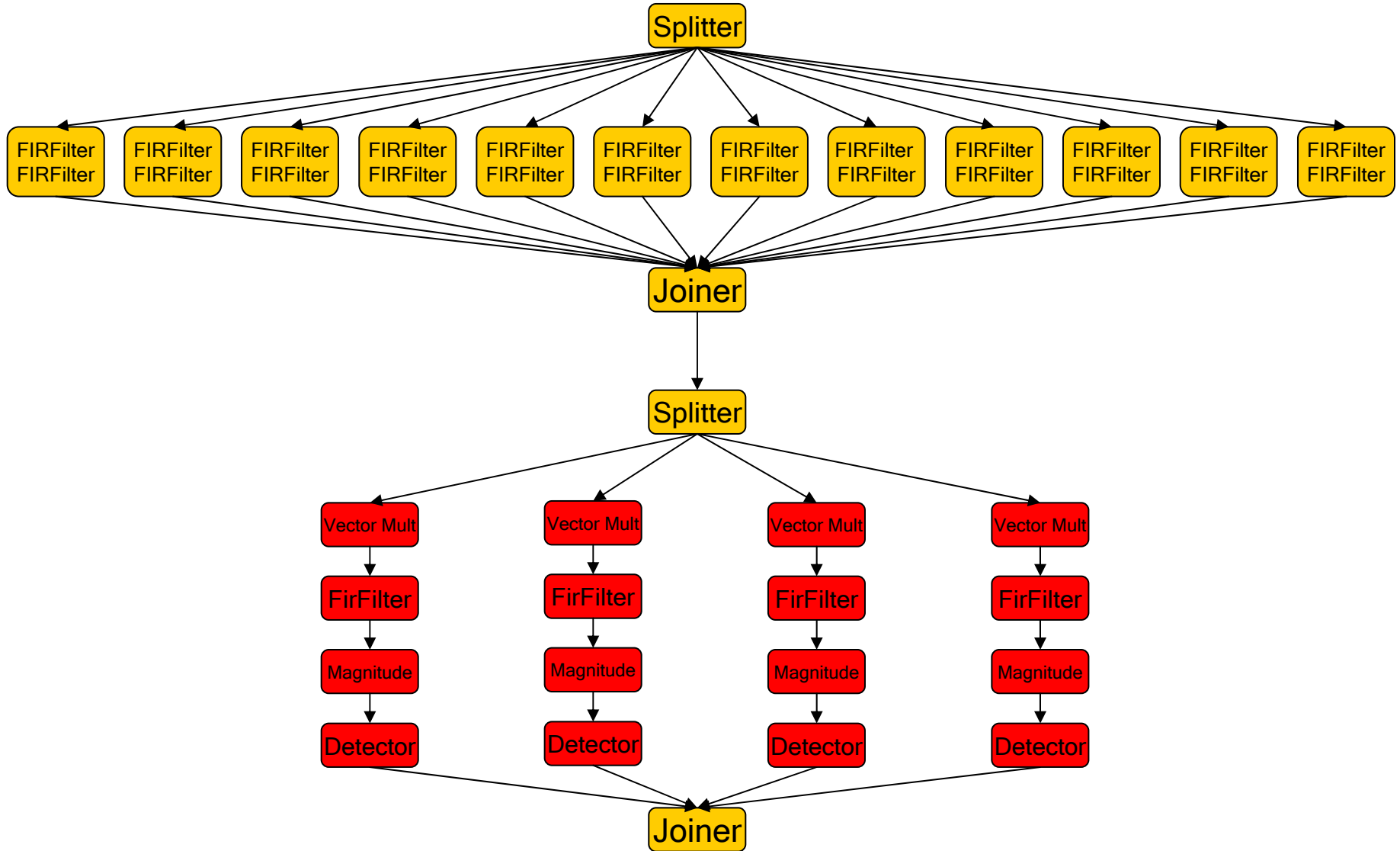
Example: Radar Array Front End



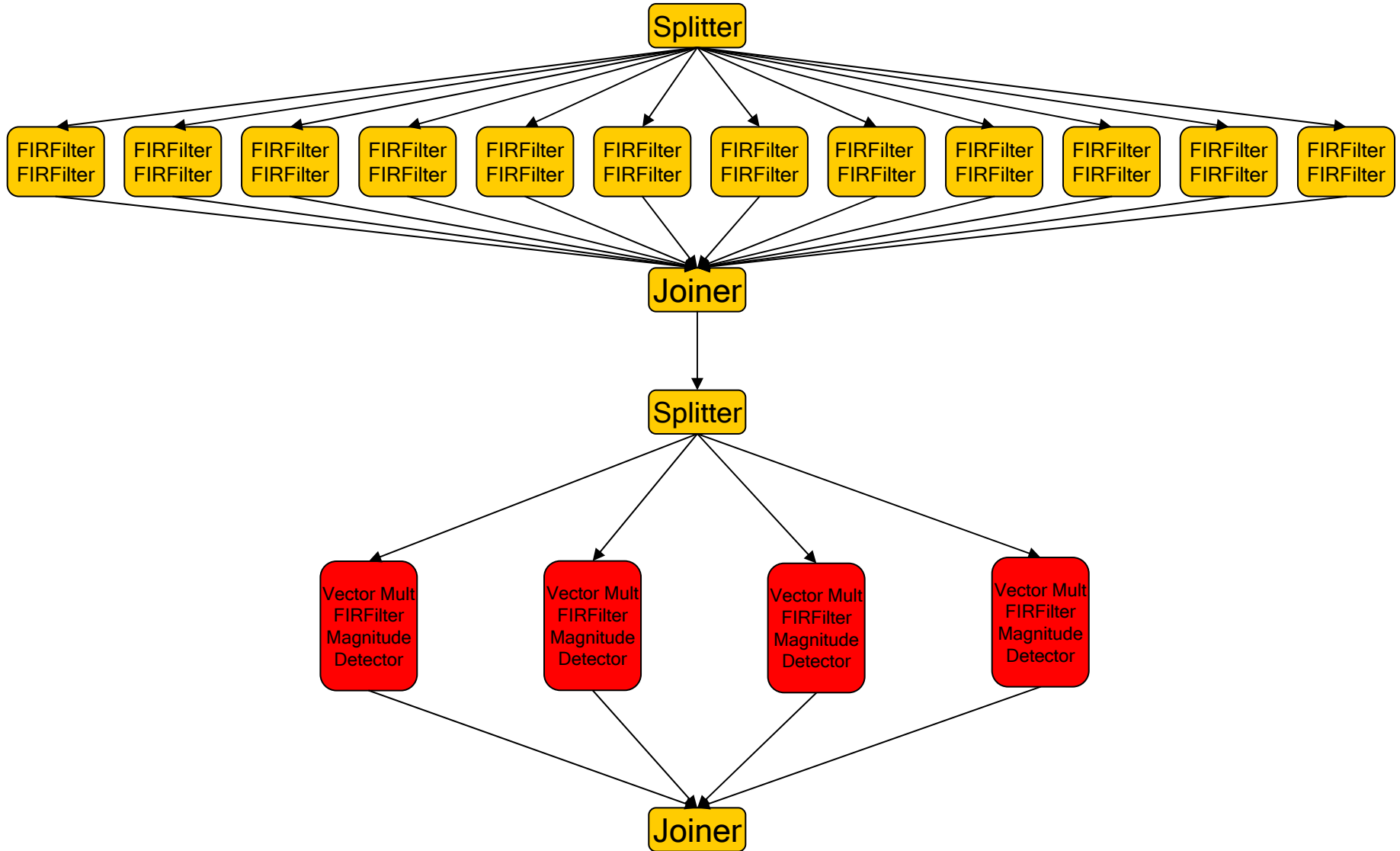
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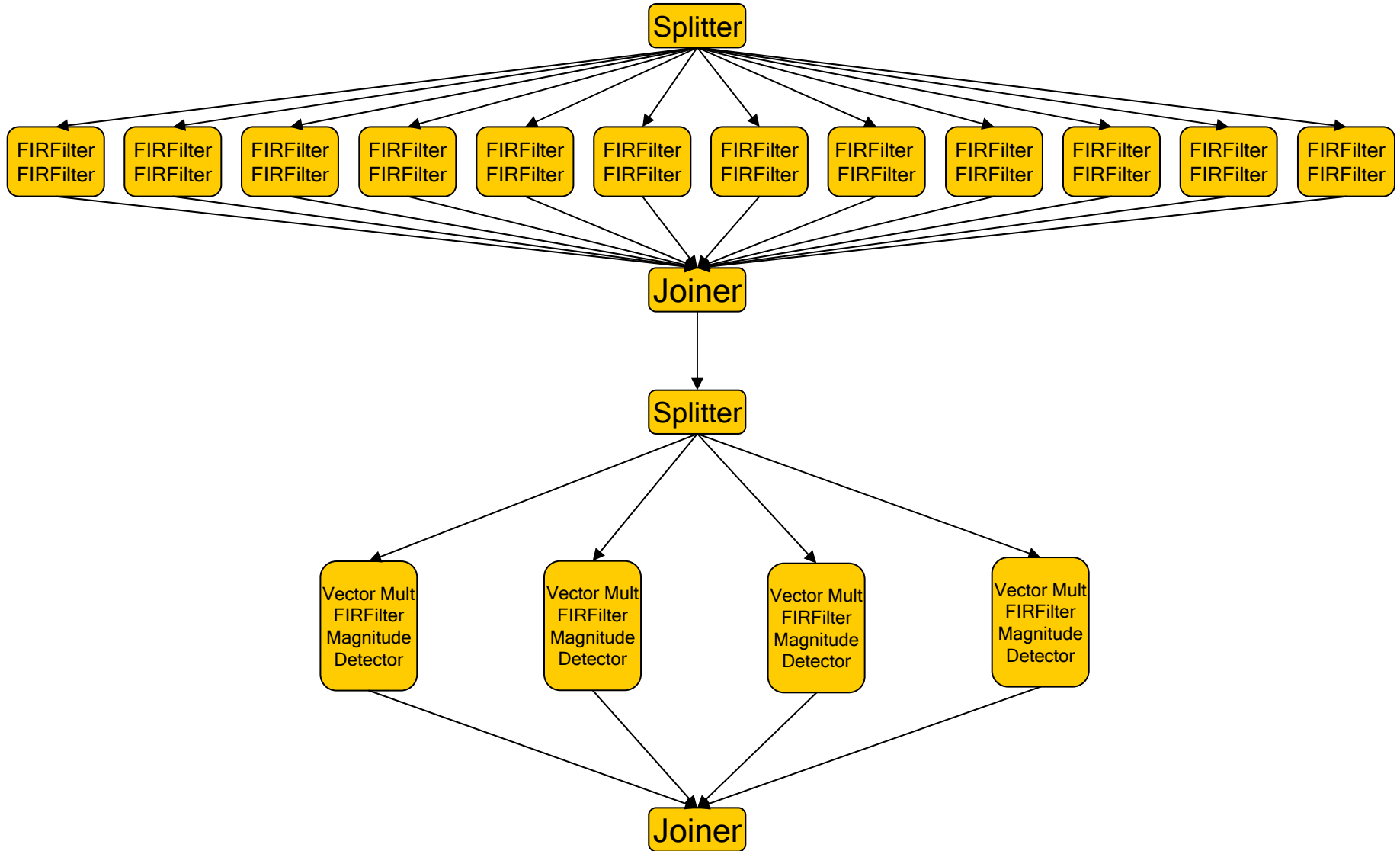
Example: Radar Array Front End



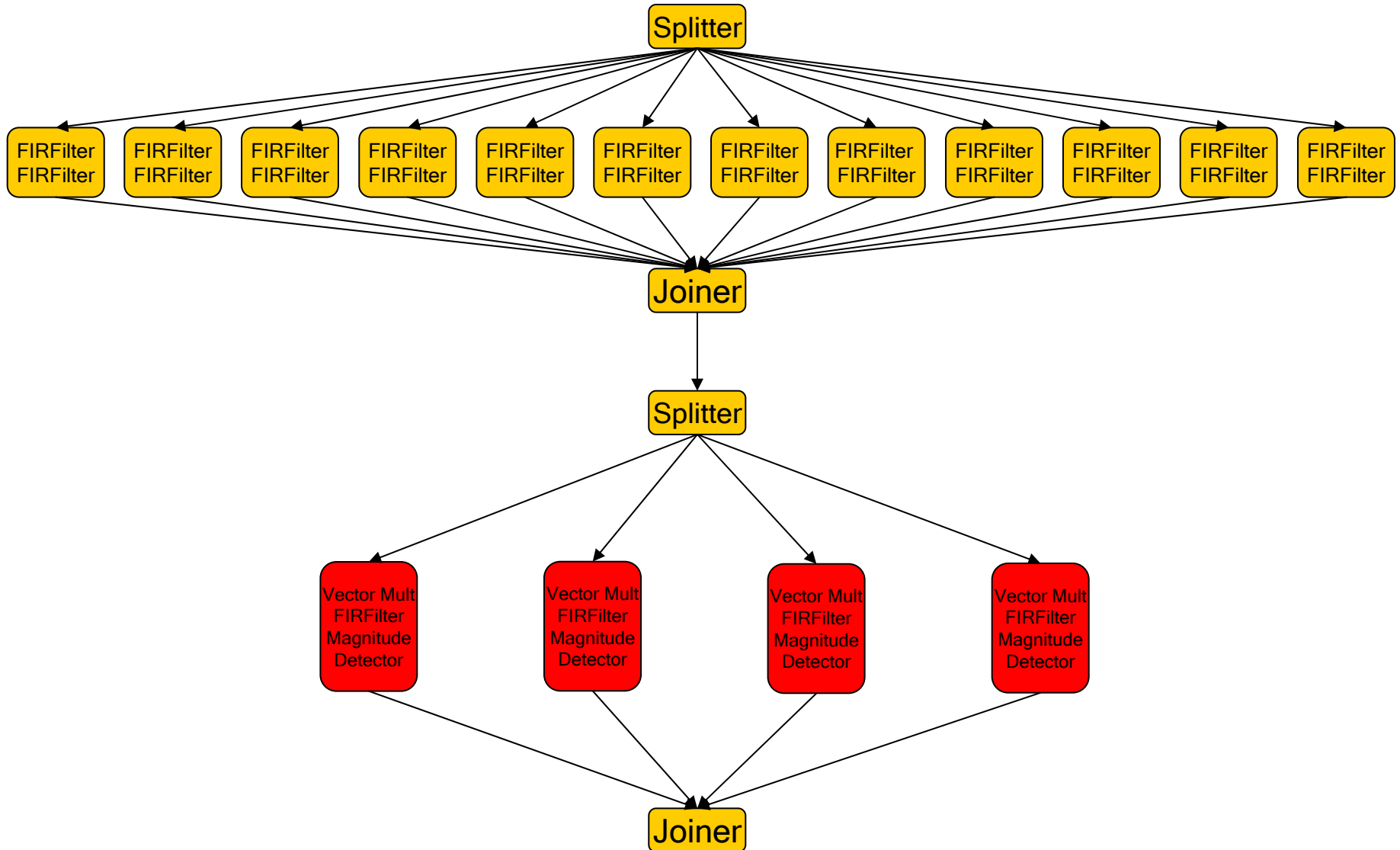
Example: Radar Array Front End



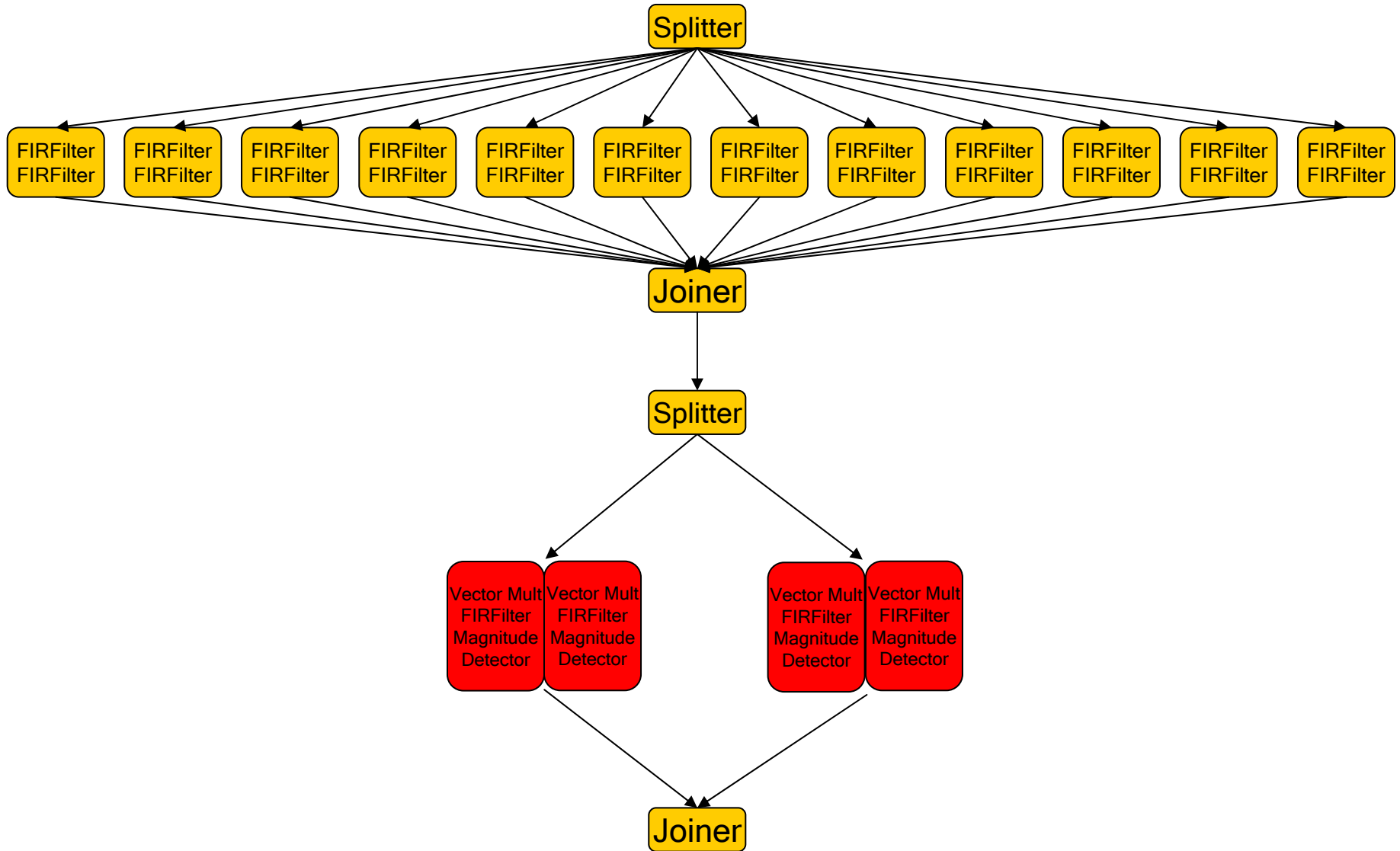
Example: Radar Array Front End



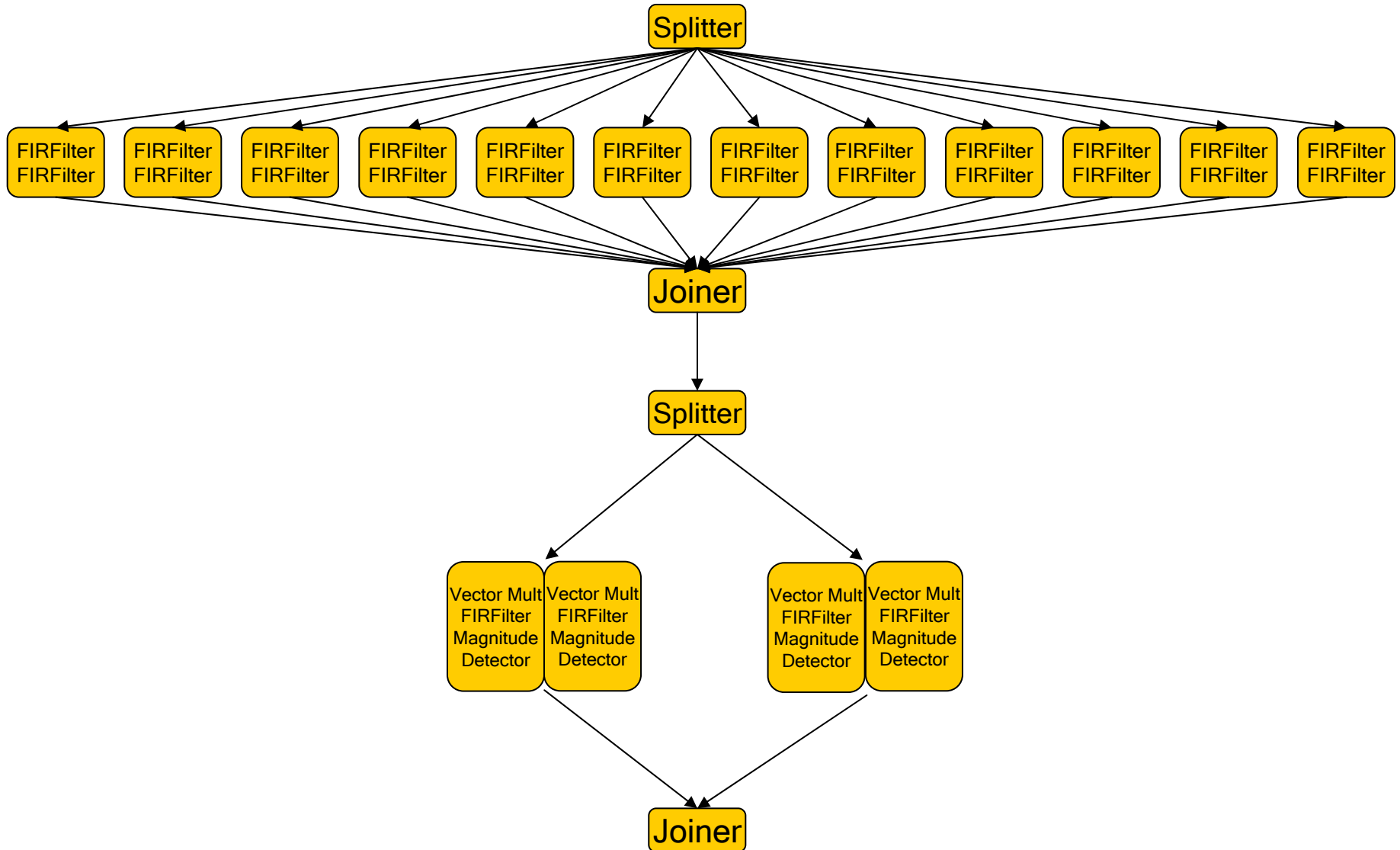
Example: Radar Array Front End



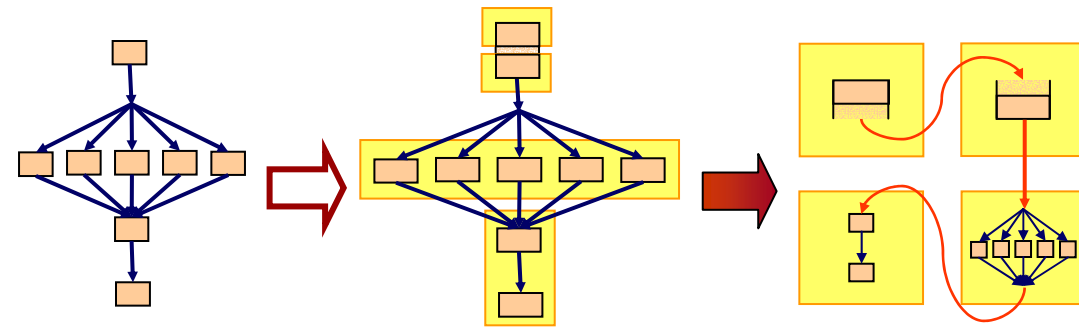
Example: Radar Array Front End



Example: Radar Array Front End (Balanced)

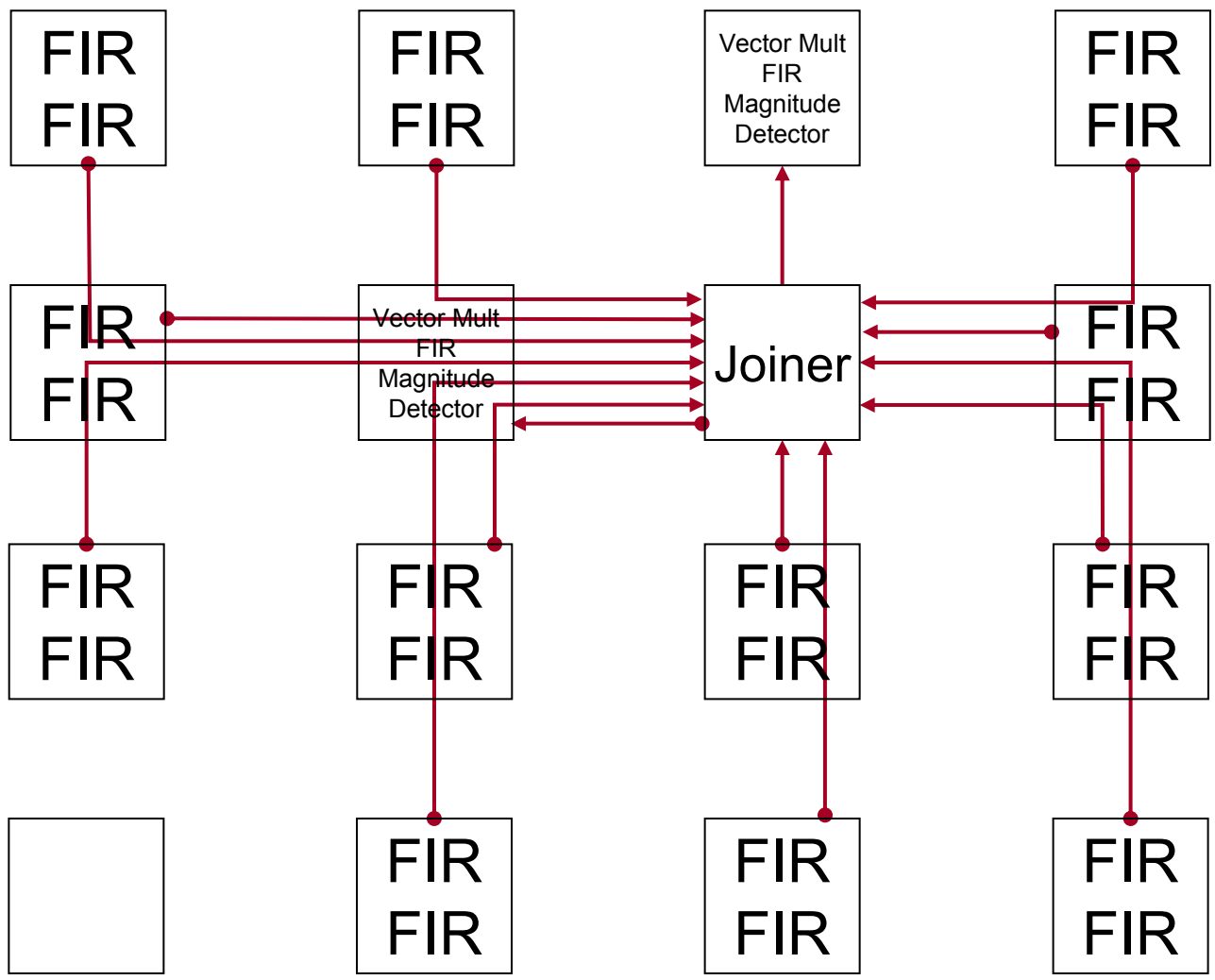


Placement: Minimizing Communication

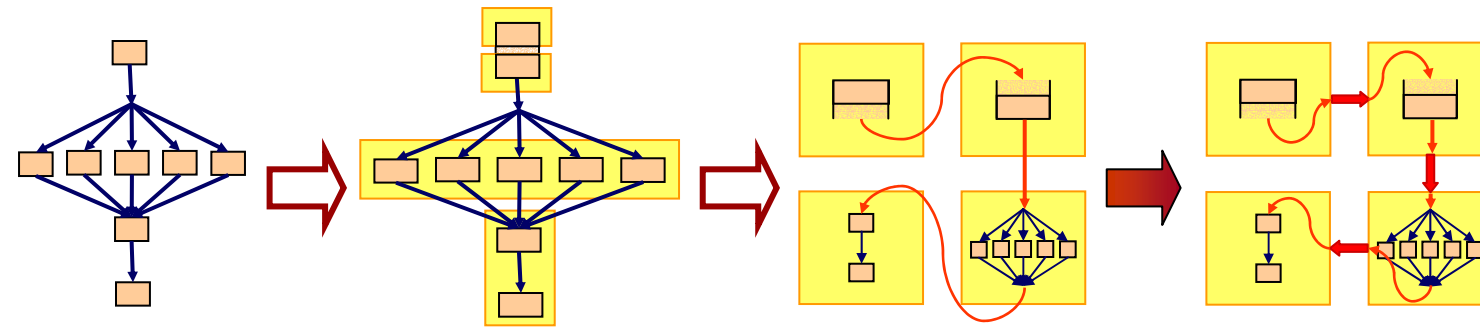


- Assign filters to tiles
 - Communicating filters → try to make them adjacent
 - Reduce overlapping communication paths
 - Reduce/eliminate cyclic communication if possible
- Compiler algorithm
 - Uses Simulated Annealing

Placement for Partitioned Radar Array Front End



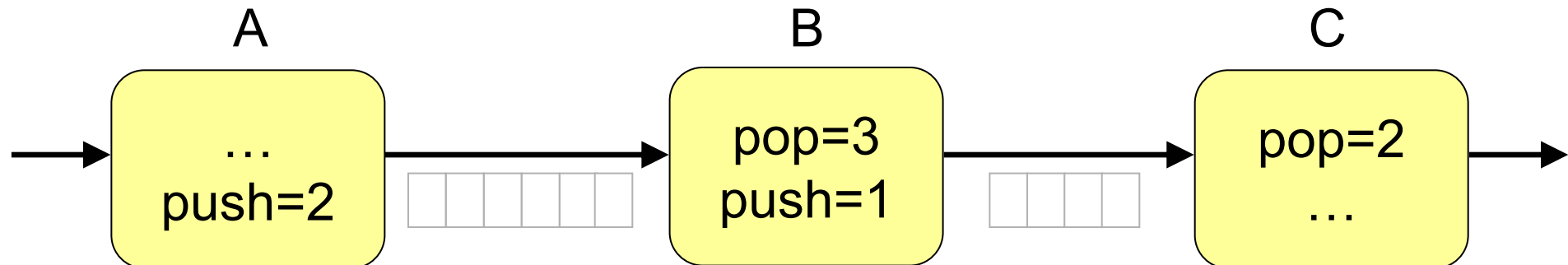
Scheduling: Communication Orchestration



- Create a communication schedule
- Compiler Algorithm
 - Calculate an initialization and steady-state schedule
 - Simulate the execution of an entire cyclic schedule
 - Place static route instructions at the appropriate time

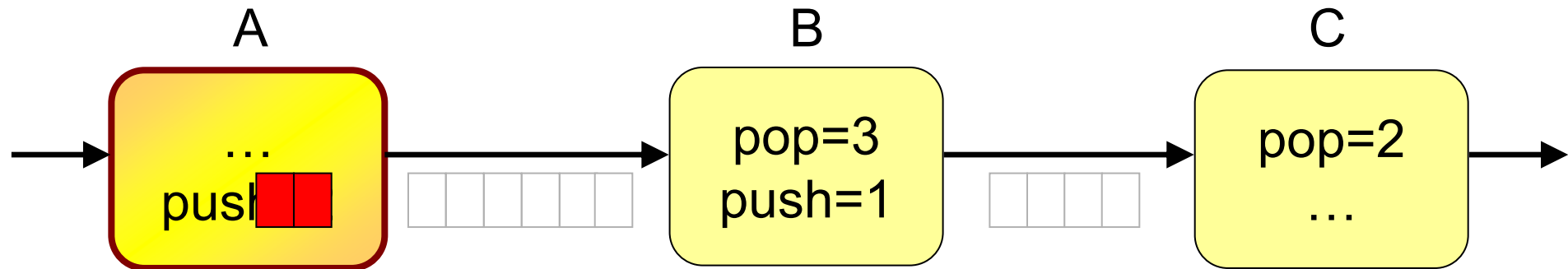
Steady-State Schedule

- All data pop/push rates are constant
- Can find a Steady-State Schedule
 - # of items in the buffers are the same before and the after executing the schedule
 - There exist a unique minimum steady state schedule
- Schedule = { }



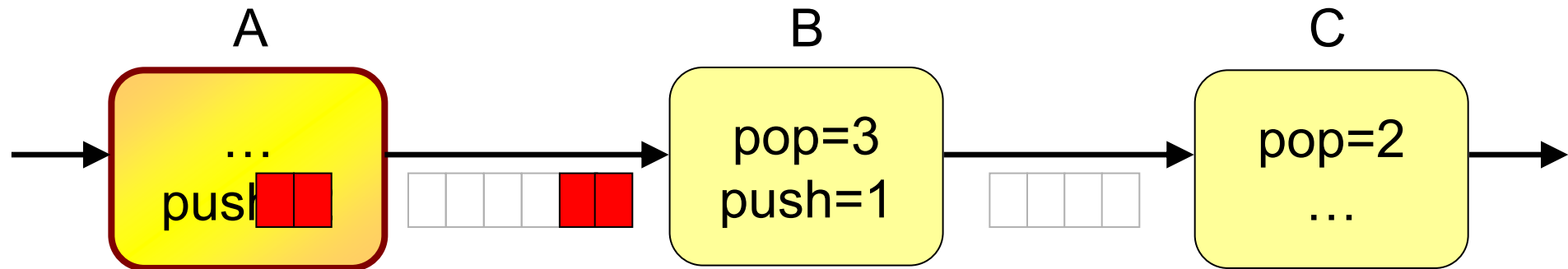
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- Schedule = { A }



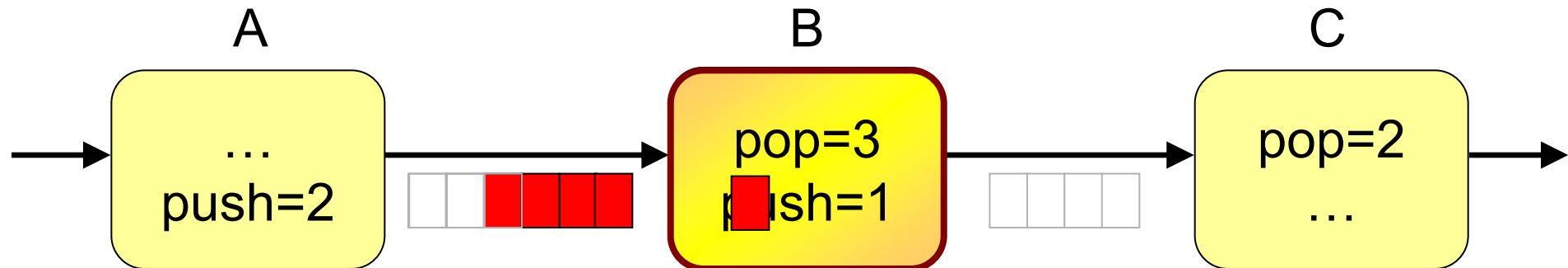
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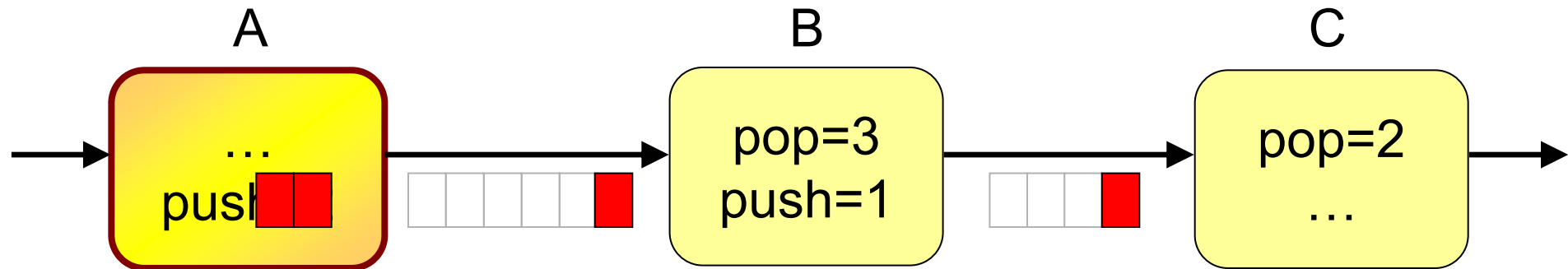
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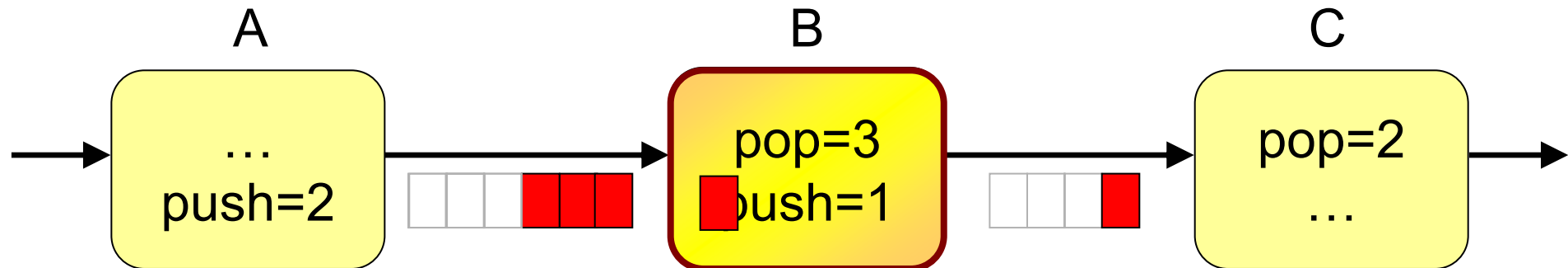
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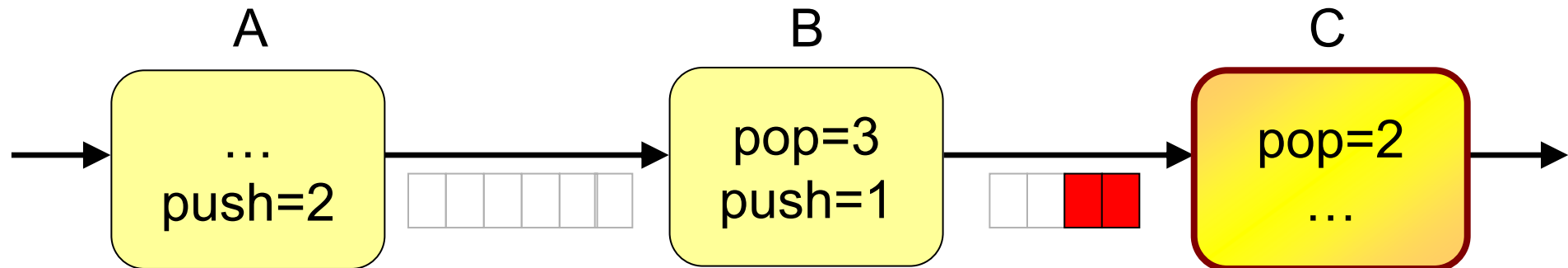
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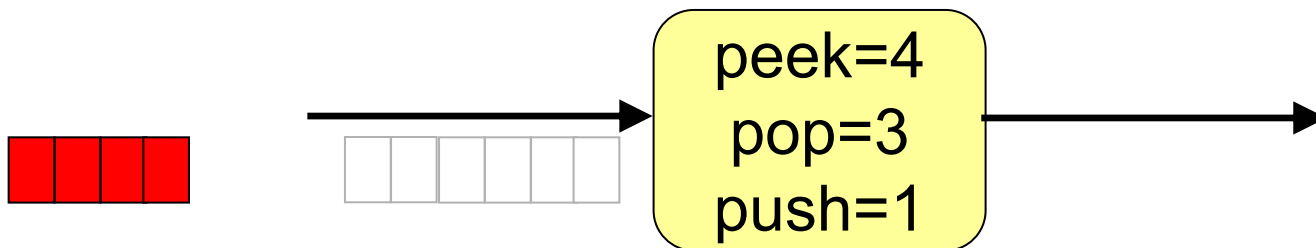
Steady-State Schedule

- All data pop/push rates are constant
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- Schedule = { A, A, B, A, B, C }



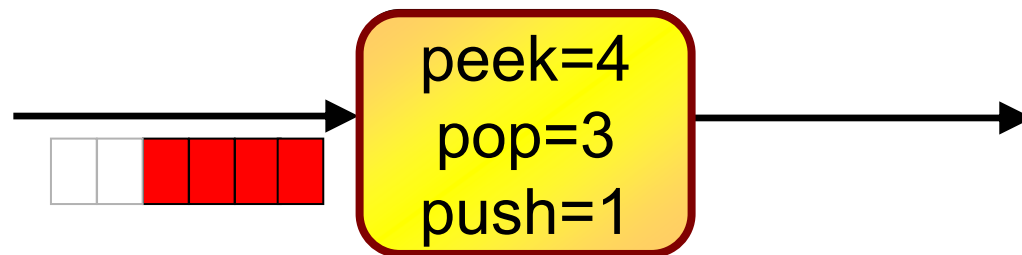
Initialization Schedule

- When $\text{peek} > \text{pop}$, buffer cannot be empty after firing a filter
- Buffers are not empty at the beginning/end of the steady state schedule
- Need to fill the buffers before starting the steady state execution

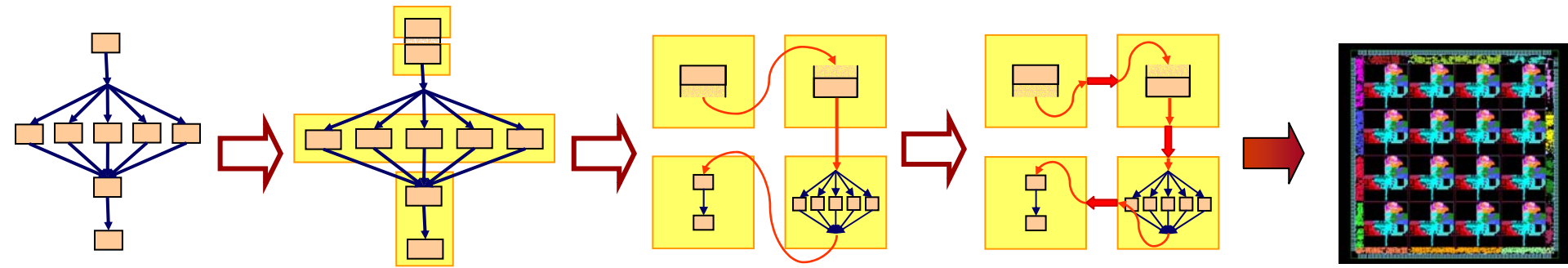


Initialization Schedule

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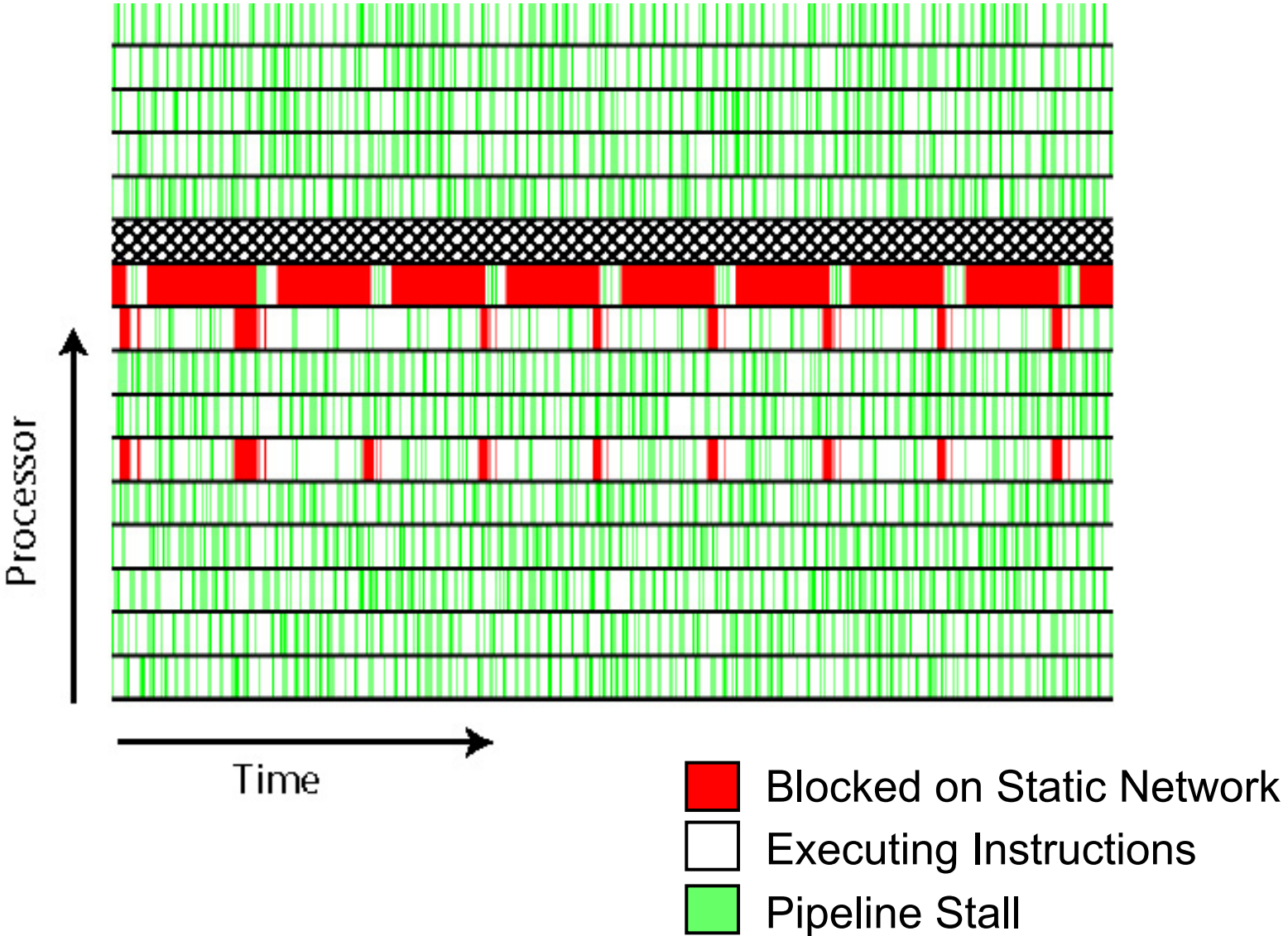


Code Generation: Optimizing tile performance

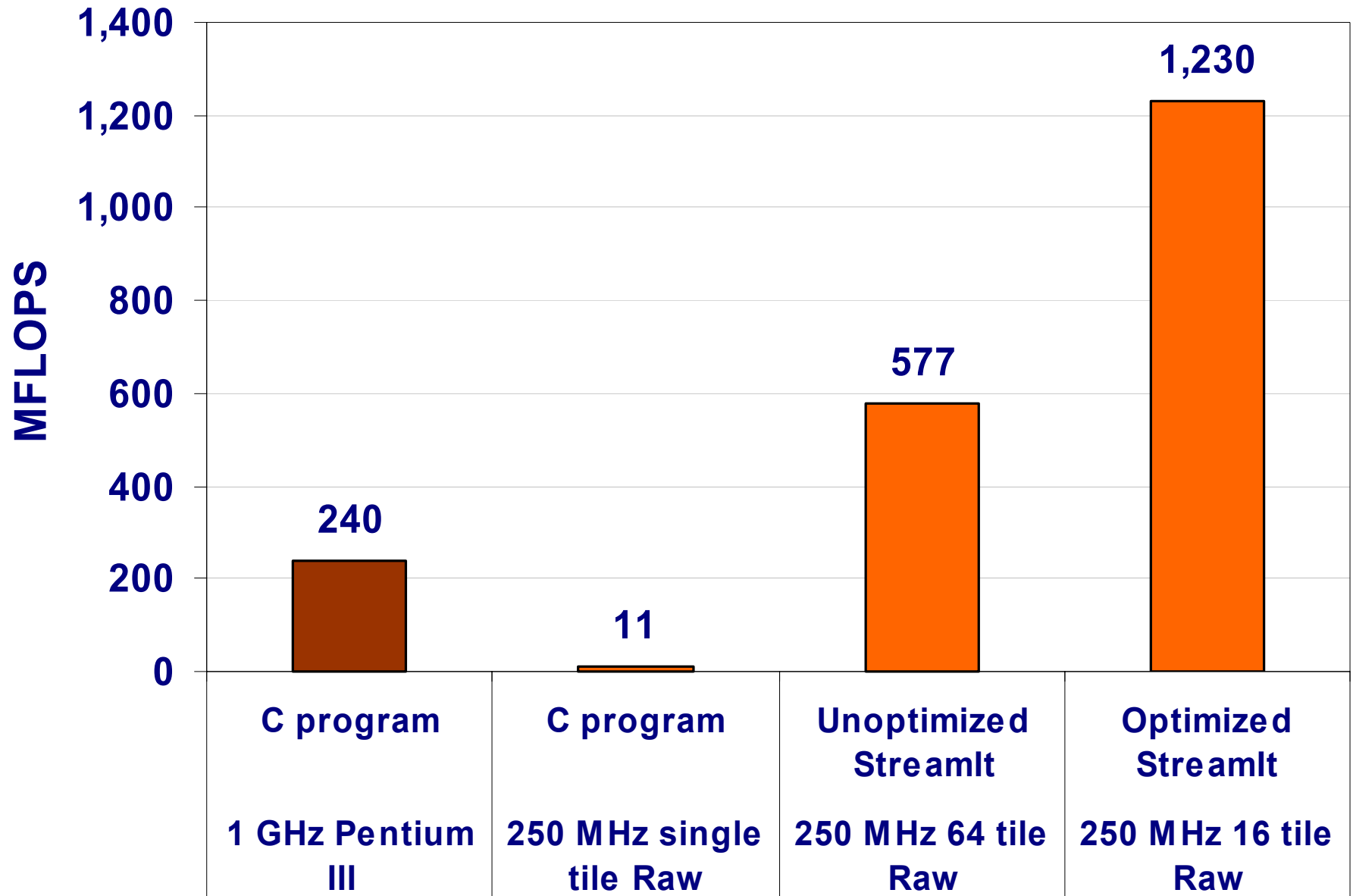


- Creates code to run on each tile
 - Optimized by the existing node compiler
- Generates the switch code for the communication

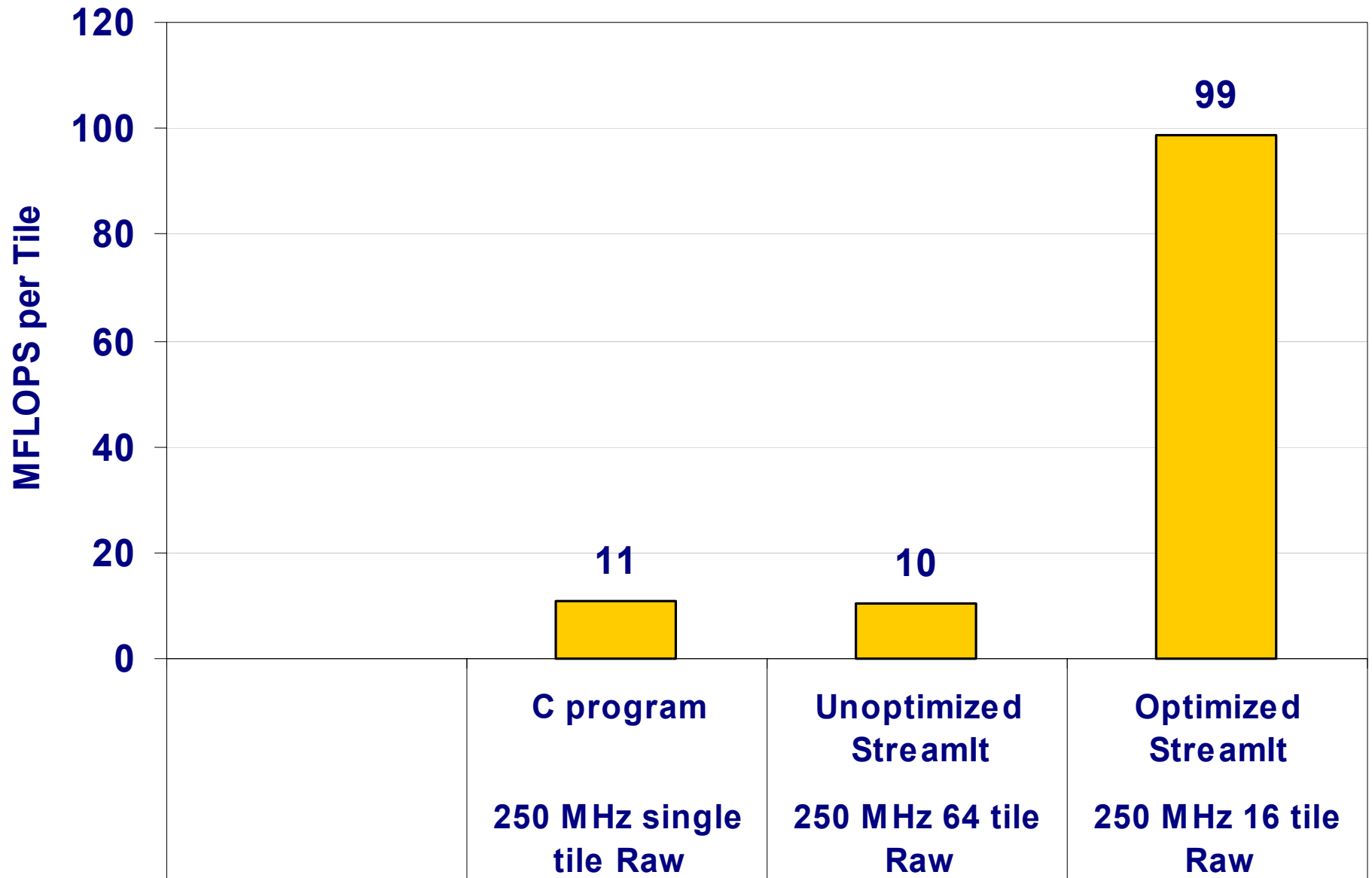
Performance Results for Radar Array Front End



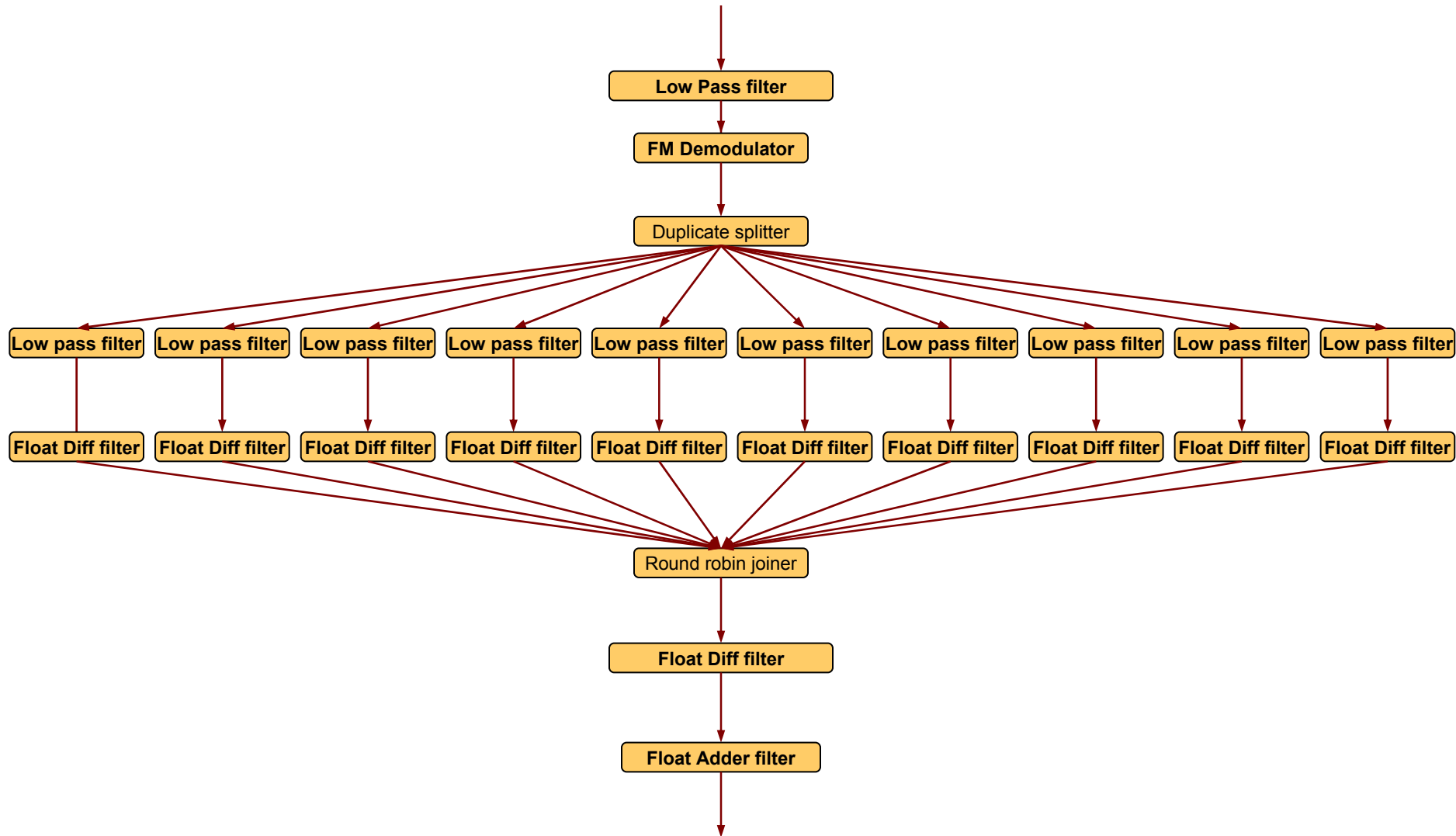
Performance of Radar Array Front End



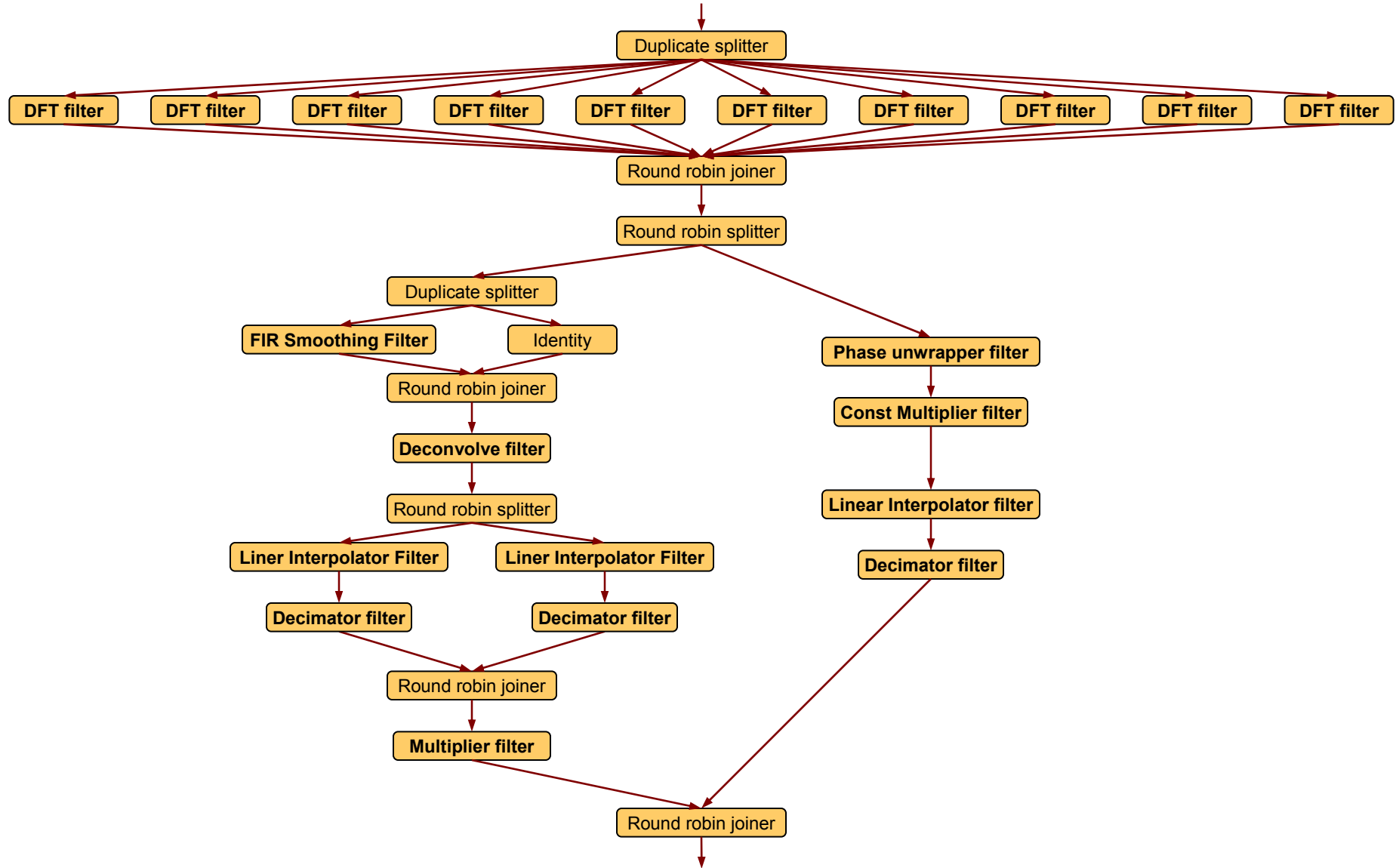
Utilization of Radar Array Front End



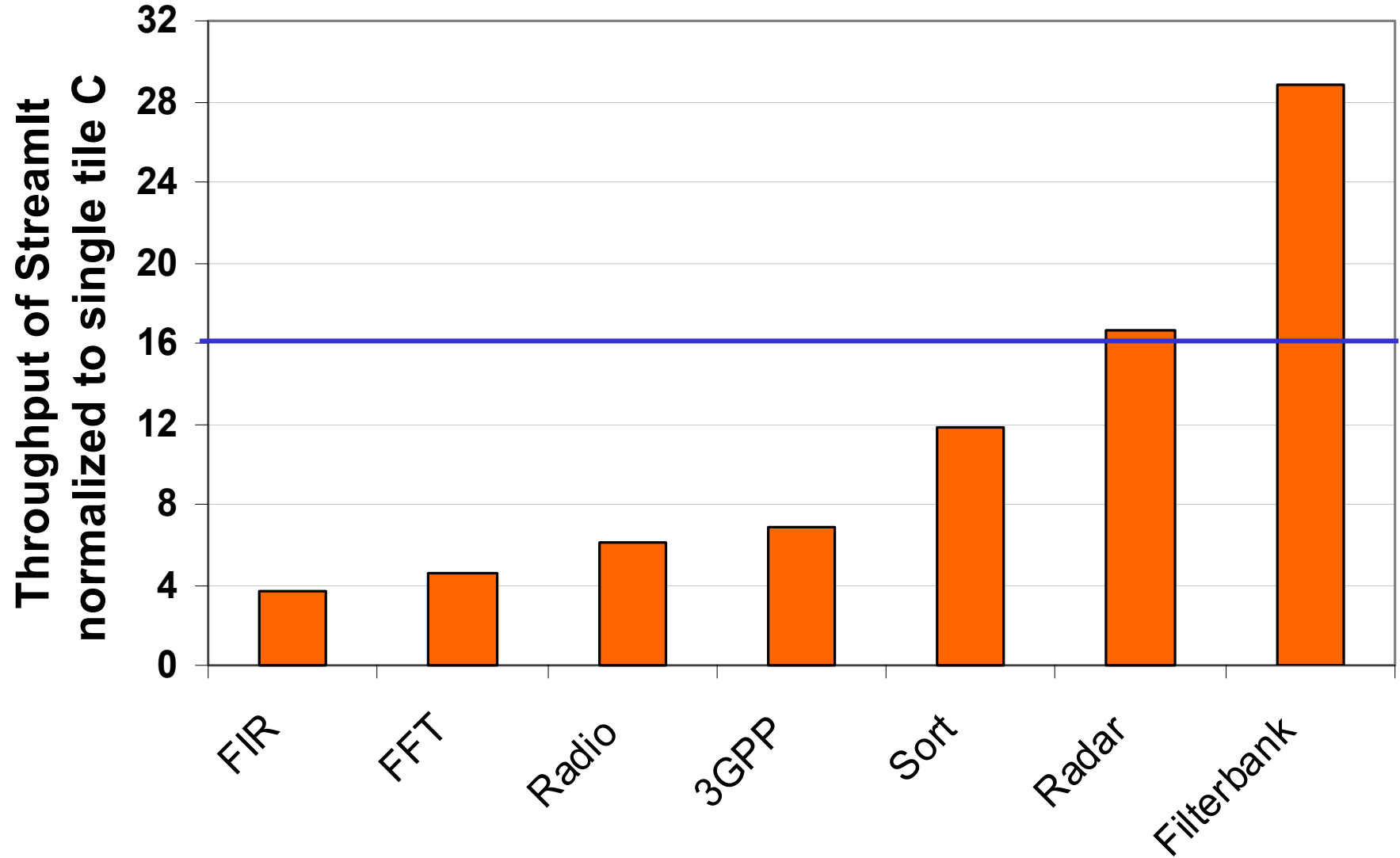
StreamIt Applications: FM Radio with an Equalizer



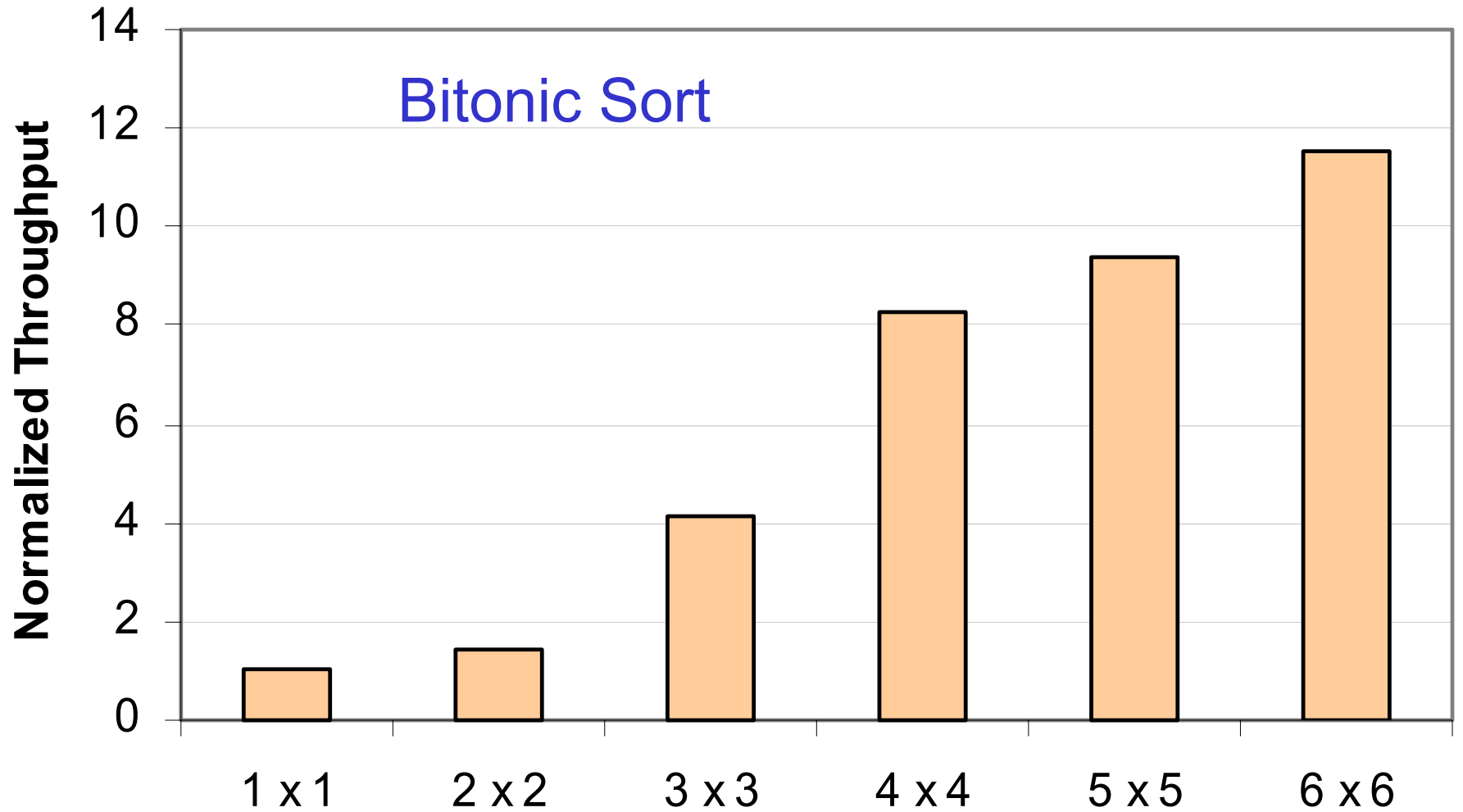
StreamIt Applications: Vocoder



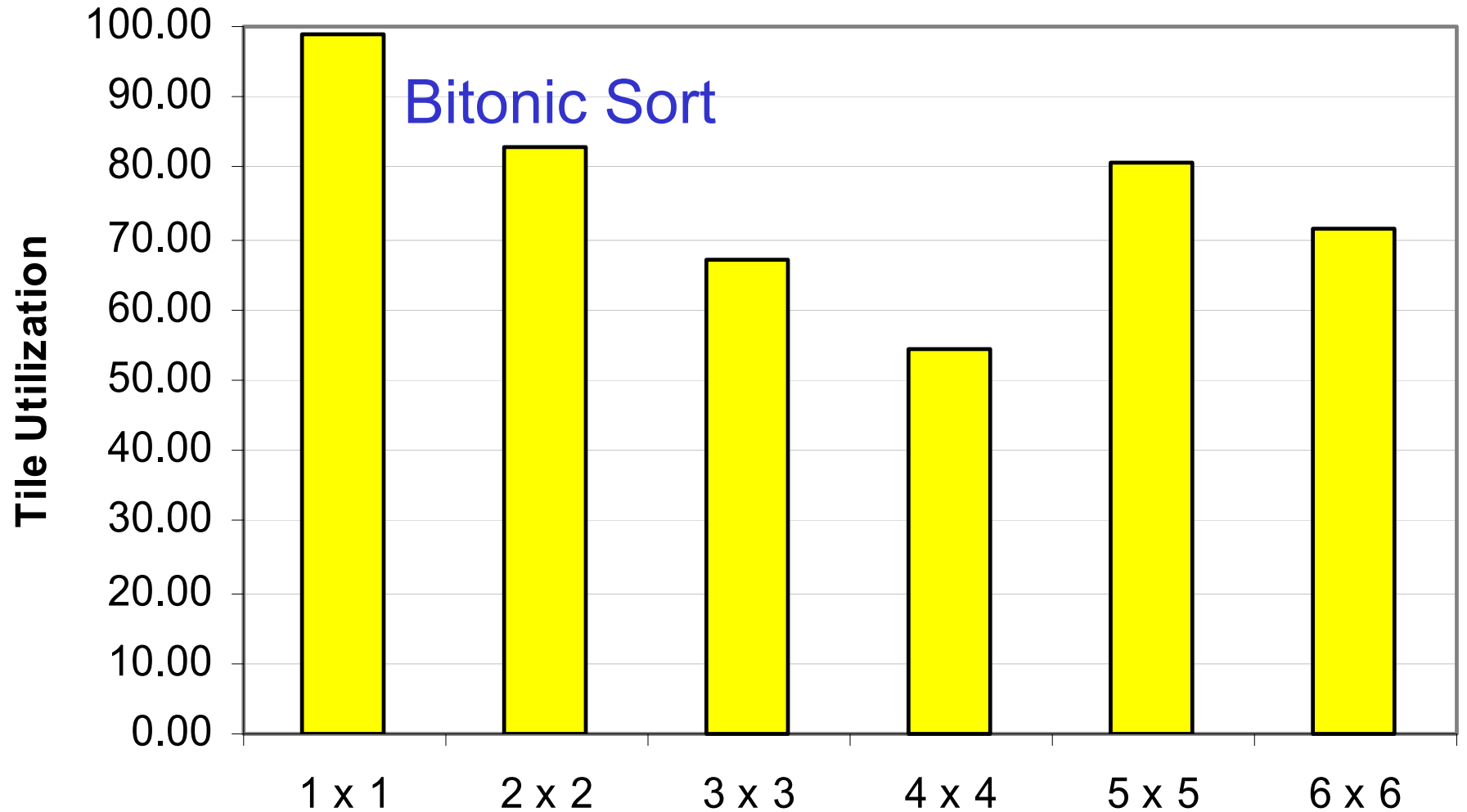
Application Performance



Scalability of StreamIt



Scalability of StreamIt



Related Work

- Stream-C / Kernel-C (Dally et. al)
 - Compiled to Imagine with time multiplexing
 - Extensions to C to deal with finite streams
 - Programmer explicitly calls stream “kernels”
 - Need program analysis to overlap streams / vary target granularity
- Brook (Buck et. al)
 - Architecture-independent counterpart of Stream-C / Kernel-C
 - Designed to be more parallelizable
- Ptolemy (Lee et. al)
 - Heterogeneous modeling environment for DSP
 - Many scheduling results shared with StreamIt
 - Don't focus on language development / optimized code generation
- Other languages
 - Occam, SISAL – not statically schedulable
 - LUSTRE, Lucid, Signal, Esterel – don't focus on parallel performance

Conclusion

- Streaming Programming Model
 - An important class of applications
 - Can break the von Neumann bottleneck
 - A natural fit for a large class of applications
 - Straightforward mapping to the architectural model
- StreamIt: A Machine Language for Communication Exposed Architectures
 - Expose the common properties
 - Multiple instruction streams
 - Software exposed communication
 - Fast local memory co-located with execution units
 - Hide the differences
 - Granularity of execution units
 - Type and topology of the communication network
 - Memory hierarchy
- A good compiler can eliminate the overhead of abstraction