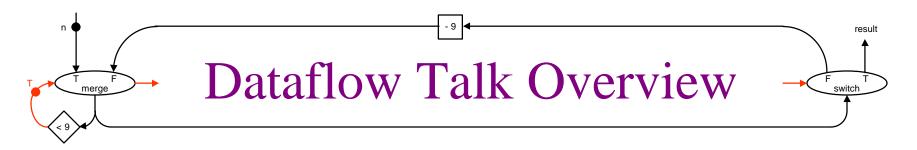
Datawhat?

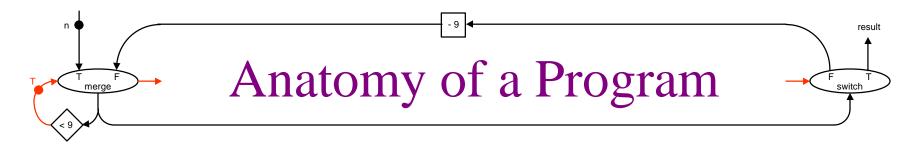
J.P. Grossman March 14, 19100



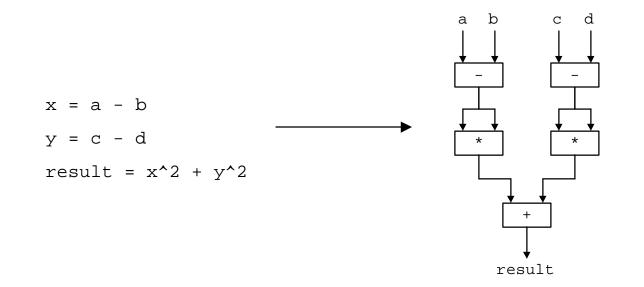
- Major focus: How, what and why
 - What is dataflow?
 - Why is it a good idea?
 - How do I build a dataflow computer?
- Briefly:
 - Existing machines
 - Programming models
 - Problems with dataflow
 - Hybrid approaches



- Given a sequential program, how do we:
 - Find and exploit parallelism?
 - Identify independent operations
 - Execute them in parallel
 - Keep the processor busy with useful work?
 - Tolerate latency
 - Avoid pipeline stalls
 - Cheap synchronization

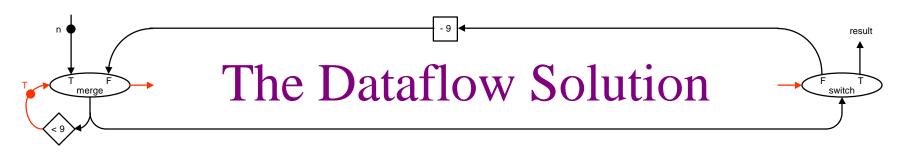


- Computation can be described by a dataflow graph
- All computers evaluate the dataflow graph

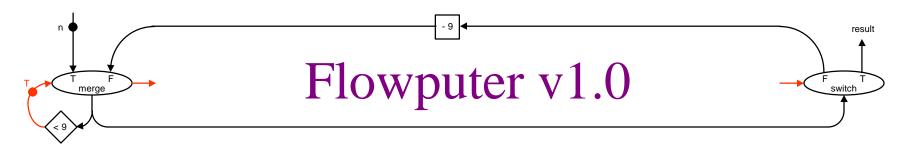




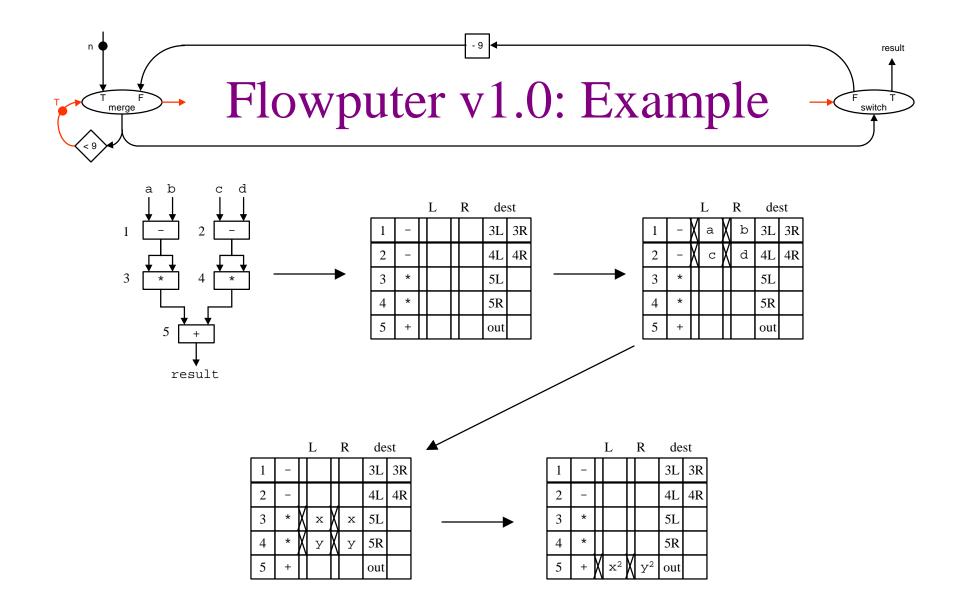
- Operations in dataflow graph are given a sequential order at compile time
- Problems:
 - Stalls can occur due to dynamic data dependencies
 - Hides parallelism rather than uses it
 - Compiling code with a finite number of registers creates artificial dependencies
 - Makes life difficult for architectures that try to be clever
 - Synchronization is expensive in parallel machines

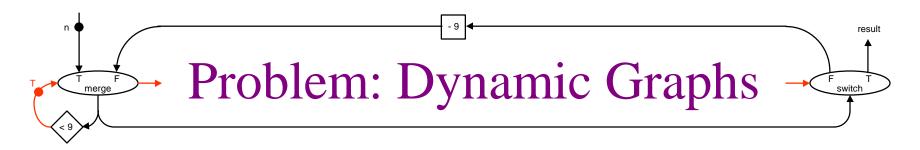


- Program Counters are evil get rid of them
 - instructions should execute when their operands are available, not when an arbitrary PC says it's time
- Compile and run the dataflow graph directly
 - compiler produces a set of operations
 - each op specifies one or more result destinations
 - an op can execute as soon its operands are available
- An "ideal" dataflow machine represents the *fastest possible* execution of a program

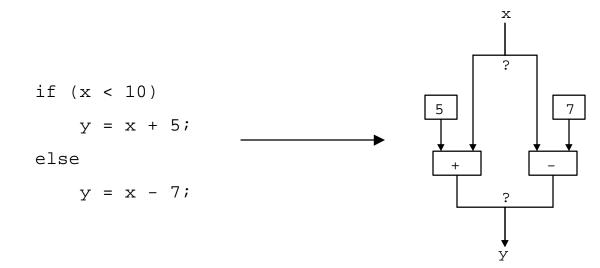


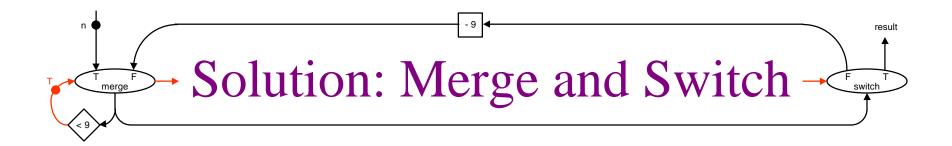
- Basic data structure is an *activity template*
 - opcode, storage for operands (with present bits), destination specifier(s)
- Flowputer maintains a set of *ready* operations
- On each cycle, grab *n* ready operations:
 - execute the operation
 - store the result in the destination(s)
 - check to see if any new operations become ready and, if so, add them to the ready set

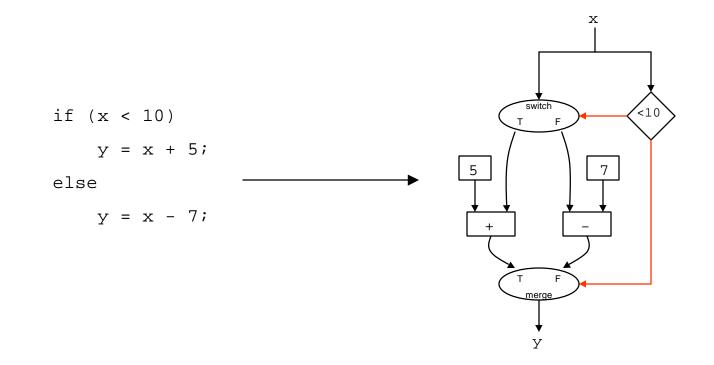


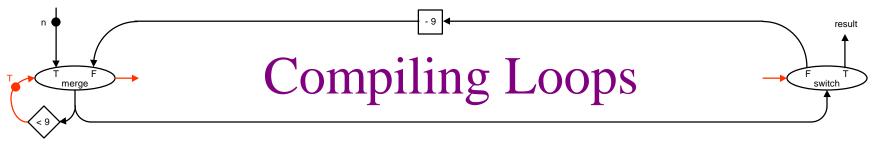


- Most program dataflow graphs are dynamically determined at run time
- Example:

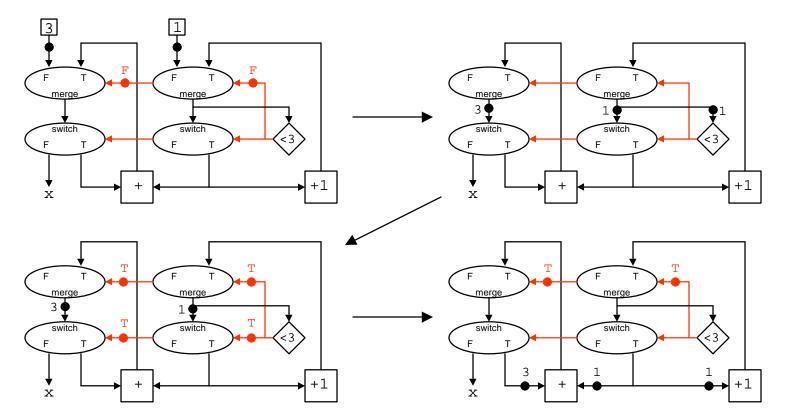


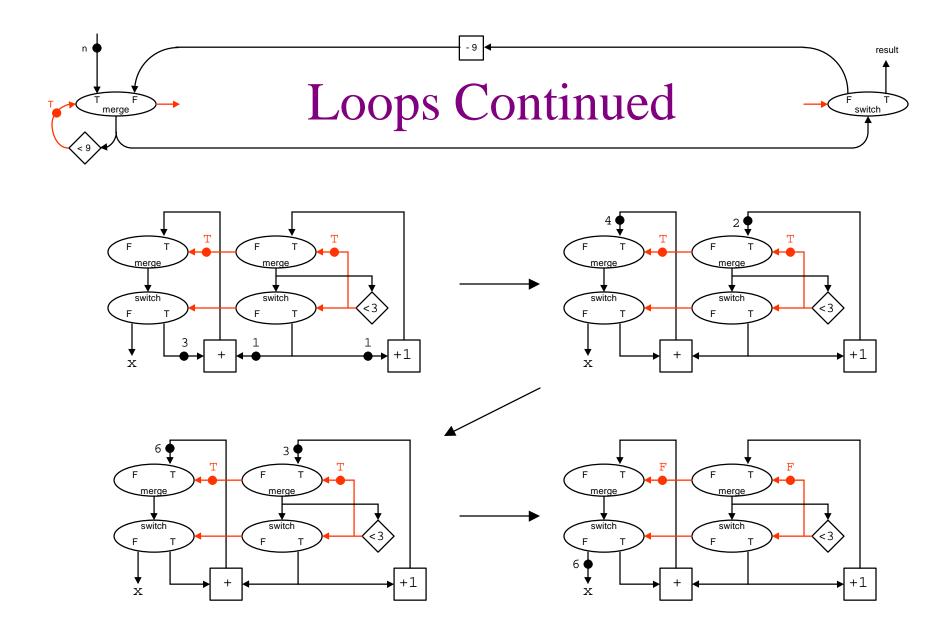


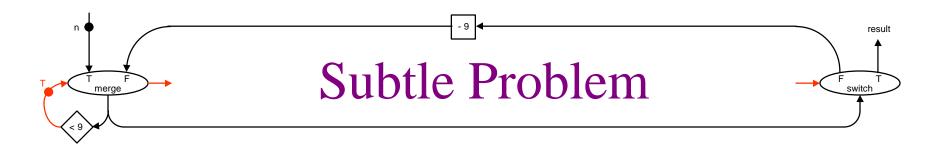


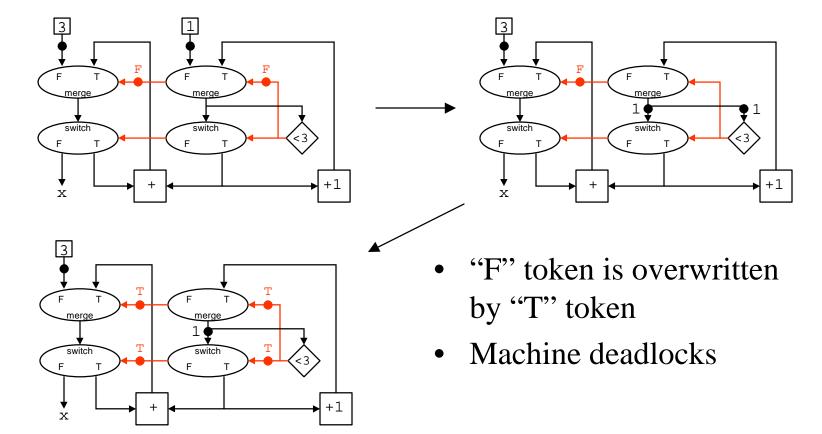


X = 3; for (i = 1 ; i < 3 ; i++) x = x + i;



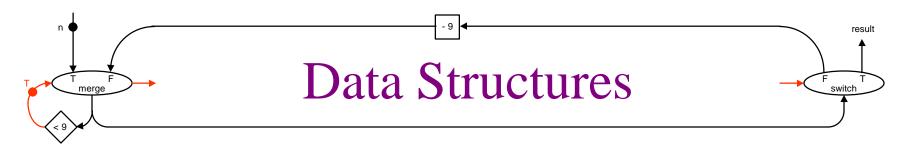




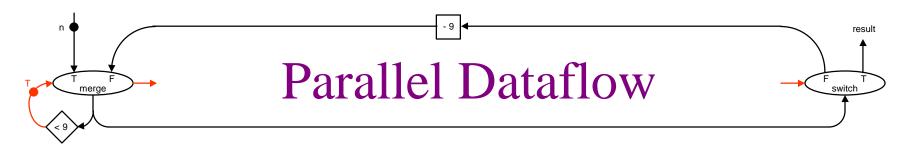




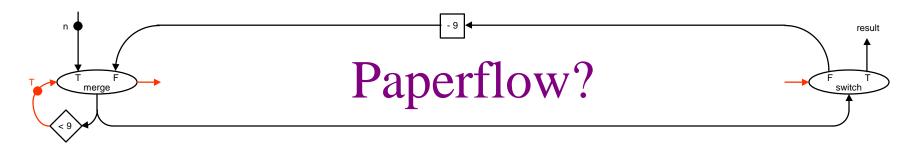
- Only allow one token per arc (Dennis)
 - Complicates hardware
 - Reduces performance and flexibility.. no recursion!
- Tagged Token Dataflow (Manchester, SIGMA-1)
 - Give each function call/loop iteration a unique tag
 - Also complicates hardware (tag matching)
- Explicit Token Store (Monsoon, EM-4)
 - Dynamically allocate frame memory for operands
 - Compiler ensures one token per arc



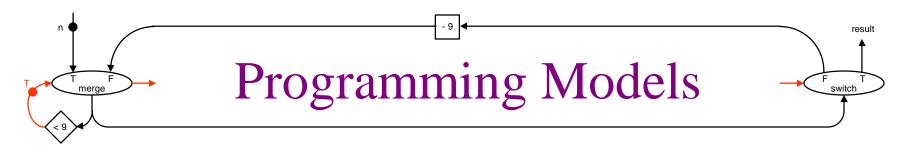
- Can't pass entire data structures in tokens
- Pass pointers instead.. but pointers to what?
 - Need to emulate "availability of operands"
- Solution: I-structure memory
 - Single write, multiple read, split-phase memory
 - Read to a full slot causes the value to be sent back
 - Read to an empty slot blocks until slot is filled
 - Write to an empty slot causes all waiting reads to be satisfied



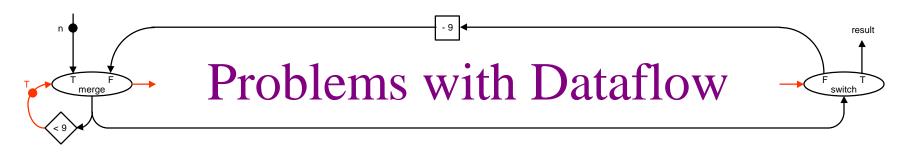
- At any point in time, a dataflow program generally has many ready operations
- Pick your favorite way to execute them in parallel
- Big question: do we get good speedup?
- Experiments with Monsoon indicate that the answer is "yes"
 - 8 processors, interconnection network for routing tokens to correct processor
 - Average speedup of 7.15x



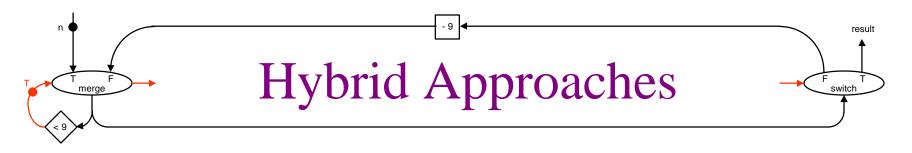
- Lots of paper, not too many actual machines
- Manchester (1981) Tagged Token
 - 12 functional units, 1 MIPS, 4-8x slower than VAX
- SIGMA-1 (1987) Tagged Token
 - 128 processors, 170 MFlops
- Monsoon (1988) Explicit Token Store
 - 8 processor configuration out-performs MIPS R3000
- EM-4 (1990) Hybrid Explicit Token Store
 - 80 processors, 1 GIPS, 14.63 GB/s peaks



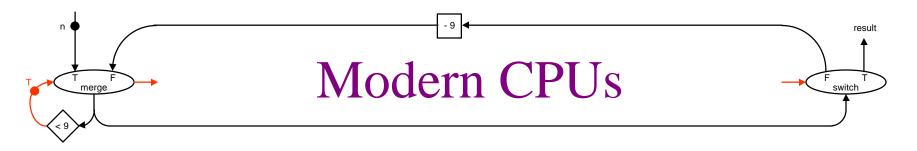
- Id (Monsoon)
 - functional language augmented with I structures
- SISAL (Manchester)
 - Pascal-like single-assignment language
- Both languages enforce a write-once read-many programming style
 - Relatively easy to compile to efficient dataflow code
 - Relatively easy to lose toes



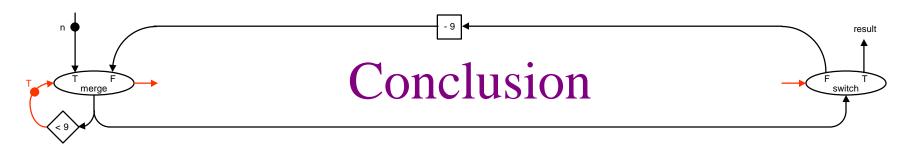
- Some obvious problems:
 - complicated and expensive hardware
 - confusing programming model
- Some not-so-obvious problems:
 - Extremely high bandwidth requirements
 - Each operation injects token(s) into the network
 - Runaway resource requirements
 - Direct consequence of the goal of limitless parallelism
 - Bounding loops works, but doesn't deal with general recursion



- Scheduling Quanta (Iannucci)
 - Use dataflow mechanisms to schedule instructions on a coarser granularity
 - Similar to "Strongly Connected Blocks" (EM-4)
- P-RISC (Nikhil, Arvind)
 - Looks like RISC with a few extra instructions
 - unless you look at the storage model (not recommended)



- Most modern CPUs are really dataflow computers on a small scale!
- Each arc should have a unique name
 - Register renaming!
- Results should be sent directly to ops
 - Out of order instruction buffer!
- Independent ops should execute in parallel
 - Superscalar!



- Advantages of dataflow:
 - Naturally finds all available parallelism
 - Will never stall when there is useful work to do
- Disadvantages of dataflow:
 - Complex hardware
 - Overall performance/efficiency is still worse than a good modern processor
 - Confusing programming model
- Five star rating: $\star \star \star$