Outline



- Part 3: Models of Computation
 - FSMs
 - Discrete Event Systems
 - CFSMs
 - Data Flow Models
 - Petri Nets
 - The Tagged Signal Model

Data-flow networks



- A bit of history
- Syntax and semantics
 - actors, tokens and firings
- Scheduling of Static Data-flow
 - static scheduling
 - code generation
 - buffer sizing
- Other Data-flow models
 - Boolean Data-flow
 - Dynamic Data-flow



Data-flow networks

- Powerful formalism for data-dominated system specification
- Partially-ordered model (no over-specification)
- Deterministic execution independent of scheduling
- Used for
 - simulation
 - scheduling
 - memory allocation
 - code generation
 - for Digital Signal Processors (HW and SW)

A bit of history



- Karp computation graphs ('66): seminal work
- Kahn process networks ('58): formal model
- Dennis Data-flow networks ('75): programming language for MIT DF machine
- Several implementations
 - graphical:
 - Ptolemy (UCB), Khoros (U. New Mexico), Grape (U. Leuven)
 - SPW (Cadence->Coware->Synopsys), COSSAP (H. Meyr, Aachen ->Cadis -> Synopsys)
 - textual:
 - Silage (UCB, Mentor)
 - Lucid, Haskell

Data-flow network



- A Data-flow network is a collection of functional nodes which are connected and communicate over unbounded FIFO queues
- Nodes are commonly called actors
- The bits of information that are communicated over the queues are commonly called tokens



Intuitive semantics



- (Often stateless) actors perform computation
- Unbounded FIFOs perform communication via sequences of tokens carrying values
 - integer, float, fixed point
 - matrix of integer, float, fixed point
 - image of pixels
- State implemented as self-loop
- Determinacy:
 - unique output sequences given unique input sequences
 - Sufficient condition: blocking read
 - (process cannot test input queues for emptiness)

Intuitive semantics



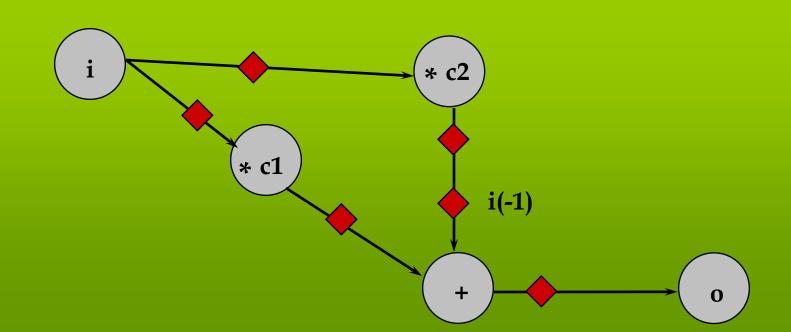
- At each time, one actor is fired
- When firing, actors consume input tokens and produce output \bullet tokens
- Actors can be fired only if there are enough tokens in the input \bullet queues



Intuitive semantics



- Example: FIR filter
 - single input sequence i(n)
 - single output sequence o(n)
 - o(n) = c1 i(n) + c2 i(n-1)



Questions



- Does the order in which actors are fired affect the final result?
- Does it affect the "operation" of the network in any way?
- Go to Radio Shack and ask for an unbounded queue!!

Formal semantics: sequences



- Actors operate from a sequence of input tokens to a sequence of output tokens
- Let tokens be noted by x_1, x_2, x_3 , etc...
- A sequence of tokens is defined as

$$X = [x_1, x_2, x_3, ...]$$

- Over the execution of the network, each queue will grow a particular sequence of tokens
- In general, we consider the actors mathematically as functions from sequences to sequences (not from tokens to tokens)



- Let X₁ and X₂ be two sequences of tokens.
- We say that X₁ is less than X₂ if and only if (by definition) X₁ is an initial segment of X₂
- Homework: prove that the relation so defined is a partial order (reflexive, antisymmetric and transitive)
- This is also called the prefix order
- Example: $[x_1, x_2] \le [x_1, x_2, x_3]$
- Example: $[x_1, x_2]$ and $[x_1, x_3, x_4]$ are incomparable



- Consider the set S of all finite and infinite sequences of tokens
- This set is partially ordered by the prefix order
- A subset C of S is called a chain iff all pairs of elements of C are comparable
- If C is a chain, then it must be a linear order inside S (otherwise, why call it chain?)
- Example: { [x₁], [x₁, x₂], [x₁, x₂, x₃], ... } is a chain
- Example: { [x₁], [x₁, x₂], [x₁, x₃], ... } is not a chain

(Least) Upper Bound



- Given a subset Y of S, an upper bound of Y is an element z of S such that z is larger than all elements of Y
- Consider now the set Z (subset of S) of all the upper bounds of Y
- If Z has a least element u, then u is called the least upper bound (lub) of Y
- The least upper bound, if it exists, is unique (basic property of partial orders)
- Note: u might not be in Y (if it is, then it is the largest value of Y)



- Every chain in S has a least upper bound
- Because of this property, S is called a Complete Partial Order
- Notation: if C is a chain, we indicate the least upper bound of C by lub(C)
- Note: the least upper bound may be thought of as the limit of the chain



Processes

Process: function from a p-tuple of sequences to a q-tuple of sequences

 $F : S^p \rightarrow S^q$

• Tuples have the induced point-wise order:

$$Y = (y_1, \dots, y_p), Y' = (y'_1, \dots, y'_p) \text{ in } S^p :$$

$$Y \le Y' \quad \text{iff } y_i \le y'_i \text{ for all } 1 \le i \le p$$

- Given a chain C in S^p, F(C) may or may not be a chain in S^q
- We are interested in conditions that make that true

Continuity and Monotonicity



- Continuity: F is continuous iff (by definition) for all chains C, lub(F(C)) exists and F(lub(C) = lub(F(C))
- Similar to continuity in analysis using limits
- Monotonicity: F is monotonic iff (by definition) for all pairs X, X' X <= X' => F(X) <= F(X')
- Continuity implies monotonicity
 - intuitively, outputs cannot be "withdrawn" once they have been produced
 - timeless causality: F transforms chains into chains



From Kahn networks to Data Flow networks

- Each process becomes an actor. set of pairs of
 - firing rule
 - (number of required tokens on inputs)
 - function

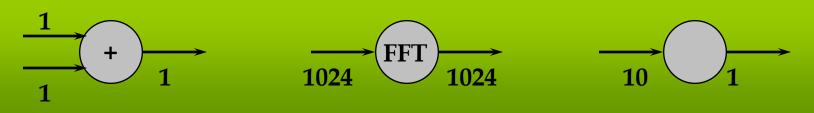
(including number of consumed and produced tokens)

- Formally shown to be equivalent, but actors with firing are more intuitive
- *Mutually exclusive* firing rules (i.e. every actor has only one firing rule active at any given time) imply monotonicity
- Generally simplified to *blocking read*



Examples of Data Flow actors

- SDF: Synchronous (or, better, Static) Data Flow
 - fixed input and output tokens



- BDF: Boolean Data Flow
 - control token determines consumed and produced tokens





Static scheduling of DF

- Key property of DF networks: output sequences do not depend on time of firing of actors
- SDF networks can be *statically scheduled* at compile-time
 - execute an actor when it is *known* to be fireable
 - no overhead due to sequencing of concurrency
 - static buffer sizing
- Different schedules yield different
 - code size
 - buffer size
 - pipeline utilization



Static scheduling of SDF

- Based only on *process graph* (ignores functionality)
- Network state: number of tokens in FIFOs
- Objective: find schedule that is *valid*, i.e.:
 - admissible

(only fires actors when fireable)

– periodic

(brings network back to initial state firing each actor at least once)

• Optimize cost function over admissible schedules



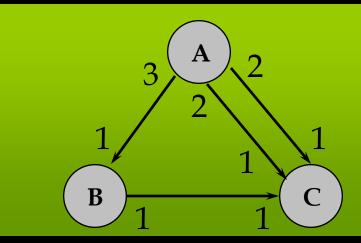
 Number of produced tokens must equal number of consumed tokens on every edge

 $A \xrightarrow{n_p \qquad n_c} B$

- Repetitions (or firing) vector v_S of schedule S: number of firings of each actor in S
- $v_{s}(A) n_{p} = v_{s}(B) n_{c}$

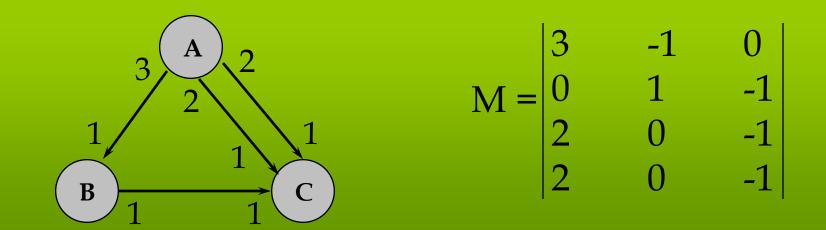
must be satisfied for each edge





- Balance for each edge:
 - $3 v_{S}(A) v_{S}(B) = 0$
 - $v_{S}(B) v_{S}(C) = 0$
 - $-2 v_{\rm S}(A) v_{\rm S}(C) = 0$
 - $2 v_{S}(A) v_{S}(C) = 0$





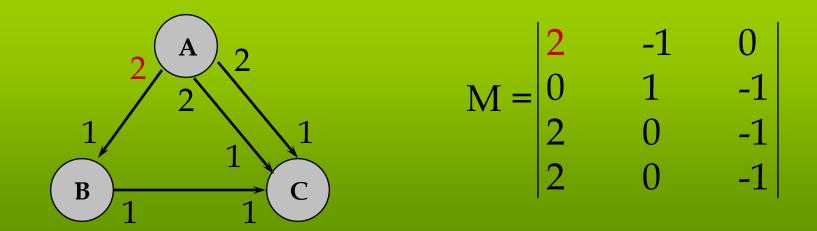
• M v_s = 0

iff S is periodic

- Full rank (as in this case)
 - no non-zero solution
 - no periodic schedule

(too many tokens accumulate on A->B or B->C)





• Non-full rank

- infinite solutions exist (linear space of dimension 1)

- Any multiple of $q = |1 \ 2 \ 2|^T$ satisfies the balance equations
- ABCBC and ABBCC are minimal valid schedules
- ABABBCBCCC is non-minimal valid schedule



Static SDF scheduling

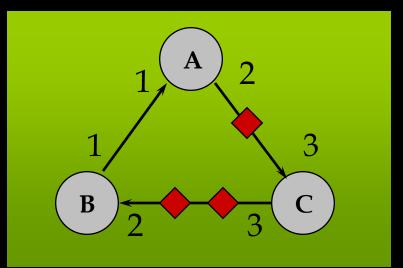
- Main SDF scheduling theorem (Lee '86):
 - A connected SDF graph with *n* actors has a periodic schedule iff its topology matrix M has rank *n*-1
 - If M has rank *n-1* then there exists a unique smallest integer solution q to

M q = 0

- Rank must be at least *n-1* because we need at least *n-1* edges (connected-ness), providing each a linearly independent row
- Admissibility is not guaranteed, and depends on initial tokens on cycles



Admissibility of schedules



• No admissible schedule:

BACBA, then deadlock...

Adding one token (delay) on A->C makes

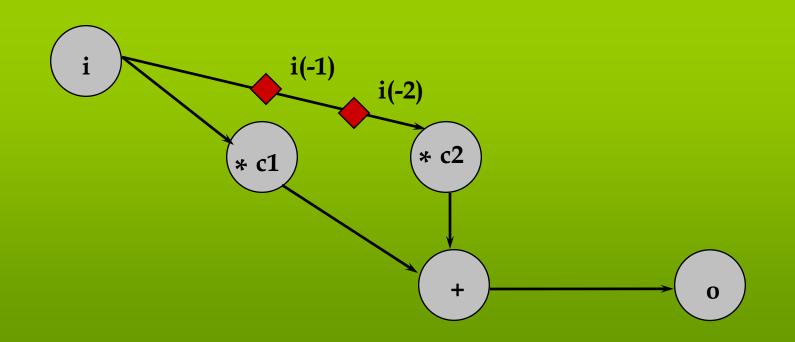
BACBACBA valid

• Making a periodic schedule admissible is always possible, but changes specification...



Admissibility of schedules

• Adding initial token changes FIR order

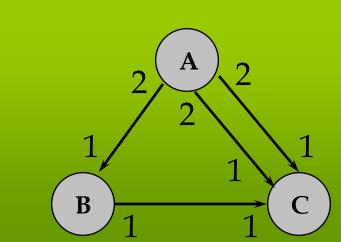




From repetition vector to schedule

Repeatedly schedule fireable actors up to number of times in repetition vector

 $q = |1 \ 2 \ 2|^{T}$



- Can find either ABCBC or ABBCC
- If deadlock before original state, no valid schedule exists (Lee '86)



From schedule to implementation

- Static scheduling used for:
 - behavioral simulation of DF (extremely efficient)
 - code generation for DSP
 - HW synthesis (Cathedral by IMEC, Lager by UCB, ...)
- Issues in code generation
 - execution speed (pipelining, vectorization)
 - code size minimization
 - data memory size minimization (allocation to FIFOs)
 - processor or functional unit allocation



Compilation optimization

• Assumption: code stitching

(chaining custom code for each actor)

- More efficient than C compiler for DSP
- Comparable to hand-coding in some cases
- Explicit parallelism, no artificial control dependencies
- Main problem: memory and processor/FU allocation depends on scheduling, and vice-versa



Code size minimization

- Assumptions (based on DSP architecture):
 - subroutine calls expensive
 - fixed iteration loops are cheap

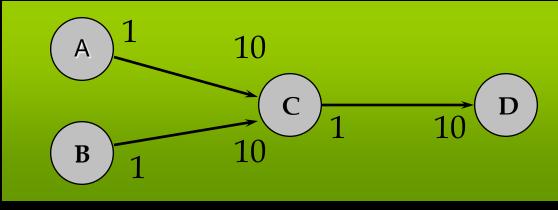
("zero-overhead loops")

- Absolute optimum: single appearance schedule
 e.g. ABCBC -> A (2BC), ABBCC -> A (2B) (2C)
 - may or may not exist for an SDF graph...
 - buffer minimization relative to single appearance schedules (Bhattacharyya '94, Lauwereins '96, Murthy '97)



Buffer size minimization

- Assumption: no buffer sharing
- Example:



q = | 100 100 10 1|^T

- Valid SAS: (100 A) (100 B) (10 C) D
 - requires 210 units of buffer area
- Better (factored) SAS: (10 (10 A) (10 B) C) D
 - requires 30 units of buffer areas, but...
 - requires 21 loop initiations per period (instead of 3)

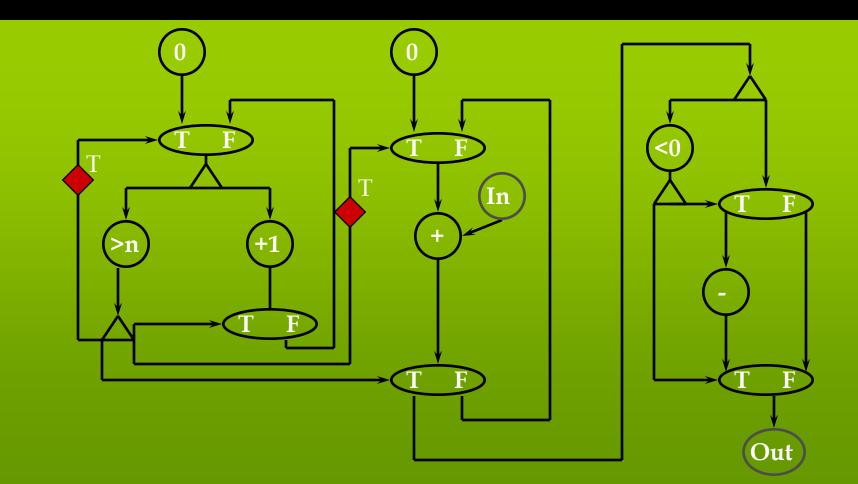
Dynamic scheduling of DF



- SDF is limited in modeling power
 - no run-time choice
 - cannot implement Gaussian elimination with pivoting
- More general DF is too powerful
 - non-Static DF is Turing-complete (Buck '93)
 - bounded-memory scheduling is not always possible
- BDF: semi-static scheduling of special "patterns"
 - if-then-else
 - repeat-until, do-while
- General case: thread-based dynamic scheduling
 - (Parks '96: may not terminate, but never fails if feasible)

Example of Boolean DF

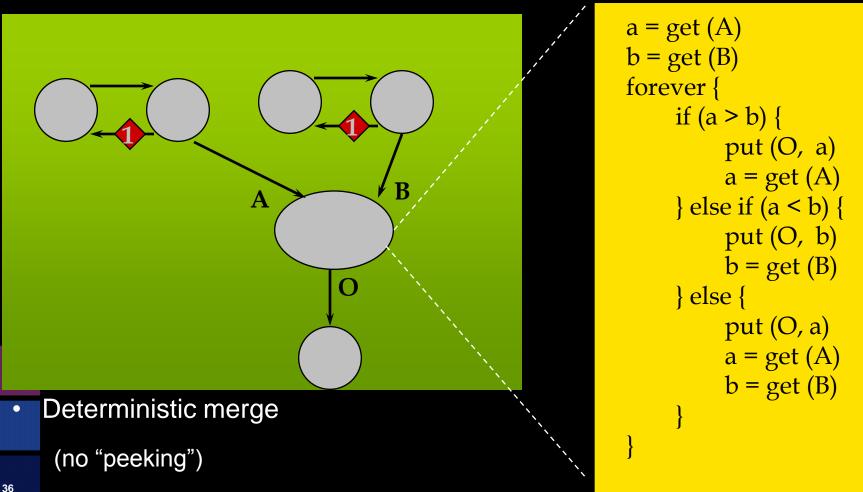
• Compute absolute value of average of *n* samples





Example of general DF

- Merge streams of multiples of 2 and 3 in order (removing duplicates)



Summary of DF networks

• Advantages:

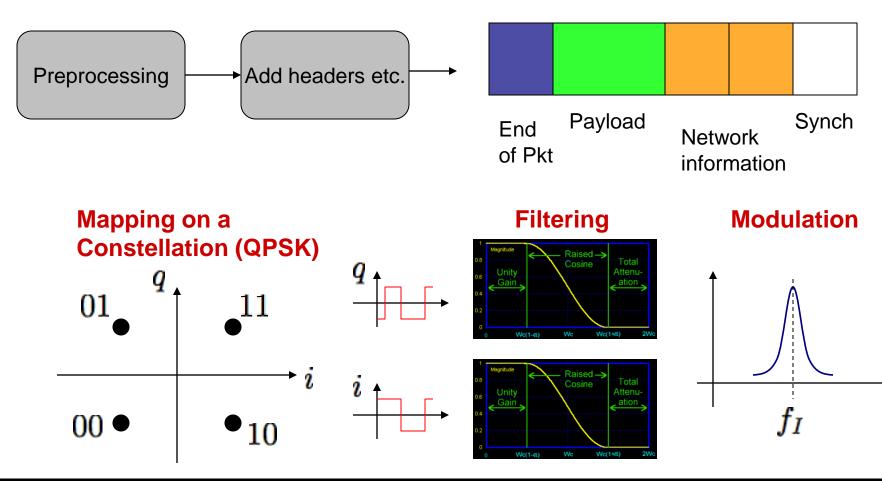
- Easy to use (graphical languages)
- Powerful algorithms for
 - verification (fast behavioral simulation)
 - synthesis (scheduling and allocation)
- Explicit concurrency
- Disadvantages:
 - Efficient synthesis only for restricted models
 - (no input or output choice)
 - Cannot describe reactive control (blocking read)

Base-band Processing in Cell Phones

QuickTime?and a TIFF (Uncompressed) decompressor are needed to see this picture.

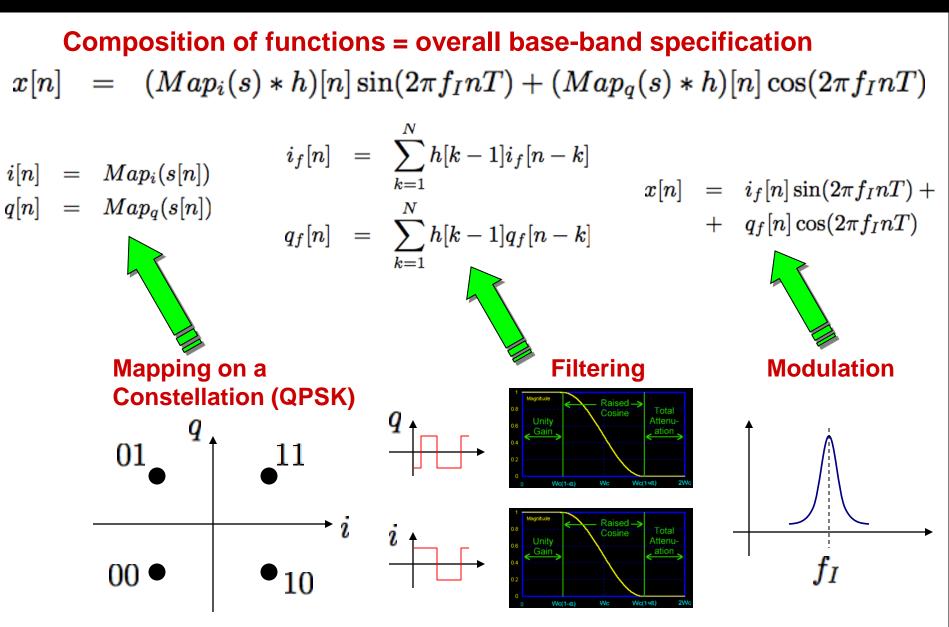






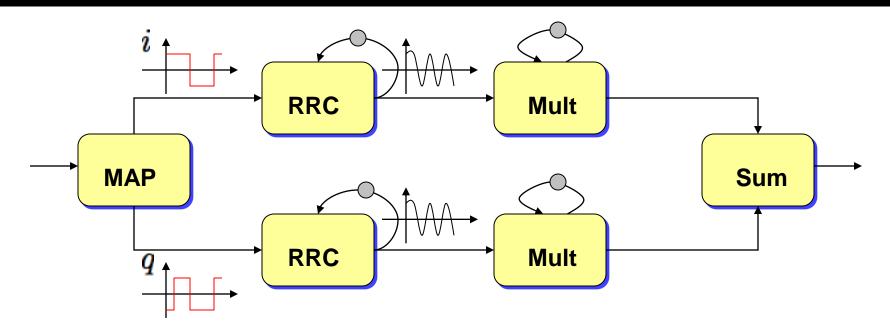
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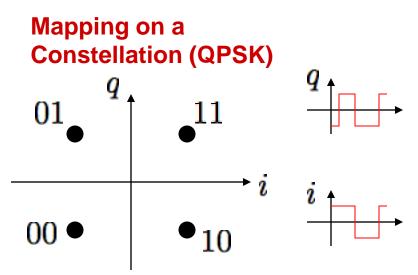
Base-band Processing: Denotation



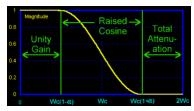
Base-band Processing: Data Flow Model

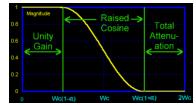




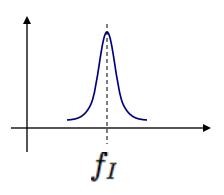


Filtering





Modulation



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Remarks



- Composition is achieved by input-output connection through communication channels (FIFOs)
- The operational semantics dictates the conditions that must be satisfied to execute a function (actor)
- Functions operating on streams of data rather than states evolving in response to traces of events (data vs. control)
- Convenient to mix denotational and operational specifications

Telecom/MM applications



- Heterogeneous specifications including
 - data processing
 - control functions
- Data processing, e.g. encryption, error correction...
 - computations done at regular (often short) intervals
 - efficiently specified and synthesized using DataFlow models
- Control functions (data-dependent and real-time)
 - say when and how data computation is done
 - efficiently specified and synthesized using FSM models
- Need a common model to perform global system analysis and optimization

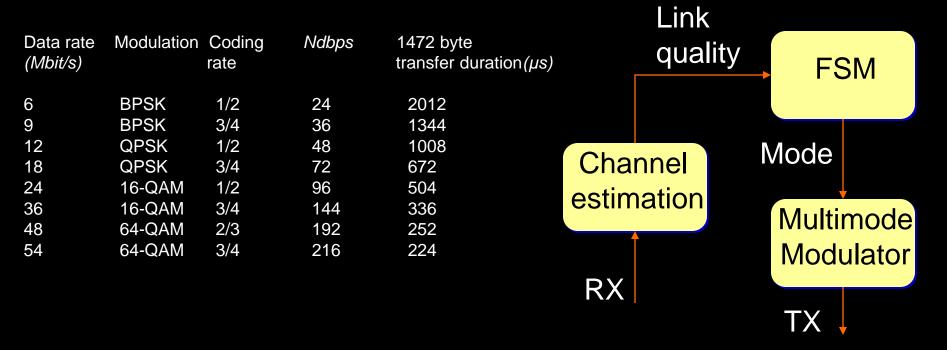
Mixing the two models: 802.11b



- State machine for control
 - Denotational: processes as sequence of events, sequential composition, choice etc.
 - Operational: state transition graphs
- Data Flow for signal processing
 - Functions
 - Data flow graphs
- And what happens when we put them together?

802.11b: Modes of operation





- Depending on the channel conditions, the modulation scheme changes
- It is natural to mix FSM and DF (like in figure)
- Note that now we have real-time constraints on this system (i.e. time to send 1472 bytes)

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