ECE 6775 High-Level Digital Design Automation Fall 2023

Scheduling

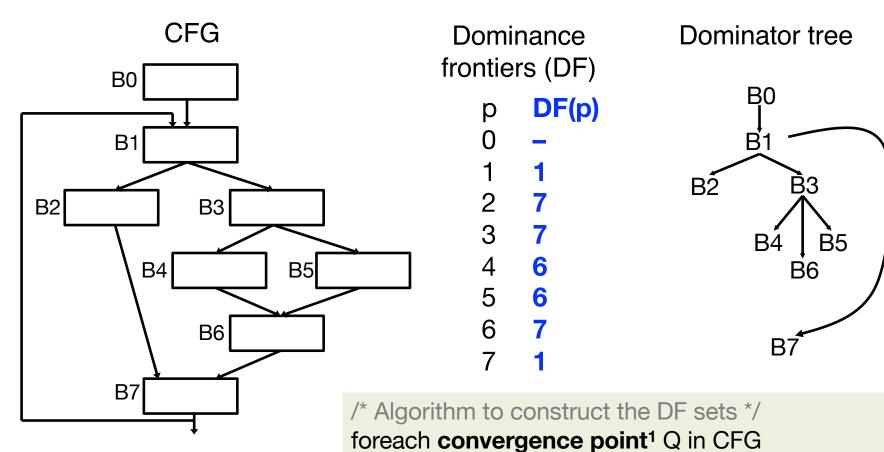




Agenda

- Unconstrained scheduling
 - ASAP and ALAP
- Constrained scheduling
 - Resource constrained scheduling (RCS)
 - Exact formulations with integer linear programming (ILP)

Review: Dominance Frontiers



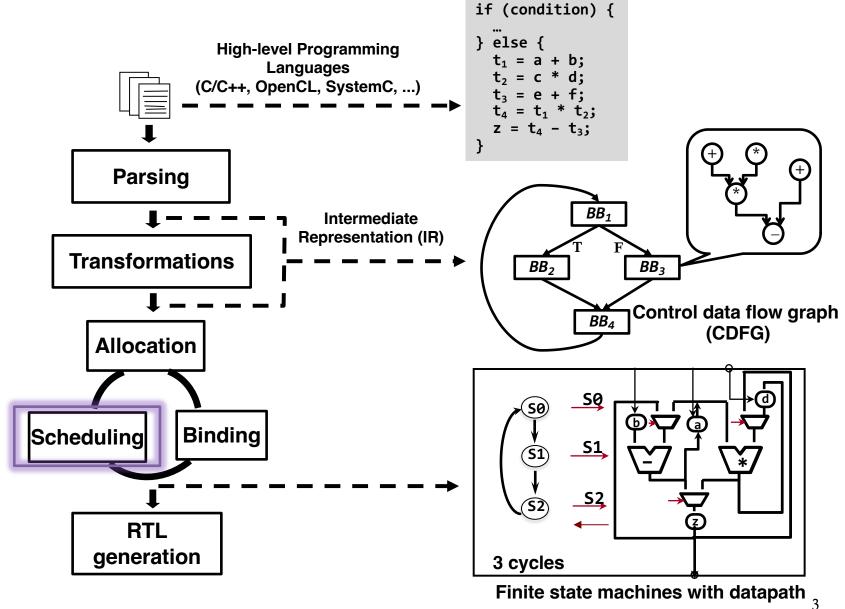
¹**convergence point** is a node in CFG with multiple predecessors

Run up to **Y=IDOM(Q)** in the dominator tree, adding Q to DF(P) for each P between [X, Y)

foreach predecessor X of Q in CFG

Only convergence points are added to the DF sets!

Recap: A Typical HLS Flow



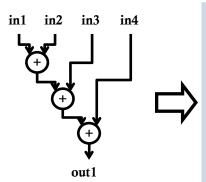
Importance of Scheduling

- Scheduling is a central problem in HLS
 - Introduces clock boundaries to untimed (or partially timed) input specification
 - Has significant impact on the quality of results
 - Frequency
 - Latency
 - Throughput
 - Area
 - Power

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Scheduling: Untimed to Timed

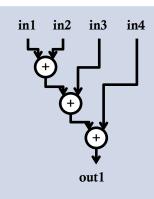
Untimed



Control-Data Flow Graph (CDFG)

out1 = f(in1, in2, in3, in4)

Combinational for Latency

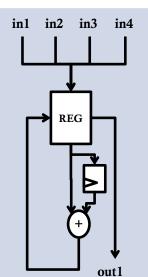


$$t_{clk} \approx 3 * d_{add}$$

$$T_{I} = 1 / t_{clk}$$

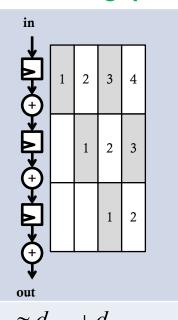
$$A_{1} = 3 * A_{add}$$

Sequential for Area



$$t_{clk} \approx d_{add} + d_{setup}$$
 $t_{clk} \approx d_{add}$
 $T_2 = 1/(3*t_{clk})$ $T_3 = 1/t_{clk}$
 $A_2 = A_{add} + 2*A_{reg}$ $A_3 = 3*A_{add}$

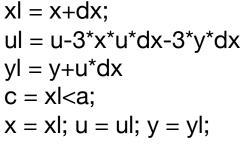
Pipelined for Throughput

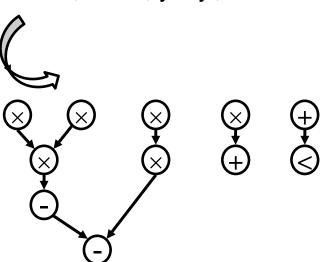


$$t_{clk} \approx d_{add} + d_{setup}$$
 $t_{clk} \approx d_{add} + d_{setup}$ $T_2 = 1/(3*t_{clk})$ $T_3 = 1/t_{clk}$ $A_2 = A_{add} + 2*A_{reg}$ $A_3 = 3*A_{add} + 6*A_{reg}$

Scheduling Input

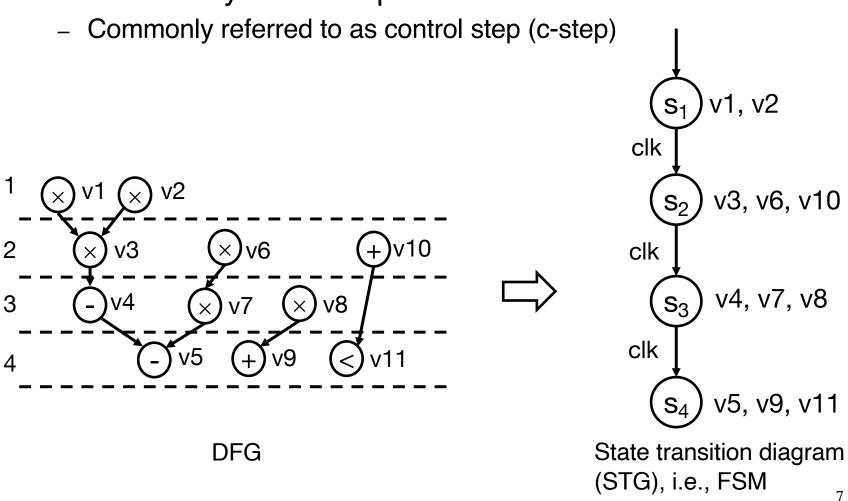
- Control data flow graph (CDFG)
 - Generated by a compiler front end from a high-level specification
 - Nodes: basic blocks & operations
 - Directed edges: data & control dependencies
- Without control flow, the basic structure is a data flow graph (DFG)





Scheduling Output

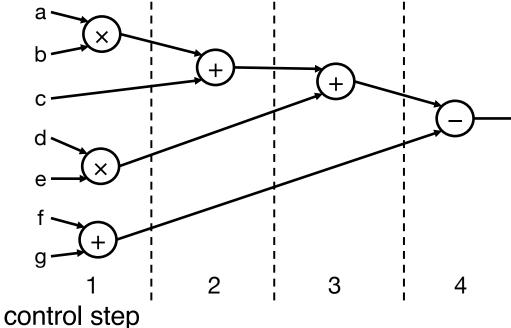
- Scheduling: map operations to states
- Each clock cycle corresponds to a state in the FSM



Unconstrained Scheduling

- Only consideration: dependence
- As soon as possible (ASAP)
 - Schedule an operation to the earliest possible step
- As late as possible (ALAP)
 - Schedule an operation to the earliest possible step, without increasing the total latency

ASAP Schedule



Assumption for simplicity:

Combinational chaining of multiple operations is not allowed (each operation occupies the full cycle)

$$Y = ((a*b)+c)+(d*e)-(f+g)$$

ASAP(G(V, E)):

V' = Topological_Sort(G)

foreach v_i in V':

// Primary inputs (PIs) to first cycle

if $v_i \in Pls$: $t_i = 1$

// Assume no chaining & single-cycle operations

else: $t_i = max_{i \in pred(i)} \{t_i + 1\}; // (v_i, v_i) \in E$

The start time for each operation is the least one allowed by the dependencies

ALAP Schedule

a b + + + + g 1 2 3 4 control step

Assumption for simplicity:

Combinational chaining of multiple operations is not allowed (each operation occupies the full cycle)

$$Y = ((a*b)+c)+(d*e)-(f+g)$$

ALAP(G(V, E), L): // L is the latency bound

V' = Reverse_Topological_Sort(G)

foreach v_i in V':

// Primary outputs (POs) to last cycle

if $v_i \in POs: t_i = L$

// Assume no chaining & single-cycle operations

else: $t_i = min_{k \in SUCC(i)} \{t_k\} - 1$; // $(v_i, v_k) \in E$

The end time of each operation is the latest one allowed by the dependencies and the latency constraint

Constrained Scheduling in HLS

- Constrained scheduling
 - General case NP-hard
 - Resource-constrained scheduling (RCS)
 - Minimize latency given constraints on area or resources
 - Time-constrained scheduling (TCS)
 - Minimize resources subject to bound on latency
- Exact methods
 - Integer linear programming (ILP)
 - Hu's algorithm for a very restricted problem
- Heuristics
 - List scheduling
 - Force-directed scheduling
 - SDC-based scheduling

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Linear Programming

- Linear programming (LP) solves the problem of maximizing or minimizing a linear objective function subject to linear constraints
 - Efficiently solvable both in theory and in practice
- Integer linear programming (ILP): in addition to linear constraints and objective, the values for the variables must be integer
 - NP-Hard in general (A special case, 0-1 ILP)
 - Modern ILP solvers can handle problems with nontrivial size
- Enormous number of problems can be expressed in LP or ILP

Canonical Form of ILP

```
\begin{array}{lll} \text{maximize} & c_1x_1 + c_2x_2 + \ldots + c_nx_n \text{ // objective function} \\ \text{subject to} & \text{// linear constraints} \\ & a_{11}x_1 + a_{12}x_2 + \ldots + a_{1n}x_n \leq b_1 \\ & a_{21}x_1 + a_{22}x_2 + \ldots + a_{2n}x_n \leq b_2 \\ & \ldots \\ & a_{m1}x_1 + a_{m2}x_2 + \ldots + a_{mn}x_n \leq b_m \\ & x_i \geq 0 \\ & x_i \in Z \end{array} \begin{array}{ll} x_i + a_{m2}x_1 + a_{m2}x_2 + \ldots + a_{mn}x_n \leq b_m \\ & x_i \leq 0 \\ & x_i \leq 0 \end{array} x_i \leq 0 x_i
```

Vector form

```
maximize \mathbf{c}^T \mathbf{x} / / \mathbf{c} = (c_1, c_2, ..., c_n), \mathbf{x} = (x_1, x_2, ..., x_n)
subject to \mathbf{A} \mathbf{x} \leq \mathbf{b} \qquad / / \mathbf{A} \text{ is a } m \mathbf{x} n \text{ matrix}; \ \mathbf{b} = (b_1, b_2, ..., b_n)\mathbf{x} \geq \mathbf{0}\mathbf{x}_i \in \mathbf{Z}
```

Example: Course Selection Problem

- A student is about to finalize course selection for the coming semester, given the following requirement
 - Minimum credits per semester: 8

| | Schedule | Credits | Est. workload (per week) |
|----------------------------|-----------------|---------|-----------------------------|
| 1. Metaverse | MW 2:00-3:30pm | 3 | 8 hrs |
| 2. How to start a start-up | TT 2:00-3:00pm | 2 | 4 hrs |
| 3. Linear programming (LP) | MW 9:00-11:00am | 4 | 10 hrs |
| 4. Analog circuits | TT 1:00-3:00pm | 4 | 12 hrs |

Question: Which courses should this student choose to minimize workload?

ILP Formulation for Course Selection

Define decision variables (i = 1, 2, 3, 4):

| | Time | CRs | Work |
|--------------|-----------------|-----|--------|
| 1. Metaverse | MW 2- 3:30pm | 3 | 8 hrs |
| 2. Start-up | TT 2-3pm | 2 | 4 hrs |
| 3. LP | MW 9- 11am | 4 | 10 hrs |
| 4. Analog | TT 1-3pm | 4 | 12 hrs |

 $8x_1 + 4x_2 + 10x_3 + 12x_4$

 $3x_1+2x_2+4x_3+4x_4$

- Total expected work hours:
- Total credits taken:
- Account for the schedule conflict:
- Complete ILP formulation (in canonical form): minimize 8x₁+4x₂+10x₃+12x₄

s.t.
$$3x_1+2x_2+4x_3+4x_4 \ge 8$$

 $x_2+x_4 \le 1$
 $x_i \in \{0,1\}$

Resource Constrained Scheduling (RCS)

- When functional units are limited
 - Each functional unit can only perform one operation at each clock cycle
 - e.g., if there are only K adders, no more than K additions can be executed in the same c-step
- A typical resource-constrained scheduling problem for DFG
 - Given the number of functional units of each type, minimize latency (in cycles)
 - NP-hard

ILP Formulation of RCS: Binary Variables

- ▶ Binary decision variables $x_{i,k}$
 - $-x_{i,k}=1$ if operation i starts at step k, otherwise = 0
 - $1 \le i \le N, 1 \le k \le L$
 - N is the total number of operations
 - L is the given upper bound on latency

ILP Formulation of RCS: Derived Variables

- ▶ Binary decision variables $x_{i,k}$
 - $-x_{i,k}=1$ if operation i starts at step k, otherwise = 0
 - $1 \le i \le N, 1 \le k \le L$
 - N is the total number of operations
 - L is the given upper bound on latency
- ightharpoonup Derived integer variables t_i

$$t_i = \sum_{k=1}^{L} k \cdot x_{i,k}$$

 t_i indicates the actual **start time** of operation i

ILP Formulation of RCS: Constraints (1)

Unique start times: an operation must start at one and only one of the available steps

$$\forall i \in [1, N] : \sum_{k} x_{i,k} = 1$$

Dependence must be satisfied (assuming no chaining)

$$\forall (i,j) \in E: t_j \geq t_i + d_i \rightarrow \sum_k k \cdot x_{j,k} \geq \sum_k k \cdot x_{i,k} + d_i$$

Operation *j* must not start before *i* completes if *j* depends on *i*

 d_i : latency of operation i

- $d_i = 1$ means a single-cycle operation
- $d_i > 1$ indicates a multi-cycle operation

Start Time vs. Active Time(s)

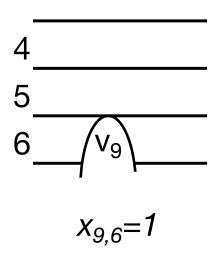
- When $d_i = 1$, the following are the same
 - Does operation i start at step k?
 - Is operation i actively running at step k?
 - Same equality check: if x_{ik} is 1
- When $d_i > 1$, the following questions are different
 - Does operation i start at step k? Simply check if x_{ik} is 1
 - Is operation i active at step k? Check if the following holds

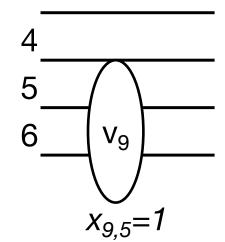
$$\sum_{l=k-d_i+1}^k x_{i,l} \stackrel{?}{=} 1$$

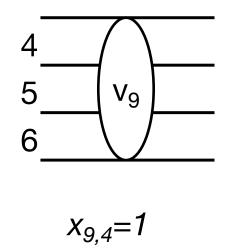
Is Operation i Still Active at Step k?

▶ Is operation 9 active (running) at step 6? assuming $d_9 = 3$

if and only if
$$x_{9,6} + x_{9,5} + x_{9,4}$$
 equals 1 $\Rightarrow \sum_{l=6-3+1}^{6} x_{9,l} \stackrel{?}{=} 1$







- Notes
 - Only one (if any) of the above three cases can happen
 - To meet resource constraints, we must check the same equality for ALL steps, and ALL operations of that type

ILP Formulation of RCS: Constraints (2)

- Physical resource limits
 - R denotes the total number of different resource types (RT)
 - $-RT(i) \in [1,R]$ is the resource type of operation i
 - $-a_r$ is the number of physically available resources of type r

At any step k, the total number of active operations of the same type r must not exceed a_r , the number of physically available resources for type r

$$\forall k \in [1, L], \forall r \in [1, R]: \sum_{i:RT(i)=r}^{k} \sum_{l=k-d_i+1}^{k} x_{i,l} \leq a_r$$
 Sum over operations of resource r If operation i is active at step k

The above summation counts the number of operations active at step *k* that use resource *r*

ILP Formulation of RCS: Putting It Together

Unique start times

$$\forall i \in [1, N] : \sum_{k} x_{i,k} = 1$$

Dependence must be satisfied (assuming no chaining)

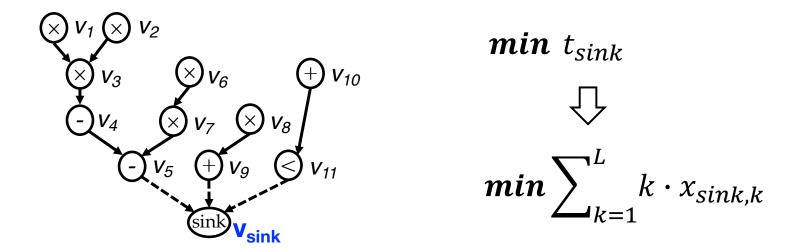
$$\forall (i,j) \in E : t_j \ge t_i + d_i \to \sum_k k \cdot x_{j,k} \ge \sum_k k \cdot x_{i,k} + d_i$$

Physical resource limits

$$\forall k \in [1, L], \forall r \in [1, R] : \sum_{i:RT(i)=r} \sum_{l=k-d_i+1}^{k} x_{i,l} \le a_r$$

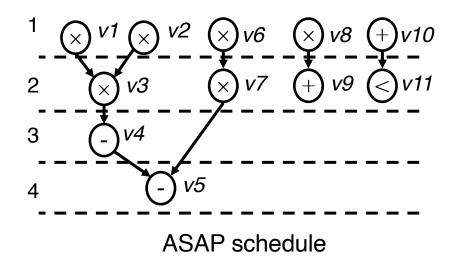
ILP Formulation of RCS: Objective Function

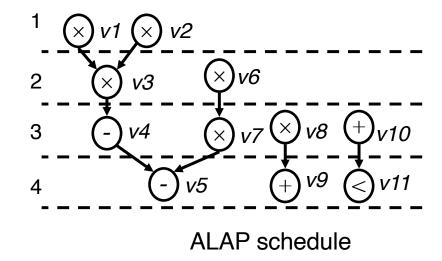
- For simplicity, we introduce a pseudo node v_{sink} to serve as a unique sink of the DFG
 - This node depends on all the original primary output nodes
- To minimize the overall latency, we simply minimize the start time of the sink node



Use of ASAP and ALAP

- In general, the following helps the ILP solver run faster
 - Minimize # of variables and constraints
 - Simplify the constraints
- We can write the ILP without ASAP/ALAP, but using ASAP and ALAP can simplify the constraints





ILP Formulation: Unique Start Time Constraints

$$\begin{vmatrix} x_{il} = 0 & for & l < t_i^S & and & l > t_i^L \\ (t_i^S = ASAP(v_i), t_i^L = ALAP(v_i)) \end{vmatrix}$$

Without using ASAP and ALAP

$$x_{1,1} + x_{1,2} + x_{1,3} + x_{1,4} = 1$$

$$x_{2,1} + x_{2,2} + x_{2,3} + x_{2,4} = 1$$

$$...$$

$$x_{6,1} + x_{6,2} + x_{6,3} + x_{6,4} = 1$$

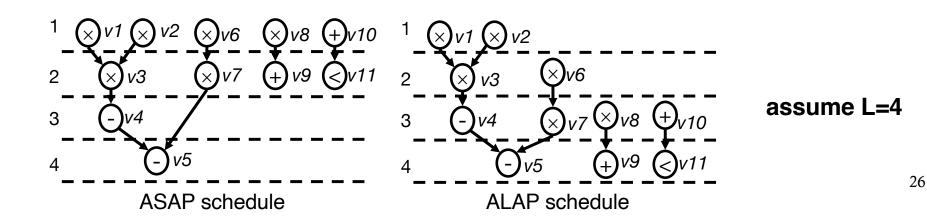
$$...$$

$$x_{9,1} + x_{9,2} + x_{9,3} + x_{9,4} = 1$$

Using ASAP and ALAP

$$x_{1,1} = 1$$

$$x_{2,1} = 1$$
...
$$x_{6,1} + x_{6,2} = 1$$
...
$$x_{9,2} + x_{9,3} + x_{9,4} = 1$$



ILP Formulation: Dependence Constraints

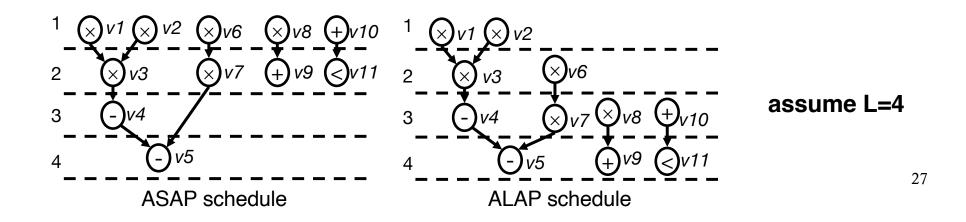
 Using ASAP and ALAP, the non-trivial inequalities are: (assuming no chaining and single-cycle ops)

$$2x_{7,2} + 3x_{7,3} - x_{6,1} - 2x_{6,2} \ge 1$$

$$4x_{5,4} - 2x_{7,2} - 3x_{7,3} \ge 1$$

$$2x_{9,2} + 3x_{9,3} + 4x_{9,4} - x_{8,1} - 2x_{8,2} - 3x_{8,3} \ge 1$$

$$2x_{11,2} + 3x_{11,3} + 4x_{11,4} - x_{10,1} - 2x_{10,2} - 3x_{10,3} \ge 1$$

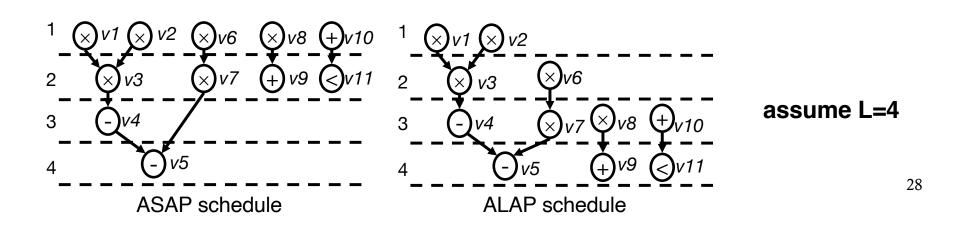


ILP Formulation: Resource Constraints

Resource constraints (assuming 2 multipliers and 1 ALU)

$$\begin{aligned} x_{1,1} + x_{2,1} + x_{6,1} + x_{8,1} &\leq 2 \\ x_{3,2} + x_{6,2} + x_{7,2} + x_{8,2} &\leq 2 \\ x_{7,3} + x_{8,3} &\leq 2 \end{aligned}$$

$$\begin{aligned} x_{10,1} &\leq 1 \\ x_{9,2} + x_{10,2} + x_{11,2} &\leq 1 \\ x_{4,3} + x_{9,3} + x_{10,3} + x_{11,3} &\leq 1 \\ x_{5,4} + x_{9,4} + x_{11,4} &\leq 1 \end{aligned}$$



ILP Summary

- Pros: versatile modeling ability
 - Can be extended to handle almost every design aspect
 - Resource allocation
 - Module selection
 - Area, power, etc.
- Cons: computationally expensive
 - #variables = O(#nodes * #steps)
 - 0-1 variables: need extensive search to find optimal solution

Next Lecture

More scheduling algorithms

Acknowledgements

- These slides contain/adapt materials developed by
 - Ryan Kastner (UCSD)