

30 Algorithms and Data Structures in 60 Minutes

Chris Lomont, May 2019

Introduction

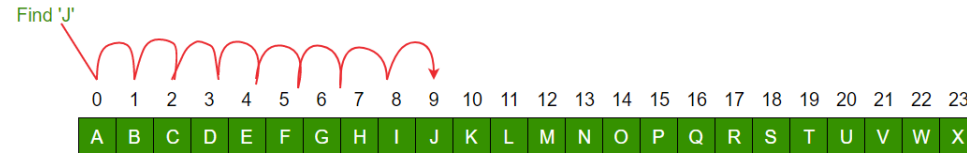
- Don't learn all details at once; simply learn they exist
- **Algorithms in blue**
- **Techniques in green** – general principles for algorithm design
- **Challenges in red**
- In practice, most problems solved with language and library features
- When something hard, need more tools
- Items start easy, get harder

Sorting

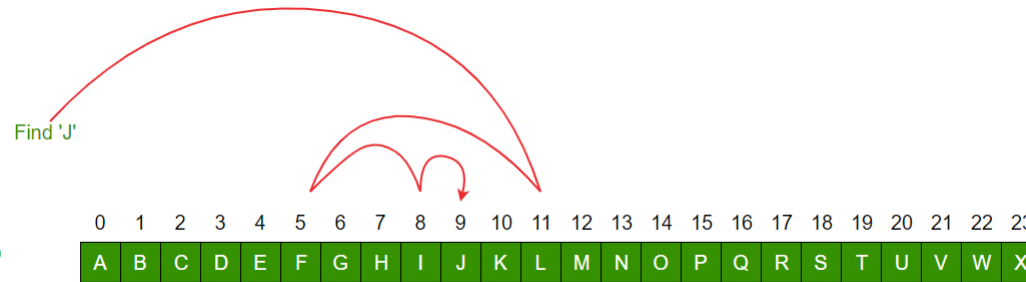
- **Bubble sort** - $O(n^2)$ - still good for small sizes
- **Quicksort** - $O(n \log n)$ expected, $O(n^2)$ worst – randomized pivot – adversary possible – **Technique: divide and conquer**
- Comparison based sort provably $O(n \log n)$
 - So don't compare. Index instead 😊
- **Radix sort** - $O(nk) \sim O(n)$, either direction, good when sparse, k passes
- **Bucket sort** - $O\left(n + \frac{n^2}{k} + k\right) \sim O(n)$, MSD to LSD, good when dense
- Other sorts for different scales: disk, internet scale, cache aware...
- **Timsort** – state of art, uses known ordered pieces to direct future work.

Searching sorted array

- **Linear** – $O(n)$



- **Binary** – $O(\log n)$
 - **divide and conquer**



- **Fibonacci** – $O(\log n)$ - move Fibonacci sized ratios – can be better
- **Dictionary** – Can be $O(1)$ for well structured data, often $O(\log \log n)$

String searching

- Search for length m word in length n text over size k alphabet.

Algorithm	Preprocess	Matching	Space
Naïve	None	$O(nm)$	none
Knuth-Morris-Pratt (KMP)	$O(m)$	$O(n)$	$O(m)$
Boyer-Moore (BM)	$O(m + k)$	Best $O(n/m)$, worst $O(mn)$	$O(k)$
Aho-Corasick (AC)		$O(n + m)$	

- KMP: when a mismatch occurs, the word itself allows skipping ahead, bypassing re-examination of previously matched characters
- BM: **preprocesses** the word for speedy decisions, useful when word searched multiple times across multiple texts. **Technique: time-space tradeoff**
- AC: find all matches to a set of words

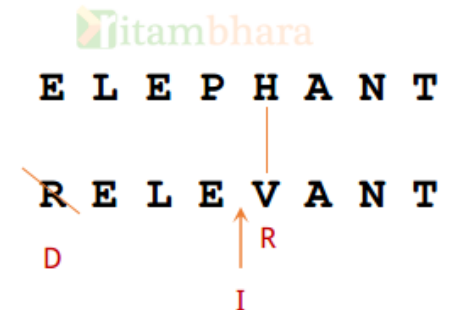
Sound matching

- **Soundex** – matches English words phonetically, patented 1918 and 1922
 - In SQL, many languages
- Algorithm
 - Keep first letter of word
 - Drop all other a, e, i, o, u, y, h, w
 - Except first, Replace with numbers:
 - b,f,p,v -> 1
 - c,g,j,k,q,s,x,z -> 2
 - d,t -> 3
 - L -> 4
 - m,n->5
 - r->6
 - Collapse certain duplicate numbers
 - Truncate to three numbers, or extend with zeroes
- Ex: Robert and Rupert -> R163, Rubin ->R150

String similarity

- Is the string “**kitten**” closer to “**sitting**” or to “**potato**”?
- **Levenshtein distance** : # of changes, insert, deletes between strings.
 - Also called **edit distance**.
- Can do in matrix, once cell at a time
 - $\text{Cell}[i,j] = \min \{ \text{up} + 1, \text{left} + 1, \text{diag} + \chi(i,j) \}$
where $\chi(i,j)$ is $\text{str1}[i] \neq \text{str2}[j]$
- Can reduce to 2 rows of memory
- **Technique: dynamic programming**
 - A form of **memorizing**
 - Very powerful way to design solutions.
- **Challenge**: time Fibonacci numbers with and without memorizing.

		E	L	E	P	H	A	N	T
	0	1	2	3	4	5	6	7	8
R	1	1	2	3	4	5	6	7	8
E	2	1	2	2	3	4	5	6	7
L	3	2	1	2	3	4	5	6	7
E	4	3	2	1	2	3	4	5	6
V	5	4	3	2	2	3	4	5	6
A	6	5	4	3	3	3	3	4	5
N	7	6	5	4	4	4	4	3	4
T	8	7	6	5	5	5	5	4	3

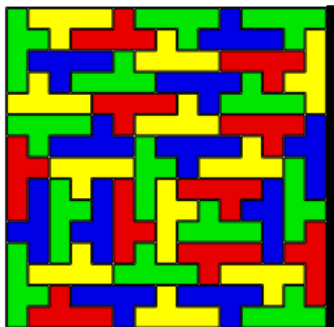


Min Edit Distance: 3

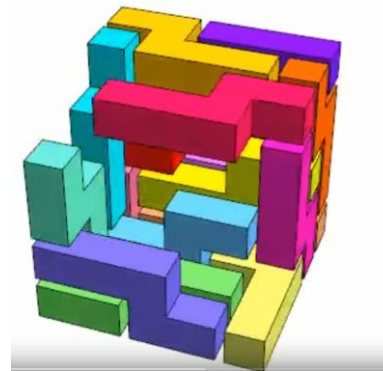
Backtracking search

- Want to search some finite space, discrete moves
- **Dancing Links** (Knuth popularized, 1979)
 - Items represented as grid of doubly linked nodes
 - DoMove/UndoMove each two pointer changes
 - Sudoku solver, combinatorial solvers
 - Heuristic S – choose branch with fewest subproblems
~exponential gain on depth, linear in cost
 - **Technique: trim choices heuristic**

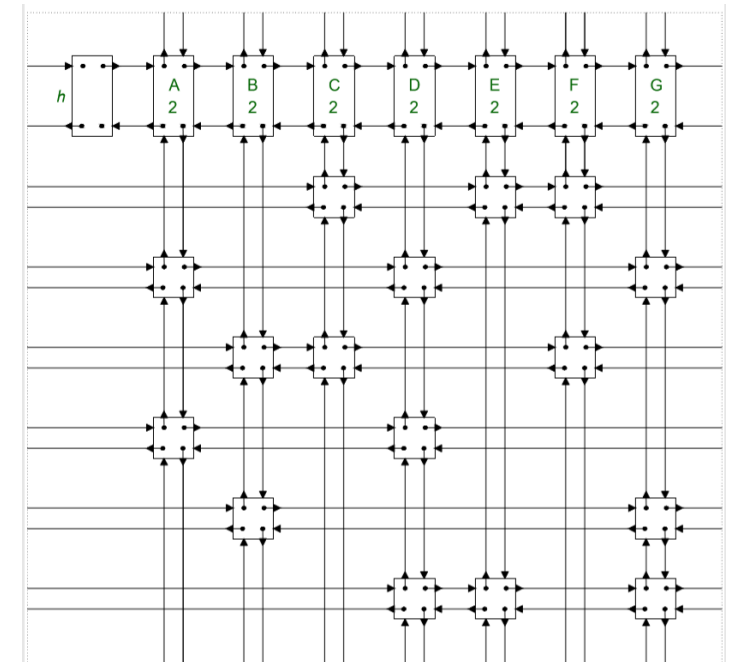
• Ex



20 solutions, 6,375,335 nodes, 806,699,079 updates



```
void Solve(State state, int depth) {  
    if (Solved(state)) {  
        Log(state, depth);  
        return; }  
    auto moves = GenerateMoves(state);  
    for (auto m : moves) {  
        DoMove(state, m);  
        Solve(state, depth + 1);  
        UndoMove(state, m); }  
}
```



Randomness

```
int getRandomNumber()  
{  
    return 4; // chosen by fair dice roll.  
             // guaranteed to be random.  
}
```

- Understand your use cases
- C/C++ rand() generally not very good or fast
- **Mersenne Twister** outdated and has bad properties
- **PCG** series for *non-crypto* pretty good, small, many tunable sizes, **very fast**
- Be careful:
 - **rand()%n** is not uniform
 - **(int)(rand01()*N)** is not uniform
 - To pick [0,N-1] uniformly, take rand and randMax, sample till in multiple of range, **then** mod
 - Assumes rand() uniform over randMax, not always true!
- **Challenge**: implement rand3 from rand2.
- **Challenge**: implement a fair coin toss from a biased coin.
- **Challenge**: implement uniform rand 1 to N.

Shuffle and sample

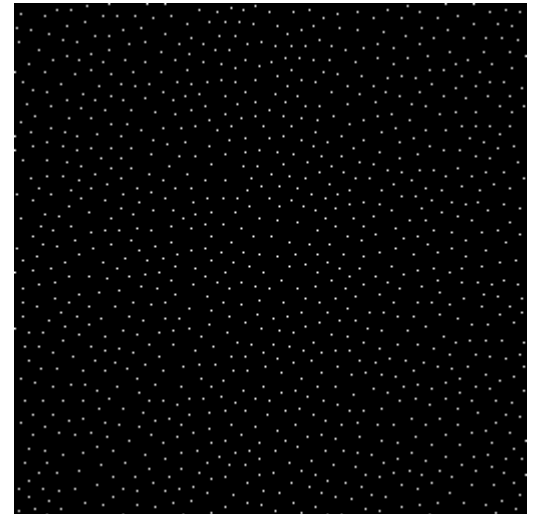
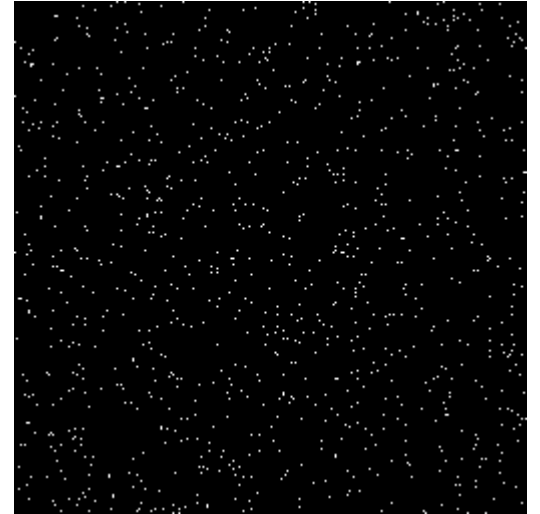
- Common error: shuffle all 1 to N. Not uniform

```
for (int i = 0; i < max; ++i)
    int j = rand() % max;
    swap(item[i], item[j]);
```

- **Fischer-Yates shuffle**

```
for (int i = 0; i < n-1; ++i)
    int j = randN(i, n); //  $i \leq j < n$ 
    swap(item[i], item[j]);
```

- Psychological issues
 - **Shuffle bag**, **Fibonacci shuffle** (not sure of name)
 - Sphere sampling, area sampling (**blue noise**)



Sampling

- Problem: stream of data, want uniform random sample size k , don't know length, cannot store.
- **Reservoir sampling:**
 - Keep first k
 - For each following sample, random to see if it should be in reservoir
 - $J = \text{rand}(1, i)$ inclusive on seeing i th sample.
 - If $j \leq k$ then $R[j] = S[i]$
- **Challenge:** can do with exponentially fewer rand calls.
 - anecdote of solving something yourself

Randomized algorithms



- Use **randomness** as a resource in an algorithm. Sometimes gives speedups
- There are efficient randomized algorithms for which no efficient non-randomized algorithm is known
- Unknown whether $P = BPP$: unknown is an arbitrary randomized algorithm that runs in polynomial time with a small error probability can be derandomized to run in polynomial time without using randomness.
- **Las Vegas algorithms** – correct answer, finite time, like QuickSort with random pivots
- **Monte Carlo algorithms** – possible incorrect answer, possible infinite time
 - Run till probability of incorrect < probability of outrageous other errors: RAM cosmic rays, hardware failures, etc.
 - $\text{IsPrime}(n)$ is currently deterministically $O((\log n)^{7.5})$, but k rounds of random Miller-Rabin is $O(k (\log n)^3)$
- Example: two places compute massively large result, want to prove to other. Can send $\log n$ bits to prove exponentially small error
- Examples:
 - In ECC, most random codes good
 - Crypto – random keys, nonces needed
 - Prime testing for RSA done probabilistically
 - QuickSort needs good random pivots
 - Network traffic backoff, related collision avoidance, retries
 - Evolutionary algorithms like **Neural Networks**, **Genetic Algorithm**, **Simulated Annealing**, **Stochastic Gradient Descent (SGD)**
- **Technique: randomness for performance or adversary thwarting**

Sequences

- Longest Common Subsequence
- Longest Increasing Subsequence
 - Optimal substructure and overlapping subproblems implies try **dynamic programming (DP)**
- Max Subarray Sum
 - Naïve $O(n^3)$
 - Kadane algorithm $O(n)$
 - 2D version is $O(n^3)$
- Summed range trick
- Summed area table (also called **Integral tables**)
- **Challenge (DP)**: # ways to change \$10.00 USD?

	Y		a	s	w	v
X	0	1	2	3	4	
	0	0	0	0	0	
a	1	0	↖ 1	← 1	← 1	
r	2	0	↑ 1	↑ 1	↑ 1	
s	3	0	↑ 1	↖ 2	← 2	
w	4	0	↑ 1	↑ 2	↖ 3	
q	5	0	↑ 1	↑ 2	↑ 3	
v	6	0	↑ 1	↑ 2	↑ 3	↖ 4

```
Initialize:
    max_so_far = 0
    max_ending_here = 0

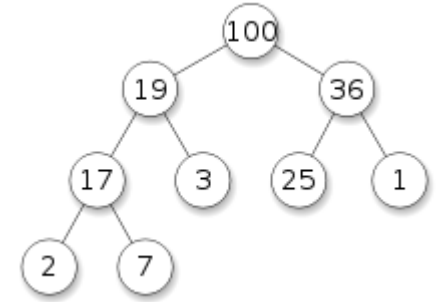
Loop for each element of the array
    (a) max_ending_here = max_ending_here + a[i]
    (b) if(max_ending_here < 0)
        max_ending_here = 0
    (c) if(max_so_far < max_ending_here)
        max_so_far = max_ending_here
return max_so_far
```

Sequence element

- Problem: Find kth largest in unordered array
- Soln 1: sort, index, $O(n \log n)$, space $O(1)$ in place, $O(n)$ otherwise
- Soln 2: heap, keep top k, walk all items, $O(n \log k)$, space $O(k)$
- Soln 3: **QuickSelect**: like QuickSort, but stop early, $O(n)$ avg, $O(n^2)$ worst
- Soln 4: **Median of Medians**: $O(n)$ always, worst space $O(\log n)$
 - Slightly tricky, but implementable
 - Time space tradeoff
 - When used in QuickSort pivot, worst case sort reduces $O(n^2) \rightarrow O(n \log n)$

Heaps

- Also called **priority queues**
- Tree: parents $\geq$ children (max-heap)
- **Binary tree** – simple code, bad merge
 - Can be stored efficiently in array
- **Binomial heap** – quick heap merges
- **Pairing heap** – simple to code
- **Fibonacci heaps** (complicated to code)
- **Brodal queue** (not good in practice)
- Many, many more

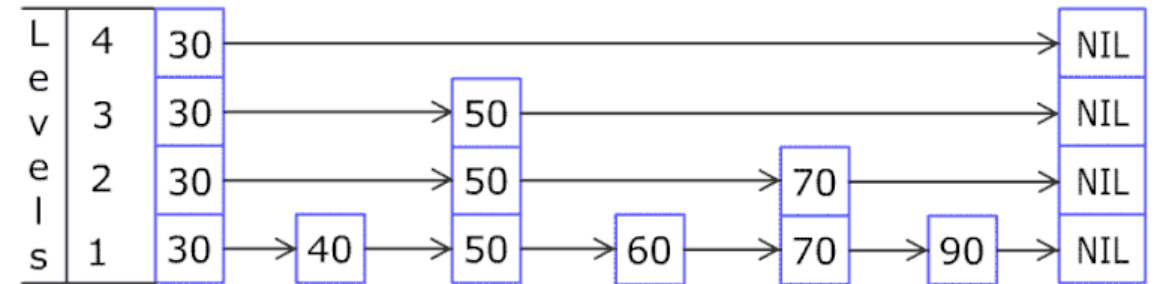


Operation	find-min	delete-min	insert	decrease-key	merge
Binary ^[8]	$\Theta(1)$	$\Theta(\log n)$	$O(\log n)$	$O(\log n)$	$\Theta(n)$
Leftist	$\Theta(1)$	$\Theta(\log n)$	$\Theta(\log n)$	$O(\log n)$	$\Theta(\log n)$
Binomial ^[8]	$\Theta(\log n)$	$\Theta(\log n)$	$\Theta(1)^{[b]}$	$\Theta(\log n)$	$O(\log n)^{[c]}$
Fibonacci ^{[8][9]}	$\Theta(1)$	$O(\log n)^{[b]}$	$\Theta(1)$	$\Theta(1)^{[b]}$	$\Theta(1)$
Pairing ^[10]	$\Theta(1)$	$O(\log n)^{[b]}$	$\Theta(1)$	$o(\log n)^{[b][d]}$	$\Theta(1)$
Brodal ^{[13][e]}	$\Theta(1)$	$O(\log n)$	$\Theta(1)$	$\Theta(1)$	$\Theta(1)$
Rank-pairing ^[15]	$\Theta(1)$	$O(\log n)^{[b]}$	$\Theta(1)$	$\Theta(1)^{[b]}$	$\Theta(1)$
Strict Fibonacci ^[16]	$\Theta(1)$	$O(\log n)$	$\Theta(1)$	$\Theta(1)$	$\Theta(1)$
2-3 heap	?	$O(\log n)^{[b]}$	$O(\log n)^{[b]}$	$\Theta(1)$	?

Skip list

- $O(\log n)$ insert, search on n ordered elements
 - Thus gives best array search speed. But exponentially faster inserts
 - Random choice of levels where to put new elements

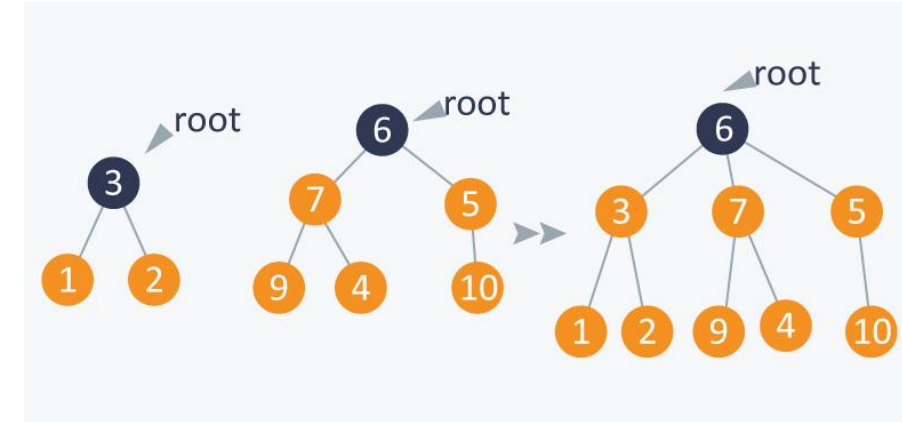
- Easy to code, good performance
 - Tricks to make cache friendly



- Not good against adversary, can be made so
- **Challenge:** implement one

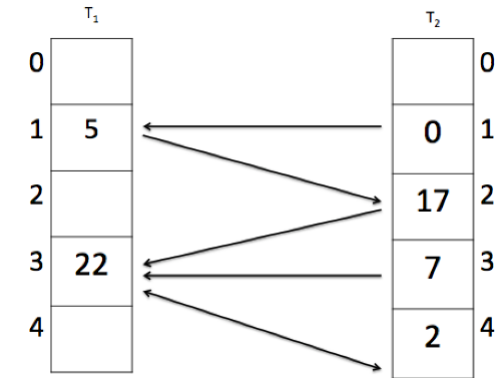
Disjoint Set

- Want to track **used** items from fixed set of items
 - Memory pages
 - FAT allocation
 - **Allocation Bitmap** nice method, uses 0/1
- More states than 0/1?
- **Disjoint-set** data structure (1964)
 - Tracks set of elements partitioned into disjoint sets
 - Near constant time add new set, merge sets, determine if elements in same set
 - Each item a node with parent pointer, can be array, also rank of item (depth)
 - Find path compression – walk pointers up over time
 - **Technique: opportunistic data structure updating**
 - Union merges smaller tree to root of larger tree

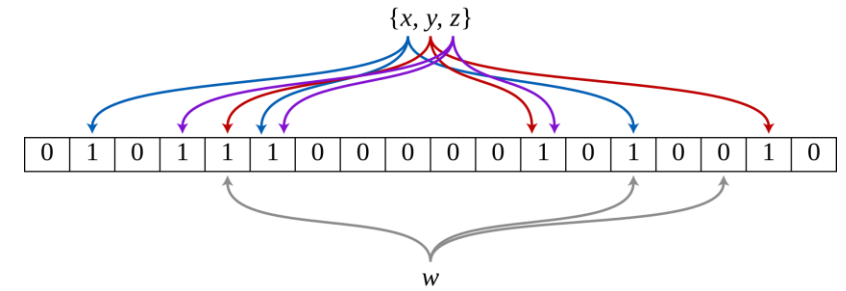


Hashing

- Hash table: array of values indexed by hashing keys.
- Hash collision: many ways to resolve
 - Linear probing – look for next free. Delete requires moving items.
- **Cuckoo hashing**: 2 hash functions, check both places first, easier to delete.
 - Collisions push items from one value to the other until stops.
Based on cuckoo pushing eggs around.
Need load factor at most 50%
- **Perfect hashing**: given fixed N items, can construct a hash that fits into N slots.
- **Hopscotch hashing**: Each bucket in a neighborhood (think cache line). If neighborhood full, resize table. Well suited for concurrent hashing.



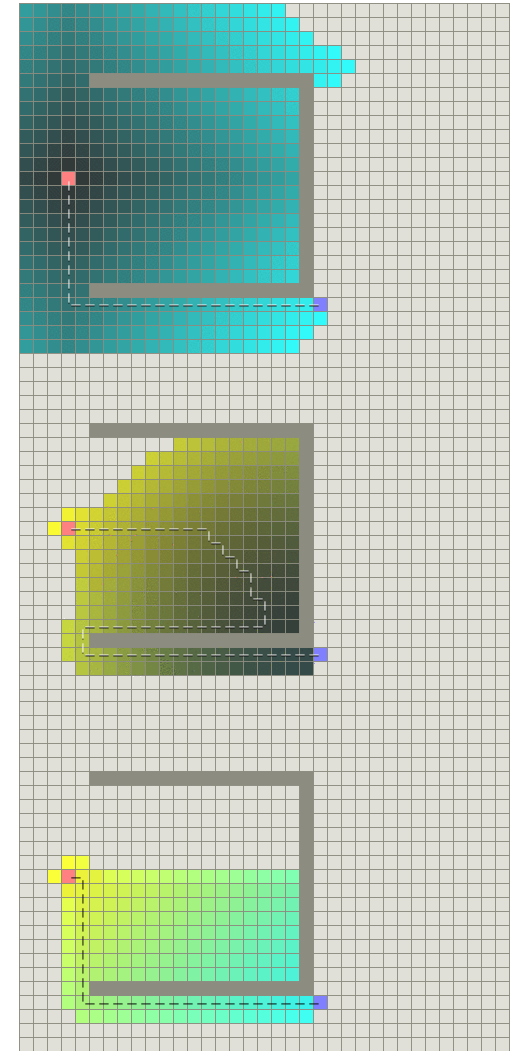
Bloom Filters



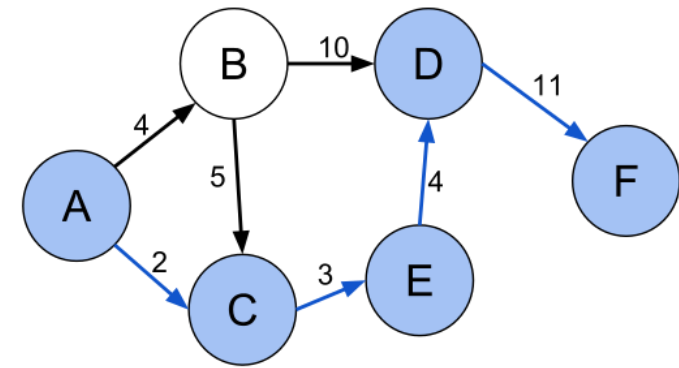
- Efficient to **test if element is *not* member of a set**.
- Set array of m bits to 0. k different hash functions, mapping each element to one of the m bits.
- False negatives not possible, false positives possible
- Vastly shrinks size of hash table
- Add: compute k hashes, set the bits
- Query: check if all k hashes bits set
- Cannot remove items (easily)
- m and k allow tuning the performance, error prob $(1 - e^{-kn/m})^k$
- Roughly, 10 bits per element gives $< 1\%$ false positive rate, no matter how many elements

Pathfinding

- Find optimal path through discrete space
- Modeled as graph search
- **A* search** (pronounced “A star”) very popular
 - Have list of paths so far
 - Use *heuristic* to find which to extend
 - Repeat
- Lots of tricks and rules to find best heuristics
- Very common in games, simple robotics, industrial optimization

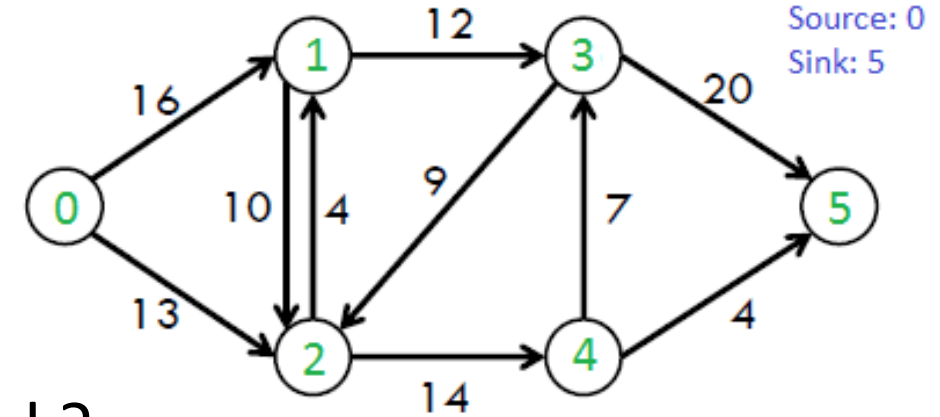


Graph



- Shortest path problem: find shortest path(s) in a graph $G = (E, V)$
- **Dijkstra's algorithm** - $O(V^2)$ – single path, non-negative edge weights
 - Trajan with Fibonacci heap 1984 $O(E + V \log V)$
 - Mark all as unvisited, each dist infinity
 - For each current in unvisited, set neighbor dists
 - Mark current as visited, add unvisited
 - Repeat till all visited
- **Bellman-Ford algorithm** if edges negative $O(V E)$, space $O(E)$
- **Floyd-Warshall algorithm** – all pairs of paths $O(V^3)$
- Lots and lots of graph algorithms

Graph



- Given graph with capacities on edges, what is max that can be pushed from source to sink?
 - Origin: 1954 analysis of Soviet railway traffic, now in all network models
- Called the **Max Flow** problem
- **Ford-Fulkerson** - $O(E \max f)$ weights f rational
- Max-flow min-cut theorem: max flow source to sink is same as the min-cut: cut minimal costs edges to disconnect source from sink
 - Useful to understand network reliability

Metrics

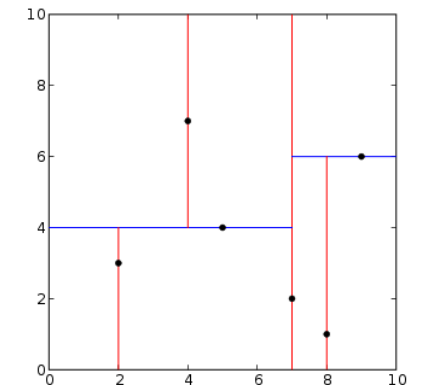
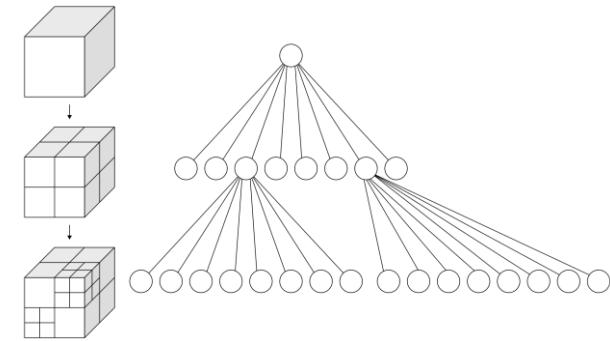
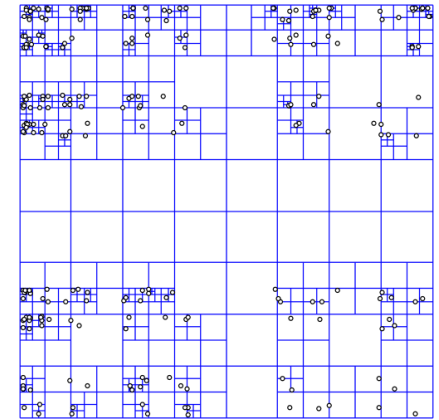
- Metric is abstract notion of distance (think Euclidean 2D or 3D space)
 - $d(x, y) \geq 0$
 - $d(x, y) = 0 \Leftrightarrow x = y$
 - $d(x, y) = d(y, x)$
 - $d(x, z) \leq d(x, y) + d(y, z)$ ***Triangle inequality***
- **Technique: leverage triangle inequality**
 - speed up all sorts of complex searching
 - Ex: explain color art speed up, photomosaic speedup
- Ex: **Hamming distance**: number of bits where strings differ
- **Sqrt trick**: long side + $\frac{1}{2}$ short side accurate to within $\sim 11.8\%$, massively faster
- Often compare distance squared instead of sqrt
- Many, many more speed tricks

Closest points

- Problem: Given n points in metric space, find closest pair of points.
- Naïve $O(n^2)$
- $O(n \log n)$ in general
 - 2D: sort on x, split in half, recurse on left and right to get 2 values, check near middle carefully
- If floor function can be computed in constant time, $O(n \log \log n)$
- If randomization + floor function, $O(n)$
 - **Technique: randomized algorithms**

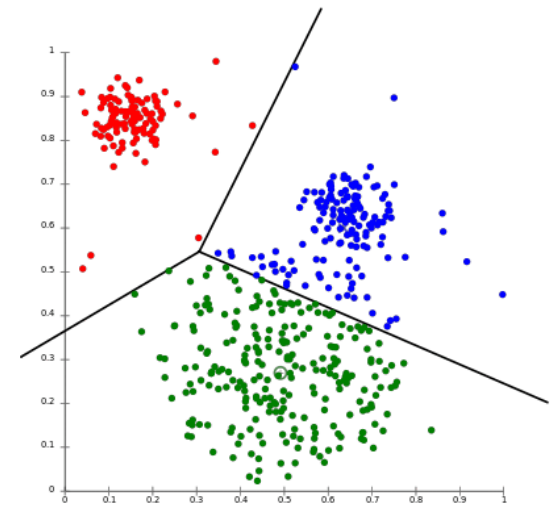
Partitioning

- For looking up things and neighbor questions in metric space
- 2D **Quadtree** – each node has 4 children
- 3D **Octree** – each node has 8 children
- **K-d tree** – each leaf is n dim point, creates splitting plane
- Most suffer from “curse of dimensionality”
 - High dimensional stuff “mostly on edges”, not “middles....”
 - Ex: n sphere mostly on edge, same as hypercube
- Many, many more (**Ball tree**, hybrid stuff....)

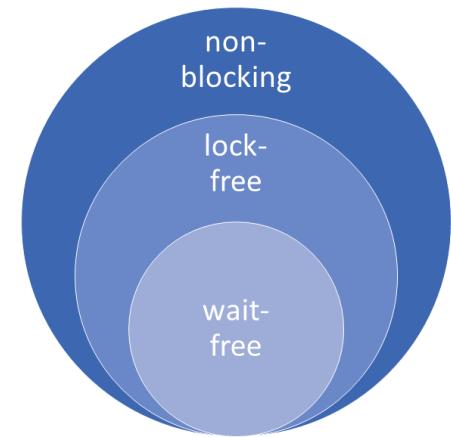


Clustering

- Want to group items by similarity under some metric
- **K-means clustering**
 - Iterative refinement (**Lloyd's Algorithm**)
 - Pick k initial items
 - Assign each to nearest of k
 - Compute new centroids
 - Repeat, stop when assignments no longer move
- Optimal solution is NP-hard



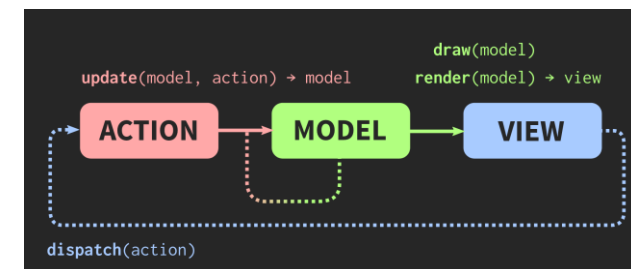
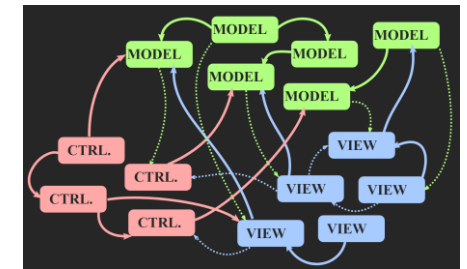
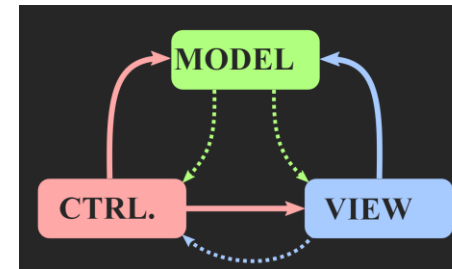
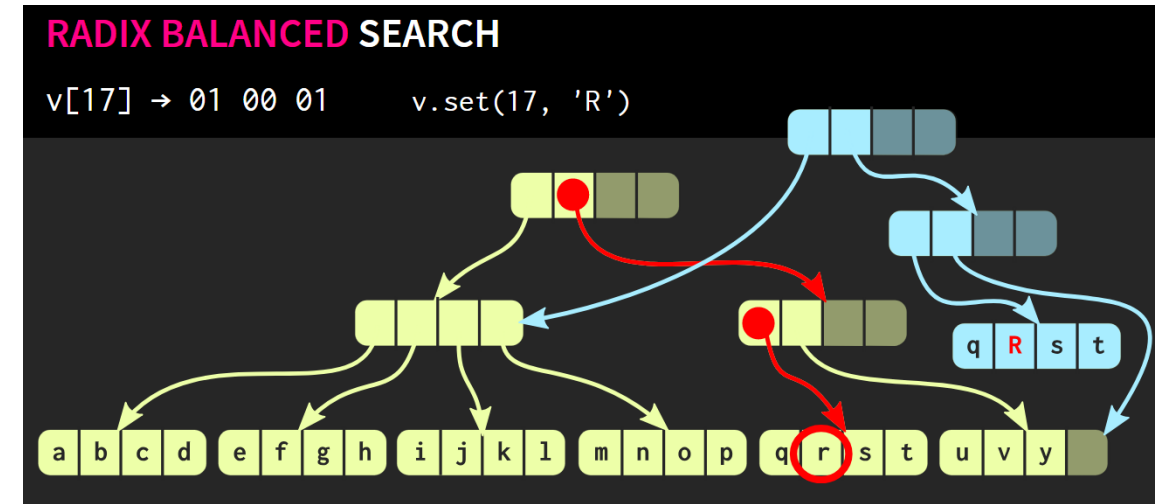
Concurrent Data Structures



- Locks: deadlock, livelock, priority inversion, problems with interrupts
 - **Coarse grained locking** loses parallelization opportunities, **fine grained** adds more locking overhead, hard to get correct.
- **Non-blocking**: a stopped thread cannot block other progress
- **Wait-free** (strongest): guaranteed systemwide progress, starvation free
- **Lock-free** (middle): guaranteed systemwide progress, individual threads can starve
- **Obstruction-free** (weakest): single thread guaranteed to make progress if all other obstructing threads suspended.
- Implemented with ***read-modify-write*** primitive. CAS most common.
- Examples:
 - **wait-free queue** (2011), **SRSW ring-buffer** with only a memory barrier, **linked lists**, **stack**, ...

Immutable data structures

- Needed as things become more parallel
- Each update returns new view of item
 - old views still valid
- Relaxed Radix Balanced Trees (**RRB Tree**)
 - Efficient immutable vector
 - Random access, update, push back, slice right/left $\sim O(1)$
 - Concat, insert, push front $O(\log n)$
- In many more modern languages (Clojure, Scala, more)
- Immer library github C++: Persistent immutable data structures
- Nicely performant
- Many, many more avenues, research, algorithms, data structures
- **Technique: immutable data structure for parallelization**



Data Compression

- Lossless

- **Run Length Encoding (RLE)**: AAAAAA => 7A
- **LZ77** – sliding window of history
- **LZ78** – explicit dictionary
- **Huffman** – optimal for power of 2 probabilities
- **Arithmetic** – optimal for fixed probabilities, is “limit” of Huffman
- **PPM** – prediction by partial matching
- **BWT** – in bzip, novel invertible “sorting” transform
- **Context-adaptive binary arithmetic coding (CABAC)** – basis of most modern

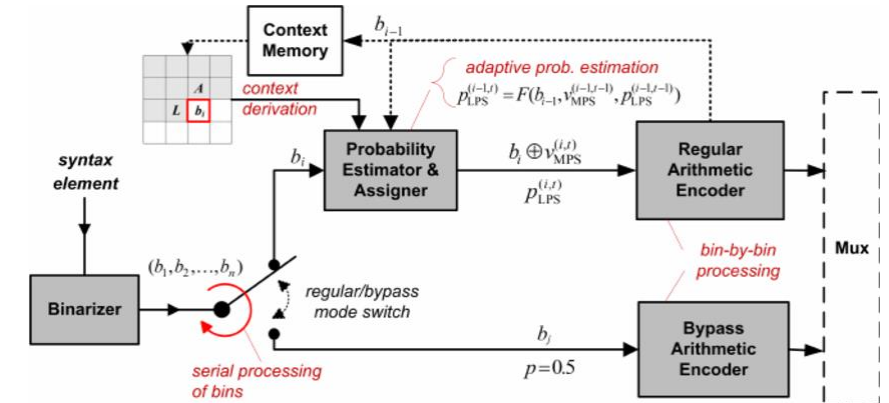
- Lossy

- Exploit perceptual bias
- Exploit temporal changes
- JPEG, MP3, H.265, Speex

- **Compressive Sensing** (Tao, Candes, .., 2004)

- Massive breakthrough : proved cases where can reconstruct signal with exponentially fewer samples than Nyquist limit
- Design of radiating systems, radar through wall imaging, antenna design, MRI

- **Challenge**: Implement a few of the first lossy versions



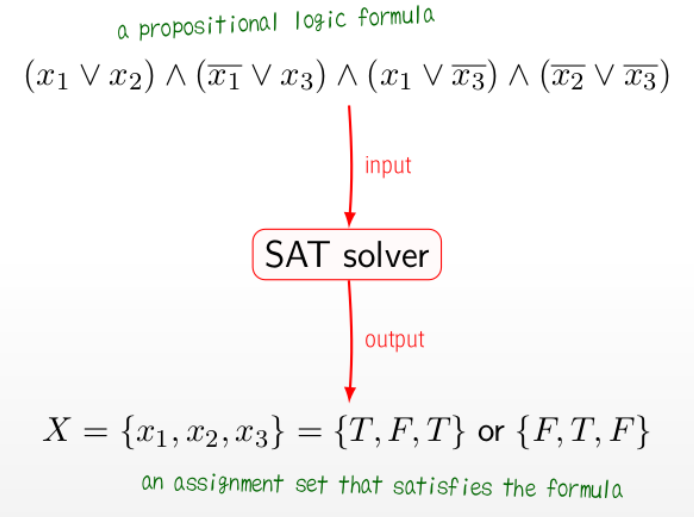
Boolean

- **SAT** – Boolean Satisfiability

- Given a Boolean expression in k variables, does there exist a True/False assignment making the expression true?
- Fundamental, many real world problems can be cast into it.
- Essential part of EDA toolbox. Even Excel has decent ones built in
- Solvers called **SAT solvers**
 - SAT solvers useful for many problems, despite NP hard
 - Can often handle millions of variables in millions of clauses
 - Knuth TAOCP chapter especially nice presentation
 - Annual competition in the 2010s drove state of the art much higher

- Boolean minimization

- **Karnaugh Map**, implemented as **Quine-McClusky** algorithm
- Faster evaluation
- Smaller circuits



AB

	00	01	11	10
00	0	0	1	1
01	0	0	1	1
11	0	0	X	1
10	0	1	1	1

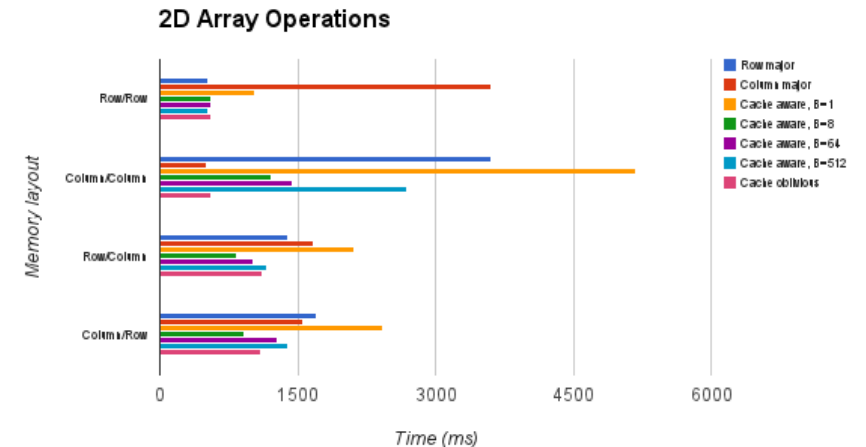
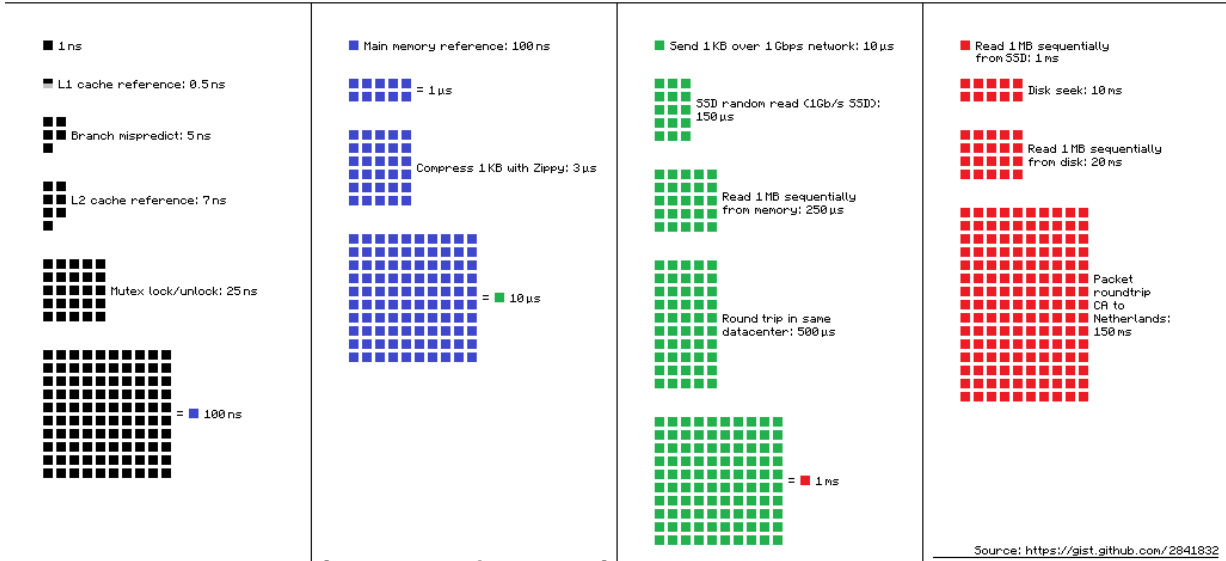
CD

$f(A,B,C,D) = E(6,8,9,10,11,12,13,14)$
 $F = A + BCD'$
 $F = (A+B)(A+C)(A+D')$

Cache aware

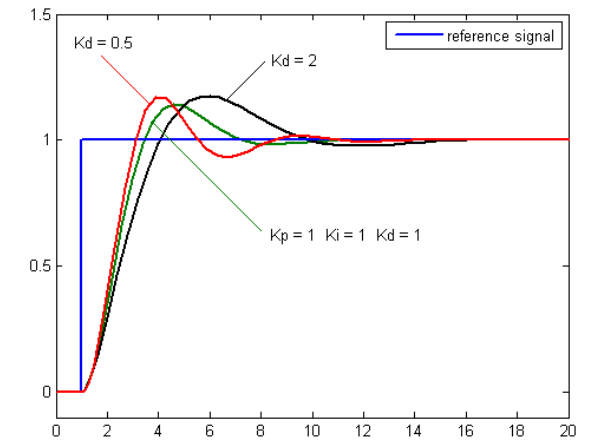
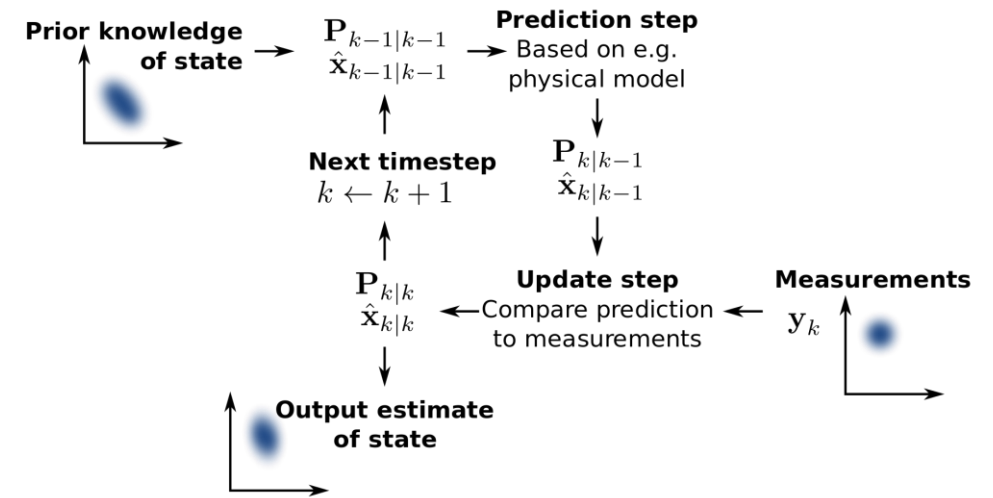
- Being cache careful can give 10-100x performance speedups
- Huge cache variety across machines
- Naïve: compile per platform, carefully tune, costly and tricky, maintenance
- Problem: Make cache aware code portable
- **Cache aware algorithms**
 - also called cache oblivious algorithms
 - H. Prokop MIT Master's Thesis (1999)
 - **Technique: break problem recursively into 2^n sizes**
 - One eventually fits in cache
- Many exist: funnel sort, quicksort, matrix transposition, array reversal, b-trees, priority queues

Latency Numbers Every Programmer Should Know



Control Theory

- Sources of information, noisy, want to merge
 - Sensor fusion (GPS, accelerometer, clock, compass, cameras,...)
 - Robots
 - Self-driving cars
 - Industrial machines
 - Phone
- **Kalman filter** – algorithm to do sensor fusion with uncertainty
 - **Extended Kalman Filter** (EKF)
- **PID controller**
 - proportional–integral–derivative controller
 - Drones
 - Engines
 - Industrial control
- **Challenge**: make simple PID that tracks some value, then poke at the value and see how the PID responds



```
previous_error = 0
integral = 0
loop:
    error = setpoint - measured_value
    integral = integral + error * dt
    derivative = (error - previous_error) / dt
    output = Kp * error + Ki * integral + Kd * derivative
    previous_error = error
    wait(dt)
    goto loop
```

The End

Questions?

Change Problem

- How many ways to make change for \$10.00?
- Coins & bills = {1,5,10,25,50,100,200,500,1000} = 9 items, S_0 to S_8
- $C(m, n)$ = number of ways to total n with items S_0 through S_m
- Key insight: split computation to those using S_m and those not
 - Technique: **split problem into mutually exclusive cases**
- $C(m, n) = C(m, n - S_m) + C(m - 1, n)$
- End cases:
 - if $n < 0$ or $m < 0$ then $C=0$
 - if $n == 0$ then $C = 1$
- Dynamic programming or memorization: $C(8, 1000) = 3,237,135$

TODO

- **Highlight general techniques**
- Final slide with techniques for review
- On final slide, list techniques one side, algo names other
- **Techniques: meet in middle (sqrt trick)**