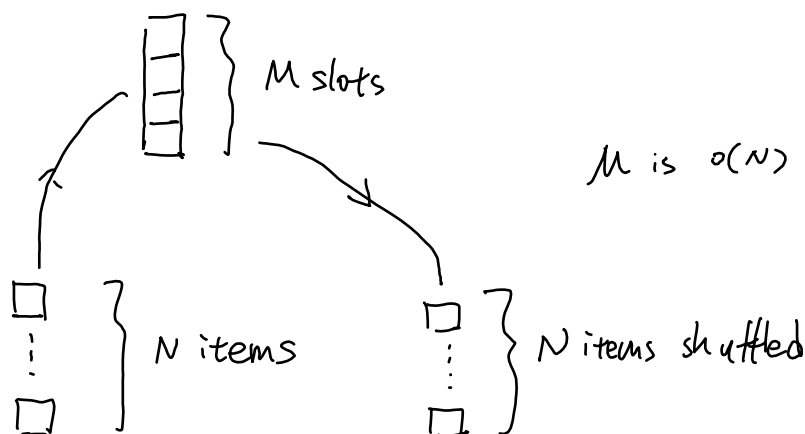


# Stream Shuffle Problem

How much entropy can we get on a data stream of length  $N$ , with  $M$  slots of random access memory?



A lossy upper bound is  $\sum_{j=1}^N \ln j \sim O(N \ln N)$

↑ uniform dist. for  $N!$  items

What kind of entropy are we looking at?

We have a random process  $X_t(\omega) : \{0, 1\}^{[1, \dots, t]}$  and  $\sum_{i=1}^t (X_t(\omega))_i \leq M$

$$X_{t+1}(\omega) \rightsquigarrow X_t(\omega)$$

↖ This works like removing one element and add a new element.

Formally, we have this statement:  $\underbrace{d_H(X_{t+1}(\omega), X_t(\omega))}_\uparrow = 2 \ \& \ (X_{t+1}(\omega))_{t+1} = 1$   
 Hamming Distance

First, we prove that if  $X_t(\omega)$  is initially filled (which is  $\sum_{i=1}^t (X_t(\omega))_i = M$ ) then the entropy of this process is the same as the shuffled stream, i.e. the selected item.

$X_0, X_1, \dots, X_t$

The probability mass is  $p_{X_{t+1}|X_t}(X_{t+1}|X_t)$



which is the probability to remove an element.

(invalid moves are just zero)

How to enumerate the elements in this space?

Suppose  $\Omega_{m,n}$  represents the space with  $m$  memory slots and  $n$  stream items.

Then the size of probability space  $|\Omega_{m,n}|$  will be

- case 1:  $m \geq n$

$$|\Omega_{m,n}| = n!$$

- Case 2:  $m < n$



$$|\Omega_{m,n}| = m! \cdot \underbrace{m^{(n-m)}}_{\text{for } [n-m+1 \dots n], \text{ they have } m \text{ choices}}$$

for  $[n-m+1 \dots n]$ , they have  $m$  choices

we randomly remove items from memory slots, so there are  $m!$  items to be removed

Question: Is the entropy of the shuffle the entropy of the distribution on  $\Omega_{m,n}$ ? (Not exactly)

$$H(p_{x,y}) = - \sum_{x,y} p(x,y) \log p(x,y)$$

$$= - \sum_{x,y} p(y|x) \cdot p(x) \cdot (\log p(y|x) + \log p(x))$$

$$= - \sum_{x,y} p(y|x) \cdot p(x) \cdot \log p(y|x) - \sum_{x,y} p(y|x) \cdot p(x) \cdot \log p(x)$$

$$= - \sum_x p(x) \cdot H(p_{y|x}(\cdot|x)) - \sum_x p(x) \log(p(x))$$

↖ It seems like we will use this twice.

First, let  $X$  be the choices (evictions) made in the first  $(n-m)$  outputs.

Let  $Y$  be the evictions made in the last  $m$  outputs.

According to the formula, no matter what  $X$  is instantiated after the  $(n-m)$  steps,

We just maximize  $P_{Y|X}(\cdot | X)$ , and it is simple: just uniformly pick items from all non-empty slots.

For now, let's just worry about the replacement part (that  $(n-m)$  thing)

We notice that the picking of slots can be independent, and when it is uniform it maximizes the entropy.

Also, there is a one-one correspondence between

Question: If we are not doing shuffle, but instead we want to perform sampling

# Grad Algorithm Lecture 2

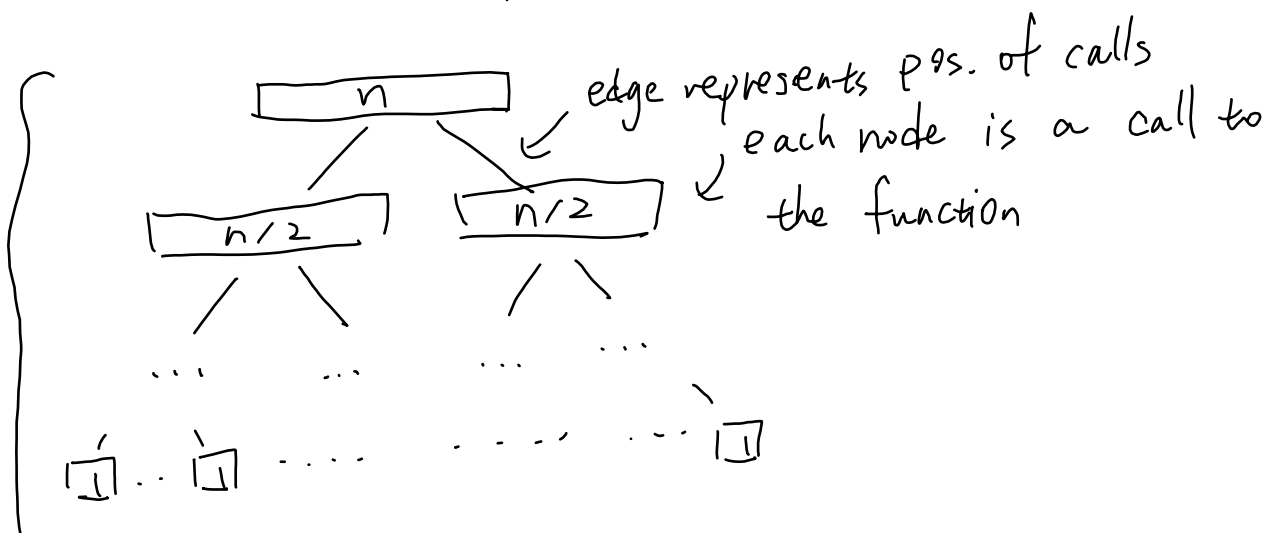
Problem Set : piazza.com ↙ whitelist that as (non-span)

Discuss Solutions ☒ Syllabus  
↑  
Home Page

Review : Merge Sort

$$T(n) = \begin{cases} 2 \cdot T(\frac{n}{2}) + O(n) & n > 1 \\ O(1) & n = 1 \end{cases}$$

Recursion Tree :



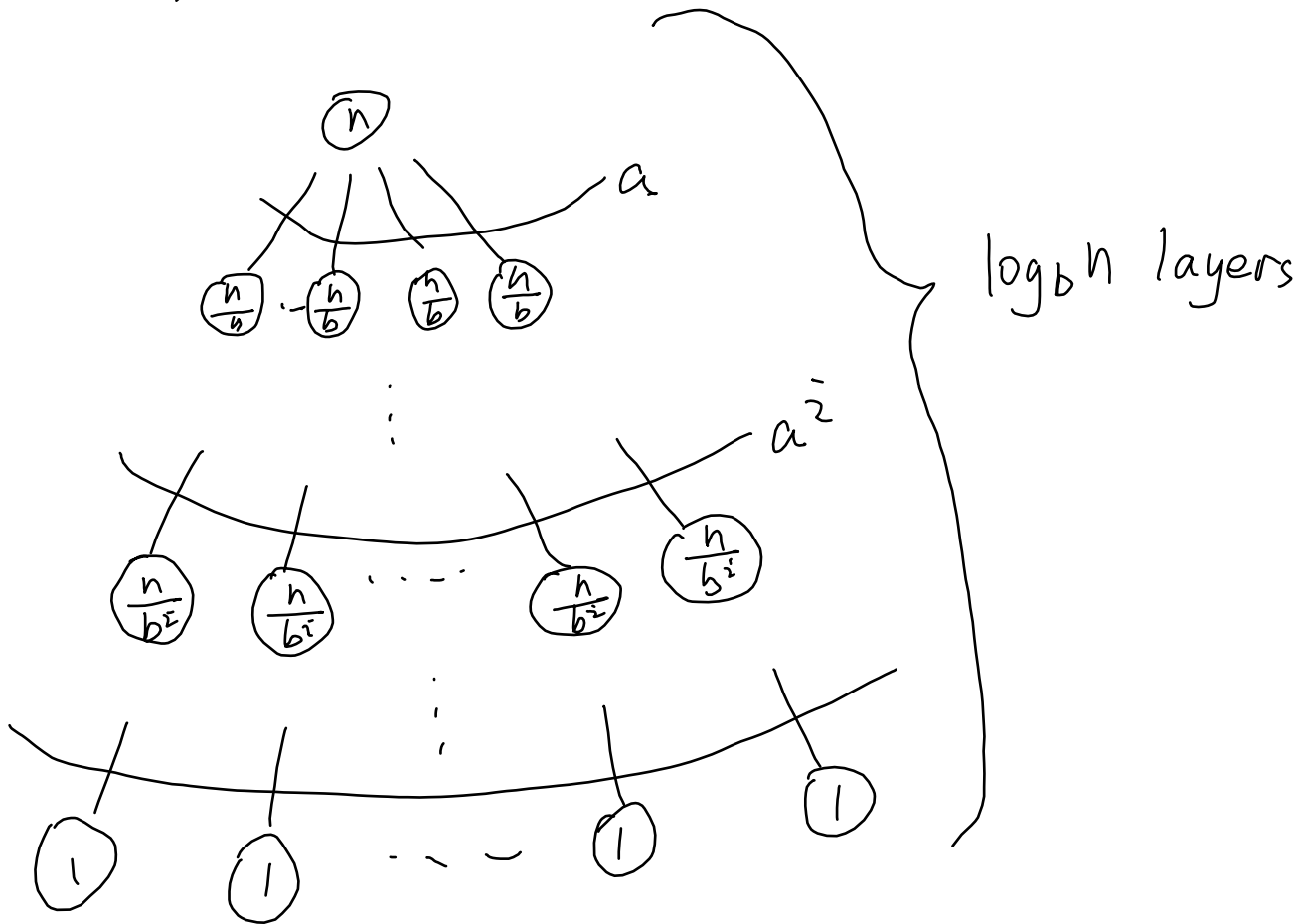
↙  $\lg n$   
height

$$O(n) + 2O(n/2) + 4O(n/4) + \dots + n O(1)$$

$\lg n$  terms

Generalize a little!

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + O(n^c)$$



$$\begin{aligned}
 T(n) &= \sum_{i=0}^{\lfloor \log_b n \rfloor} a^i \cdot O\left(\frac{n^c}{b^{ic}}\right) \\
 &= \sum_{i=0}^{\lfloor \log_b n \rfloor} \frac{a^i}{b^{ic}} \cdot O(n^c) \quad \text{not if } a = b^c \\
 &= \frac{1 - \left(\frac{a}{b^c}\right)^{\lfloor \log_b n \rfloor}}{1 - \frac{a}{b^c}} \cdot O(n^c) = \begin{cases} O(n^{\log_b a}) & a > b^c \\ \log n \cdot O(n^c) & a = b^c \\ O(n^c) & a < b^c \end{cases}
 \end{aligned}$$

Another D&C

$$A = [3 \ 1 \ 2 \ 8 \ 6 \ 4 \ 5]$$

Find maximum  $A[i] - A[j]$  where  $i < j$

# Polynomial Multiplication

$$f(x) = \sum_{i=0}^d x^i f_i$$

$$f(x) \cdot g(x) = \sum_{i=0}^{d_f+d_g} x^i \left( \sum_{j=0}^i f_j g_{i-j} \right)$$

Naive implementation gives  $O(n^2)$  ← where  $n = \text{degree}$

Polynomial Multiplication is convolution

---

Applications & Examples :

3-SUM :

Input : 3 sets  $A, B, C \subseteq \{1, 2, \dots, n\}$

Is there  $a \in A, b \in B$  s.t.  $a+b \in C$

$$f_A(x) = \sum_{a \in A} x^a \quad f_B(x) = \sum_{b \in B} x^b$$

$$f_C(x) = f_A(x) \cdot f_B(x)$$

Conjecture : For  $|A| \leq n$  w/o upper bound of values

$$|B| \leq n$$

$$|C| \leq n$$

Algo returns true if there is  $c \in C$  with

$$\sum_{k=0}^c f_{c-k} g_k \neq 0 \quad (\Rightarrow) \quad \exists k : k \in B \wedge c-k \in A \wedge c \in C$$

## Image detection & Pattern Matching

Task: Picture A  $\leftarrow$  larger

Picture B  $\leftarrow$  smaller

Find B in A (simplified, 1-d)



$$f_A(x) = \sum_{i=0}^n x^i A[i]$$

$$f_B(x) = \sum_{i=0}^m x^i B[i]$$

---

## FFT Algorithm

Representation of Polynomials P

1) coefficient

2) point representation (degree d), we have  $d+1$  points

$\{x_1, \dots, x_i, \dots, x_{d+1}\}$ , and  $\{y_i : y_i = p(x_i)\}$  evaluated at  $x_i$

If  $f$  and  $g$  are two polynomials in point representation,  
(evaluated on the same set of points)

compute  $h(x) = f(x) \cdot g(x)$  will be just  $h(x_i) = f(x_i) \cdot g(x_i)$



Remember, that  $h$  has degree  $2d$

Evaluate one of the parts on  $f(x)$  always takes  $O(n)$ .

But we can share some computation between different points,

# Ford-Fulkerson Method

Idea 1: start with some flow  $f$

Gradually improve  $f$  by finding an augmenting path in  
a modified graph  $G^f \leftarrow$  residual graph

Idea 2: we can increase  $f_e$  up to capacity  $c_e$   
decrease  $f_e > 0$

Residual Graph  $G^f$ :

- 1) same vertices  $V$
- 2) Edges  $e \in E$  have capacity  $c_e - f_e$
- 3) Edges  $-e \in E$  for  $e \in E$  with capacity  $f_e$

Alg:

- 1) Start with  $f = (0, \dots, 0)$
- 2) Create  $G^f$
- 3) Find  $s$ - $t$  path  $P$  in  $G^f$ , break if no such path exists
- 4)  $f \leftarrow f + P$ , go back to step 2

Time:

$$O((m+n) \cdot \text{"max-flow"})$$

Thm: This Alg returns maximum flow.

Def: min s-t cut of a graph

$G = (V, E)$  edge capacity  $c_e$ , source  $s$ , sink  $t$ ,  $s \neq t$ .

find  $X \ni s \wedge X \not\ni t$ , s.t.  $\sum_{e \in X \times V \setminus X} c_e$

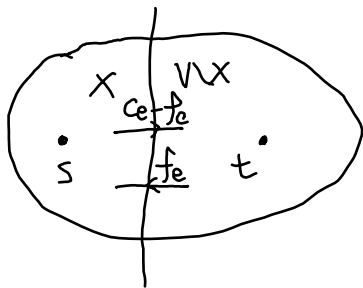
Obs: max s-t flow  $\leq$  min s-t cut

Thm: max s-t flow = min s-t cut

Pf: If  $G^f$  has no s-t path then  $\exists$  a s-t cut  $X$

s.t.  $\sum_{e \in (X, V \setminus X)} c_e = \sum_{(s, v) \in E} f_{sv}$

Let  $X$  denote all vertices reachable from  $s$  in  $G^f$

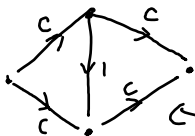


①  $\forall e \in (X, V \setminus X) \quad f_e = c_e$   $\leftarrow$  the edge weight  $c_e - f_e$

②  $\forall e \in (V \setminus X, X) \quad f_e = 0$

Alg: a polynomial algorithm finds s-t flow

in time  $O((m+n)m \log C)$   $C = \text{maximum capacity of edges}$



send maximum possible flow doesn't work

if the path is bad  $\nearrow \searrow \nearrow \searrow \dots$

Modification: To find the augmenting path, don't try an arbitrary path. Find an augment path with capacity over  $M$ .

Run the algorithm for  $M, M/2, M/4, \dots, 1$

After iteration for  $M$ , the difference between current flow and optimal flow is upper-bounded by  $M$ .

# Linear Programming

Continuous variables  $x_1, \dots, x_n$

$$\max / \min \quad \sum_{i=1}^m c_i x_i$$

$$\text{s.t.} \quad \sum_{i=1}^m a_{ij}^{(j)} x_i \leq b^{(j)} \quad \text{for } j \in \{1, \dots, m\}$$

Theorem: LP can be solved in time that is polynomial of bits of inputs.

Test Optimality of  $x$ : show a linear combination of constraints such that its coefficient is  $(c_1, \dots, c_m)$

# Homework 1

1.1  $n^c = O(d^n)$

case 1

$d > 1$  then  $\forall c > 0: \exists a : n \rightarrow \infty : n^c \leq a d^n$

case 2

$d = 1$  then we need  $c = 0$  for  $n^c$  to be constant

case 3

$d < 1$  then  $n \rightarrow \infty: d^n \rightarrow 0$ : no  $c > 0$  feasible

1.2  $\exists a > 1: n \rightarrow \infty : (\log n)^c \leq a^c n^d$

$c > 0$ , so  $(\cdot)^c$  is mono-incr,  $\log n \leq a n^{d/c}$

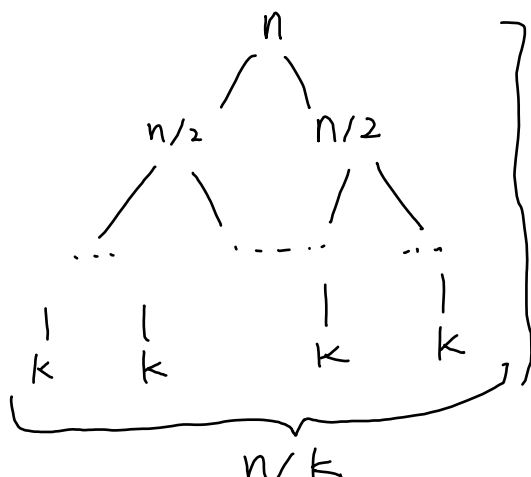
This requires  $d/c > 0$ , which always holds.

2.1  $T(n) = 4 \cdot T(n/2) + n^2$

Apply master theorem,  $4 = 2^2$ , so  $T(n) = n^2 \log n$

2.2  $T(n, k) = 2 T(n/2, k) + k$      $T(k, k) = O(k \log k)$

We should suppose  $n \geq k$  and  $T(n, k)$  increases when  $n \uparrow, k \uparrow$



$$T(n, k) = O((n/k) \cdot k \log k) \\ = O(n \cdot \log k)$$

similar to master theorem, the  
dominate term is  $O((n/k)^{\log_2 2})$

$$2.3 \quad T(n) = T(n/2) + T(n/3) + n$$

$$T(n) + an = T(n/2) + a(n/2) + T(n/3) + a(n/3)$$

so we have  $a = -6$

$$F(n) = T(n) - 6n$$

$$F(n) = F(n/2) + F(n/3)$$

# 1. Prove Cauchy-Binet Formula

Attempt 1:

$$\det(A \times B) = (\det A)(\det B) \quad \leftarrow \text{when } A, B \text{ are both } n \times n$$

which can be written as:

$$\begin{matrix} n \\ \{ \end{matrix} \left| \begin{matrix} \boxed{A} & \boxed{\text{shaded}} \\ \hline 0 & \boxed{B} \end{matrix} \right| \begin{matrix} \sigma(i) \\ \hline \end{matrix} = \prod_{\sigma} \text{sgn}(\sigma) \cdots a_{i\sigma(i)} \cdots b_{i-n\sigma(i)-n} \cdots$$

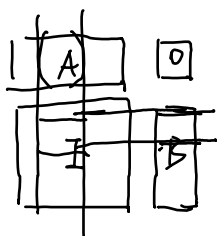
$\downarrow$

For a term to be non-zero, all elements must be selected from  $A, B, \boxed{\text{shaded}}$ . Selecting a term  $(i, \sigma(i))$  in  $\boxed{\text{shaded}}$  means  $\forall j: a_{j\sigma(j)}$  cannot be selected, but each col  $j \in [n]$  needs to select a  $\sigma(j)$ , so  $\sigma(j) \in [n+1..2n]$ , which selects a zero term.

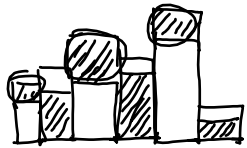
$$\text{So } \det \begin{pmatrix} A & \boxed{\text{shaded}} \\ 0 & B \end{pmatrix} = \det \begin{pmatrix} A & 0 \\ 0 & B \end{pmatrix} = \det(A) \det(B)$$

$$\begin{aligned} \det \begin{pmatrix} A & I \\ 0 & B \end{pmatrix} &= \det \begin{pmatrix} A & I \\ -BA & 0 \end{pmatrix} = \det \begin{pmatrix} -BA & 0 \\ A & I \end{pmatrix} \\ &= \det(BA) \end{aligned}$$

If it works for Cauchy-Binet formula, that will be wonderful!



$$d_{TV}(\mu, \nu) = \sup_A |\mu(A) - \nu(A)|$$



compute the max difference for each side

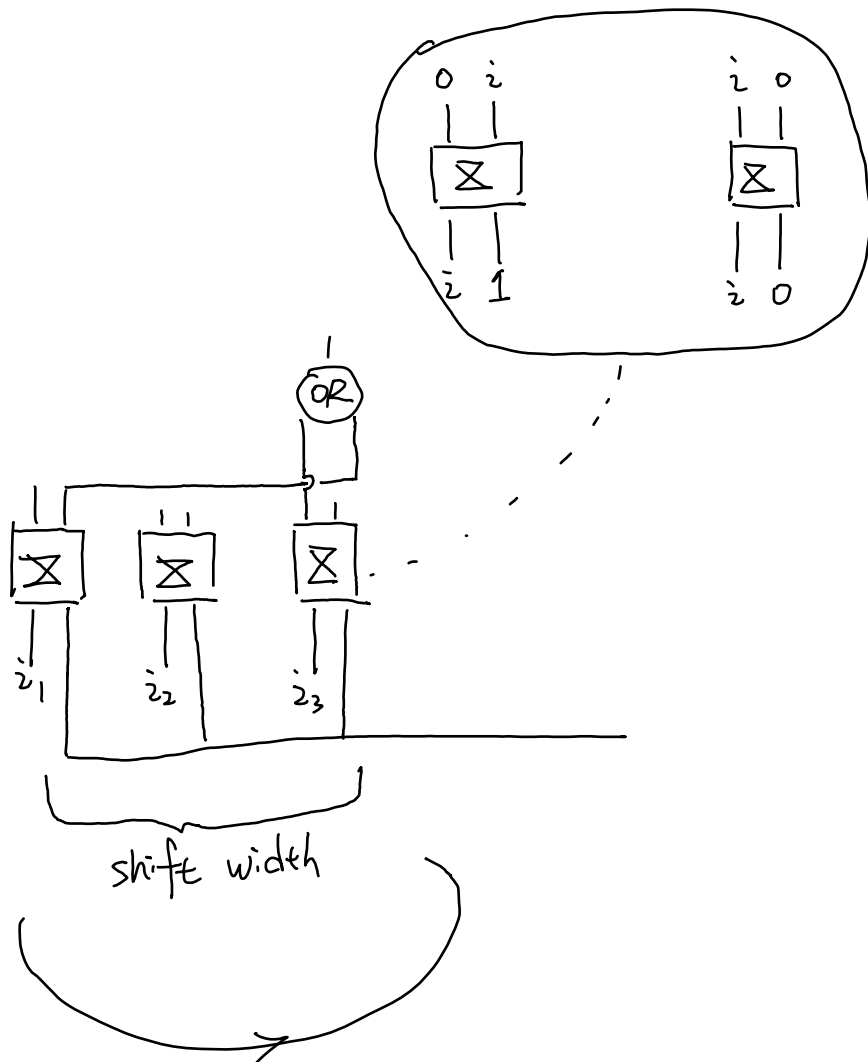
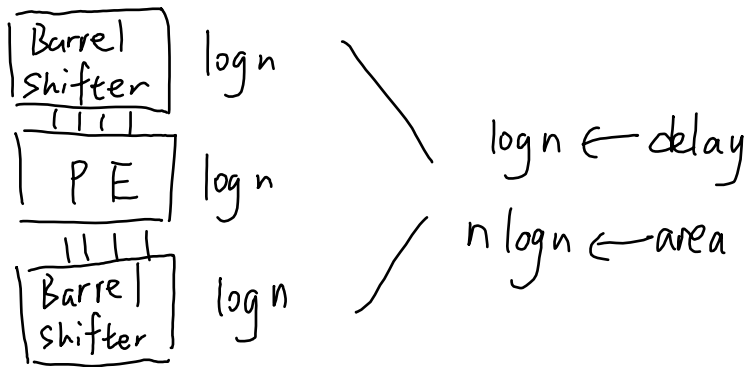
$$T_x(\epsilon) = \min \{t : d_{TV}(P^t(x, \cdot), \pi) \leq \epsilon\}$$

$$\begin{aligned} \forall x : T_x(\epsilon) &= \min \{t : d_{TV}(P^t(x, \cdot), \pi) \leq \epsilon\} \\ &\leq T_{\pi}(1/4) \log_2(1/\epsilon) \end{aligned}$$

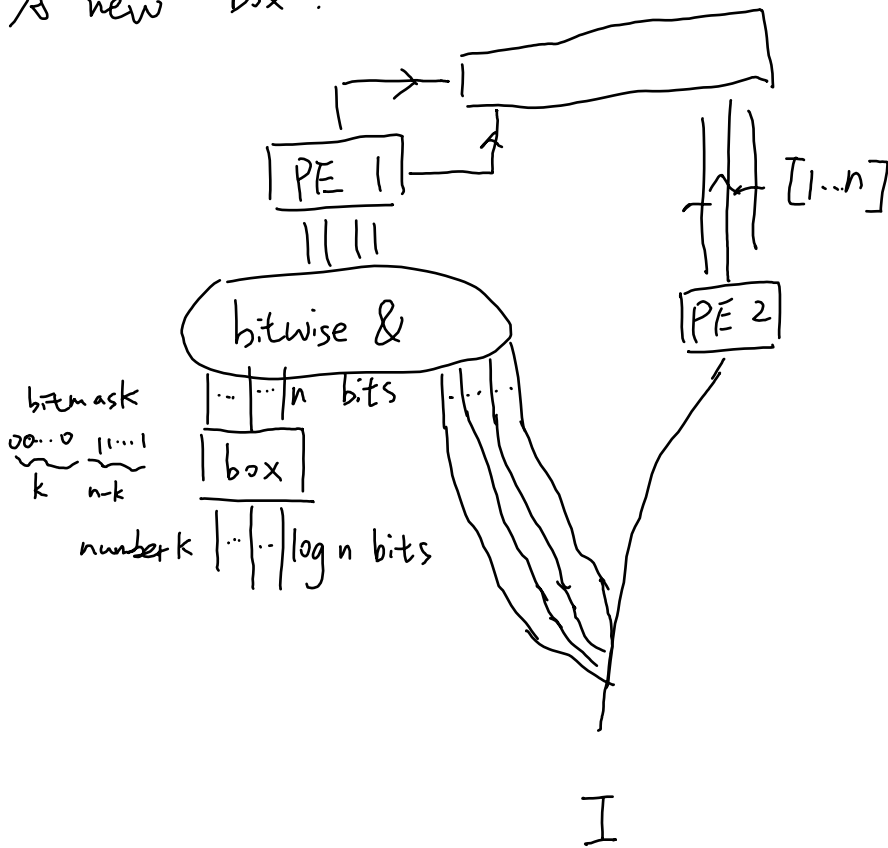




# Programmable Priority Encoder



A new box :



I

If  $a[k..n]$  has a one, then take  $a[k..n]$   
 otherwise take  $a[0..n]$ .

Register 1 cycle  
 SRAM 10ns  
 DRAM 40~60ns page mode 4kB  
 interleaved DRAM

Pipeline Design :

Other techniques

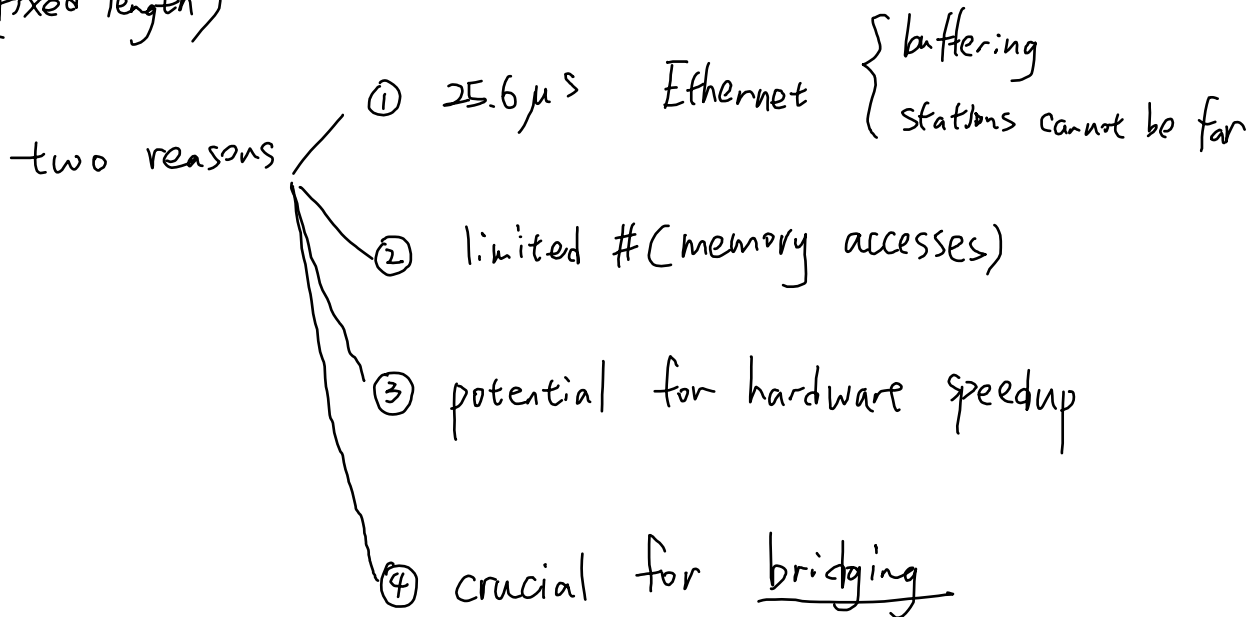
combine DRAM w. SRAM

wide word parallelism

---

Exact-match Lookups

(Fixed length)



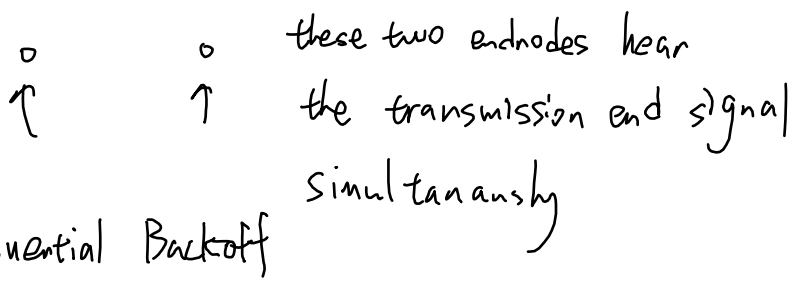
CSMA/CD



RJ-45 P2P

← only one endnode transmits.

Consider this scenario :



# Ethernet Bridge



Time is sliced into frame in Ethernet

$$(1 + o(1)) \frac{\log n}{\log \log n} \leftarrow \text{max collision in hashmap}$$

## Perfect hashing

# of elements in the bucket

$$\text{Binomial}(n, \frac{1}{n})$$

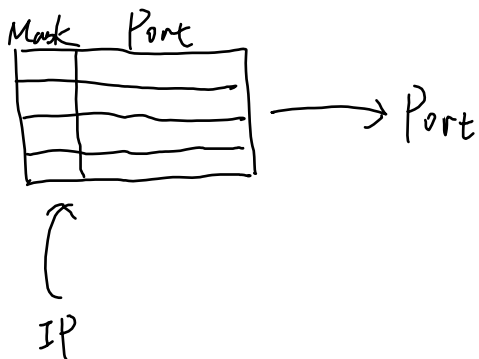
when  $n \rightarrow \infty$

Poisson Distribution with avg 1.

d-left (Another kind of hash table)

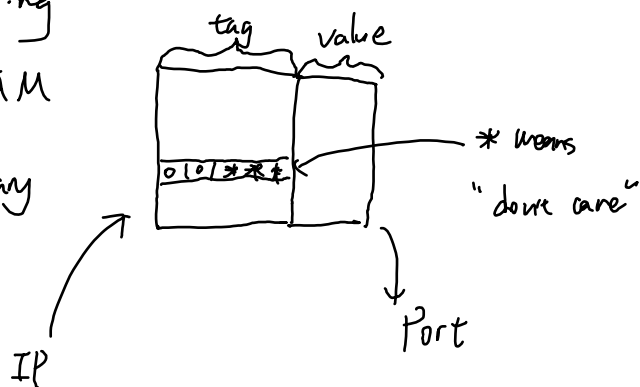
$$\log \log n / \log 2 + o(1)$$

# Longest Prefix Match



## "Engineer Solutions"

1. Caching
2. TCAM  
Ternary



## trie-based algorithm

1. binary-encoded edges (unibit trie)
2. when it matches, walks down the edge

## optimization:

1. multi-bit edges (typically 8 bits)
2. duplicate values over matching entries

if multiple prefixes end at the same level, then the port for longer prefix is stored,

fixed stride optimization: determine optimal step length with  $k$  strides

$$dp[l, k] = \min_{1 \leq j \leq 32-l} \{ 2^j \cdot |\text{forest}(l)| + dp[l+j, k-1] \}$$

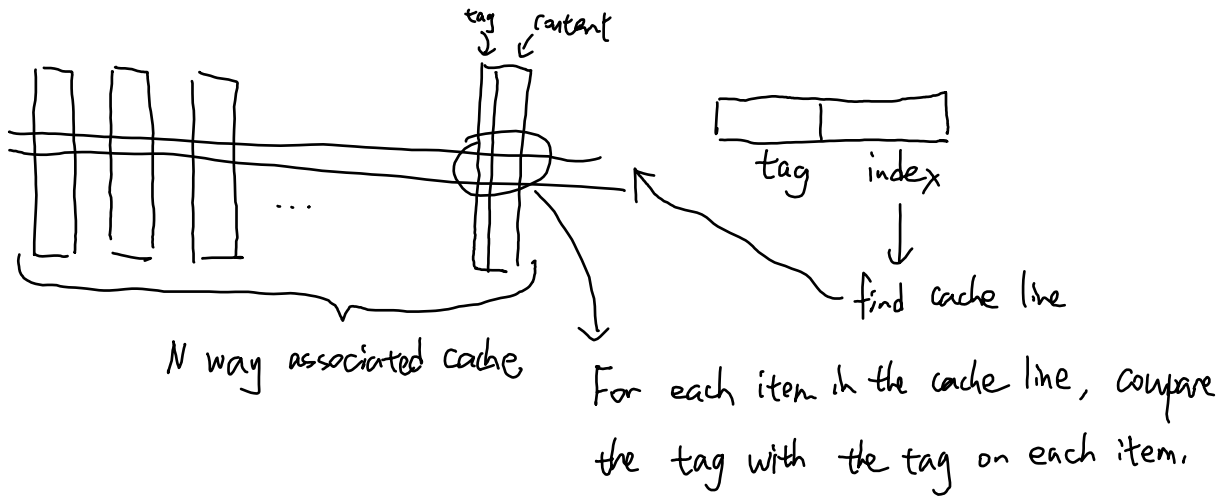
$$dp[l, 1] = 2^{32-l} \cdot |\text{forest}(l)|$$

↖ the number of prefixes

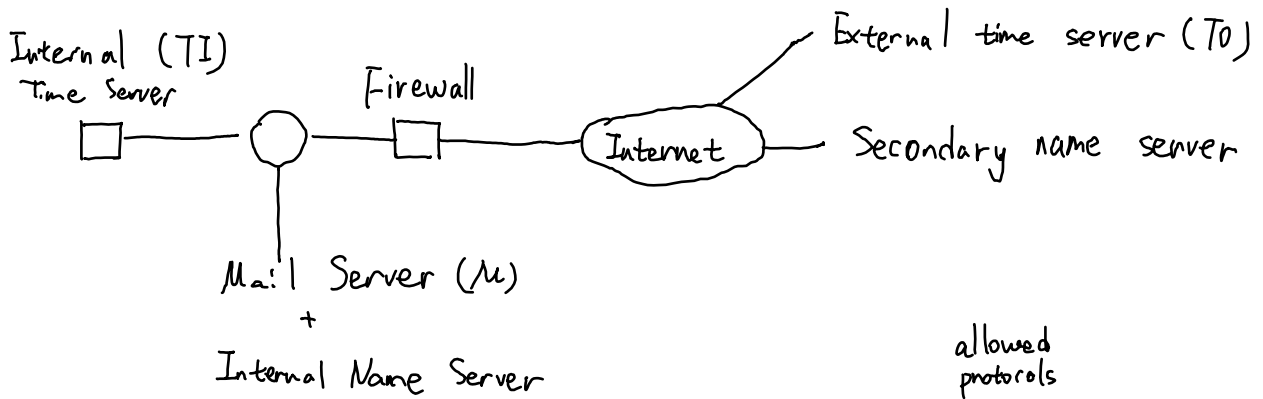
Variable stride optimization: determine optimal step length

$$dp[r, k] = \min_{1 \leq j \leq 32 - \text{level}(r)} \{ 2^j + \sum_{t \in \text{forest}(r, j)} dp[t, k-1] \}$$

## Set Associated Cache



## Firewall in 80s



D: Destination

S: Source

P: Port

A: Action

DIP	SIP	DP	SP	Flags	ACT
M	*	25	*	*	allow incoming mail
				⋮	

EGT: extended grid of tries

## Decision Tree for Firewall