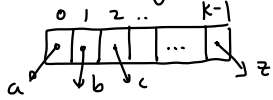


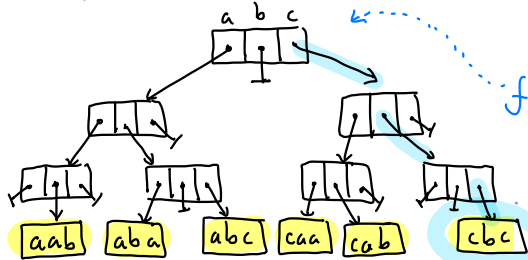
Tries: History

- de la Briandais (1959)
- Fredkin - "trie" from "retrieval"
- Pronounced like "try"

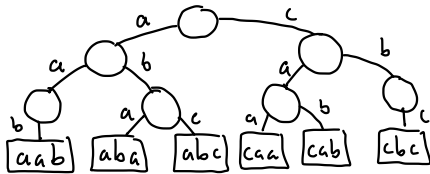
Node: Multiway of order k



Example: $\Sigma = \{a=0, b=1, c=2\}$
 Keys: $\{aab, aba, abc, caa, cab, cbc\}$



Same structure/Alt. Drawing



Digital Search:

- Keys are strings over some alphabet Σ
- E.g. $\Sigma = \{a, b, c, \dots\}$
 $\Sigma = \{0, 1\}$ Let $k = |\Sigma|$
- Assume chars coded as ints: $a=0, b=1, \dots, z=k-1$

Tries and Digital Search Trees I

find("cbc")

Analysis:

Search: \sim length of query string $[O(1)$ time per node]

Space:

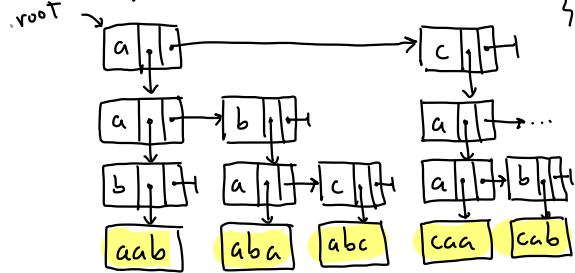
- No. of nodes \sim total no. of chars in all strings
- Space $\sim k \cdot (\text{no. of nodes})$

Large!

Analysis:

- Space: Smaller by factor k
- Search Time: Larger by factor of k

Example:



How to save space?

de la Briandais trees:

- Store 1 char. per node
- $\boxed{x} \rightarrow \neq x \Rightarrow$ try next char in Σ
 $= x \Rightarrow$ advance to next character of search string
- First-child/next-sibling

Patricia Tries:

- Improves trie by compressing degenerate paths
- PATRICIA = Practical Alg. to Retrieve Info. Coded in Alpha...
- Late 1960's: Morrison + Guchenberger
- Each node has **index field**, indicates which char to check next (Increase with depth)



Dealing with long Paths:

- To get both good spaces + query time efficiency, need to avoid long, degenerate paths.



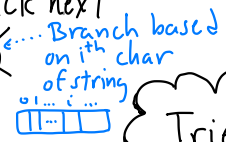
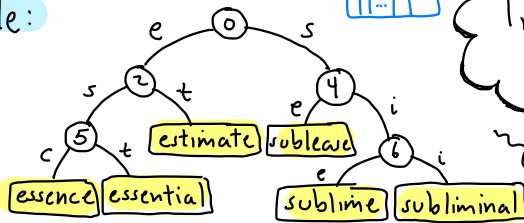
Example:

ID	String	Prefix	Identifier
S_5	ajam...	aj	aj
S_{10}	\$		
S_4	pajam...	paj	paj
S_9	a#	a#	a#
S_3	apaja...	ap	ap
S_8	ma#	ma#	ma#
S_2	mapaj...	map	map
S_7	ama#	ama#	ama#
S_1	amapaj...	amap	amap
S_6	jama#	j	j
S_0	pamapa...	pam	pam

Path compression!

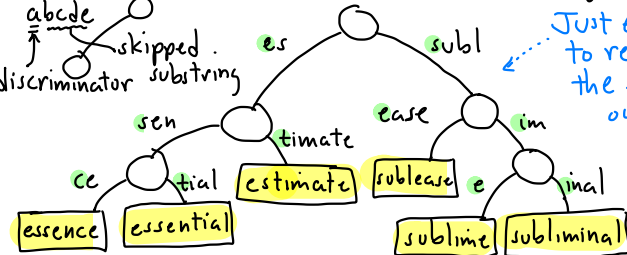
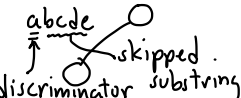
Example:

- essence
- essential
- estimate
- sublease
- sublime
- subliminal



Tries and Digital Search Trees II

Same data structure - Drawn differently



Just easier to read the strings out... same data struct.

Analysis:

- **Query time:** (Same as std trie) \sim search string length (may be less)
- **Space:**
 - No. nodes:** \sim No. of strings (irresp. of length)
 - Total space:** $K \cdot$ (No. of nodes) + (Storage for strings)

Example: $S = \text{pamapajama}\#$

- $S_{10} = \#$
- $S_9 = a\#$
- $S_8 = ma\#$
- $S_7 = ama\#$

Def: Substring identifier for

- S_i is shortest prefix of
- S_i unique to this string
- Eg. $ID(S_1) = \text{"amap"}$
- $ID(S_7) = \text{"ama\#"}$

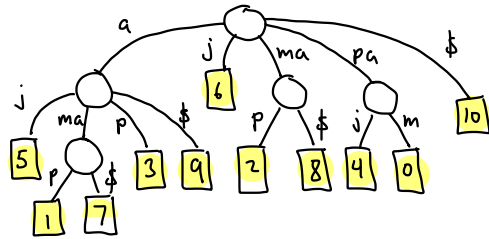
Suffix Trees:

- Given single large **text** S
- Substring queries: "How many occurrences of "tree" in CMSC 420 notes"

Notation: $S = a_0 a_1 a_2 \dots a_{n-1} \#$

- **Suffix:** $S_i = a_i a_{i+1} \dots a_{n-1} \#$ (special terminal)
- **Q:** What is minimum substring needed to identify suffix S_i ?

Example: $S = \overset{0}{p}\overset{1}{a}\overset{2}{m}\overset{3}{a}\overset{4}{p}\overset{5}{a}\overset{6}{j}\overset{7}{a}\overset{8}{m}\overset{9}{a}\overset{10}{\$}$



E.g. $ID(S, a) = \text{amap}$ $ID(S, a) = \text{ama\$}$

Suffix Trees (cont.)

S - text string $|S| = n$

$S_i = i^{\text{th}}$ suffix

Substring ID = min substr. needed to identify S_i

A suffix tree is a Patricia trie of the $n+1$ substring identifiers

Substring Queries:

How many occurrences of t in text?

- Search for target string t in trie
- if we end in internal node (or midway on edge) - return no. of extern. nodes in this subtree
- else (fall out at extern. node)
 - compare target with string
 - if matches - found 1 occurrence
 - else - no occurrences

Example:

Search("ama") → End at intern node
Report: 2 occs. ← 1, 7

Search("amapaj") → End at extern node
Go to S_i + verify ← 1

Tries and Digital Search Trees III

Analysis:

- Space: $O(n)$ nodes
 $O(n \cdot k)$ total space
($k = |Σ| = O(1)$)

- Search time: \sim total length of target string

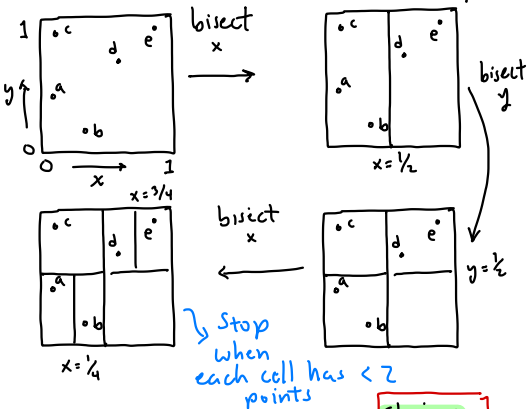
- Construction time:
- $O(n \cdot k)$ [nontrivial]

PR k-d tree: Can be used for answering same queries as point kd-tree (orth. range, near. neigh)

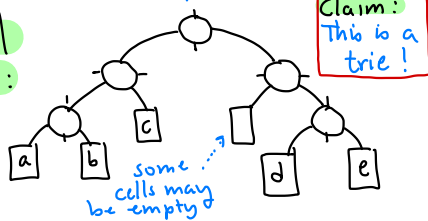
Geometric Applications:

PR kd-Tree: kd-tree based on midpoint subdivision

Assume points lie in unit square



Final tree:



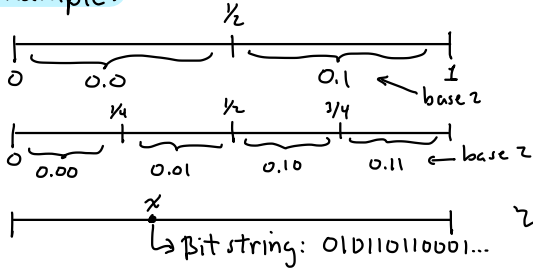
Binary Encoding:

- Assume our points are scaled to lie in **unit square**
 $0 \leq x, y < 1$ (can always be done)
- Represent each coordinate as **binary fraction**:

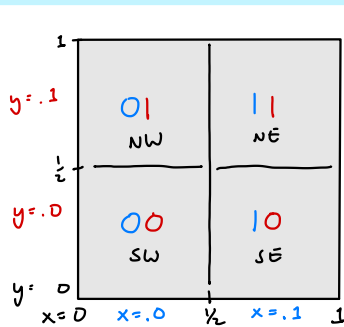
$$x = 0.a_1a_2a_3\dots \quad a_i \in \{0,1\}$$

$$x = \sum a_i \cdot \frac{1}{2^i}$$

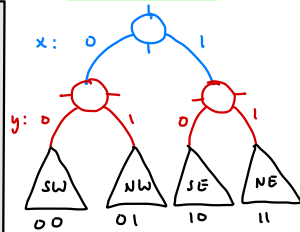
Example:



How do we extend to 2-D?



PR kd-tree



PR kd-Tree \equiv Trie ??

- Approach: Show how to map any point in \mathbb{R}^2 to bit string
- Store bit strings in a trie (alphabet $\Sigma = \{0,1\}$)
- Prove that this trie has same structure as kd-tree

Tries and Digital Search Trees IV

Bit Interleaving:

Given a point $p = (x, y)$
 $0 \leq x, y < 1$

let: $x = 0.a_1a_2\dots$ in binary
 $y = 0.b_1b_2\dots$

Define:

$$\phi(x, y) = a_1b_1a_2b_2a_3b_3\dots$$

Called **Morton Code** of p

Further Remarks:

- Techniques for efficiently encoding, building, serializing, compressing... tries **apply immediately to PR kd-tree**
- Can generalize to **any dimension**
 $x = 0.a_1a_2\dots$
 $y = 0.b_1b_2\dots$
 $z = 0.c_1c_2\dots$

Lemma: Given a pt set $P \subseteq \mathbb{R}^2$

(in unit square $[0,1]^2$) let
 $P = \{p_1, \dots, p_n\}$ where $p_i = (x_i, y_i)$
 Let $\Phi(P) = \{\phi(p_1), \phi(p_2), \dots, \phi(p_n)\}$
 (n binary strings)
 Then the PR kd-tree for P is equivalent to binary trie for $\Phi(P)$.

Proof: By induction on no. of bits

Let $x = 0.a_1a_2\dots$ $y = 0.b_1b_2\dots$
 and consider just $\phi(x, y) = a_1b_1\dots$

