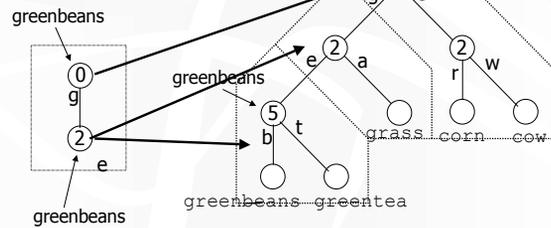




## Searching Patricia trie in Index Fabric

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- ❖ Start searching in the root layer
- ❖ One block access per layer
- ❖ Example: "greenbeans"



## Refined paths in Index Fabric

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- ❖ Queries supported by Index Fabric so far:
  - Label paths from the root (e.g., /Invoice/Buyer/Name/)
  - How about //Buyer/Name, or //Buyer/Name|Address?
- ❖ Refined paths: frequent queries
  - Just invent labels for these queries and index them in the same Patricia trie
  - Example: find invoices where X sold to Y



☞ Extra refined paths → more space required

## ViST: a sequence-based index

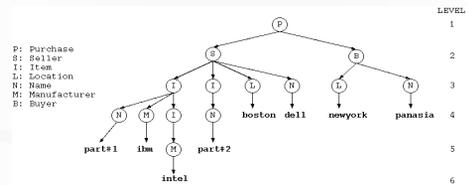
9

Wang et al. "ViST: A Dynamic Index Method for Querying XML Data by Tree Structures." SIGMOD 2003

- ❖ Use a sequence-based encoding for XML
- ❖ Turn twig queries to subsequence matches
- ❖ Index sequences in a virtual trie using interval-based encoding

## Sequence representation of XML

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- ❖ A sequence of (symbol, prefix) pairs, in depth-first order:
  - (P, ε), (S, P), (I, PS), (N, PS), (v<sub>1</sub>, PSIN), (M, PSD), (v<sub>2</sub>, PSIM), (I, PSD), (M, PSII), (v<sub>3</sub>, PSIIIM), (I, PS), (N, PSD), (v<sub>4</sub>, PSIN), (L, PS), (v<sub>5</sub>, PSL), (N, PS), (v<sub>6</sub>, PSN), (B, P), (L, PB), (v<sub>7</sub>, PBL), (N, PB), (v<sub>8</sub>, PBN)
- ☞ What is the worst-case storage requirement?
- ☞ Would listing symbols in depth-first order be sufficient?

## Sequence representation of twigs

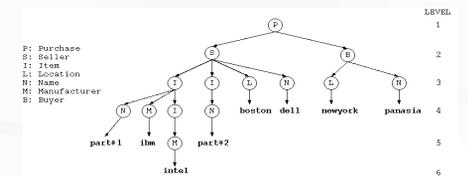
11

- ❖ Twigs can be represented sequences as well

Path Expression	Structure-Encoded Sequence
Q <sub>1</sub> : /Purchase/Seller/Item/Manufacturer	(P, ε)(S, P)(I, PS)(M, PSI)
Q <sub>2</sub> : /Purchase/Seller[Loc = v <sub>5</sub> ]/Buyer[Loc = v <sub>7</sub> ]	(P, ε)(S, P)(L, PS)(v <sub>5</sub> , PSL)(B, P)(L, PB)(v <sub>7</sub> , PBL)
Q <sub>3</sub> : /Purchase/*[Loc = v <sub>5</sub> ]	(P, ε)(L, P*)(v <sub>5</sub> , P*L)
Q <sub>4</sub> : /Purchase/[Manufacturer = v <sub>3</sub> ]	(P, ε)(M, P)(v <sub>3</sub> , P*M)

## Matching twigs as sequences

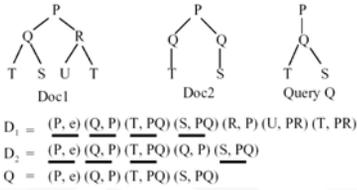
12



- ❖ Data: (P, ε), (S, P), (I, PS), (N, PS), (v<sub>1</sub>, PSIN), (M, PSD), (v<sub>2</sub>, PSIM), (I, PSD), (M, PSII), (v<sub>3</sub>, PSIIIM), (I, PS), (N, PSD), (v<sub>4</sub>, PSIN), (L, PS), (v<sub>5</sub>, PSL), (N, PS), (v<sub>6</sub>, PSN), (B, P), (L, PB), (v<sub>7</sub>, PBL), (N, PB), (v<sub>8</sub>, PBN)
- ❖ Query (Boston seller New York buyer): (P, ε), (S, P), (L, PS), (v<sub>5</sub>, PSL), (B, P), (L, PB), (v<sub>7</sub>, PBL)
- ☞ Find a (non-contiguous) subsequence of data that matches the query

## False alarms

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### ❖ $/P/Q[T]/S$

- Match sequences for  $/P/Q[T]/S$  and  $/P/[Q/T]/Q/S$
- Compute the difference between the answers
- But what if a document exhibits both structures?

## Indexing sequences with a trie

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### ❖ Just insert sequences into a trie

### ❖ Search the trie for subsequences matching the query

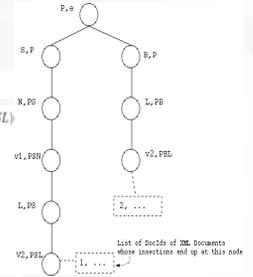
- Expensive because subsequences do not need to be contiguous

$Doc_1 : (P, \epsilon)(S, P)(N, PS)(\epsilon_1, PSN)(L, PS)(\epsilon_2, PSL)$

$Doc_2 : (P, \epsilon)(B, P)(L, PB)(\epsilon_2, PBL)$

$Q_1 : (P, \epsilon)(B, P)(L, PB)(\epsilon_2, PBL)$

$Q_2 : (P, \epsilon)(L, P)(\epsilon_2, P \cdot L)$



## “Virtual trie” idea

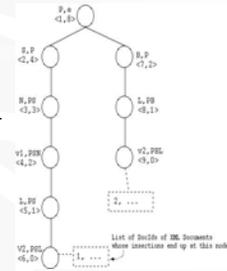
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### ❖ Use (*left*, *size*) to encode trie nodes

- $size = right - left$
- Supports efficient “skipping”

### ❖ Index in a regular $B^+$ -tree

### ❖ No need to store the trie itself



## ViST structures

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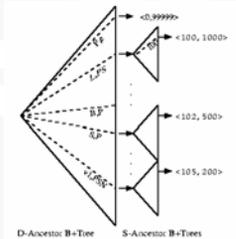
### ❖ D-Ancessor $B^+$ -tree indexes trie nodes by (*symbol*, *prefix*)

- Facilitates prefix matching (checking for ancestor-descendent relationships in documents)

### ❖ Leaf nodes point to S-Ancessor $B^+$ -trees, which further index nodes by (*left*, *size*)

- Facilitates skipping in the trie (checking for ancestor-descendent relationships in the trie)

### ❖ Subsequence matching → repeated index lookups



## Lore’s DataGuide: a structural index

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Goldman & Widom. “DataGuides: Enabling Query Formulation and Optimization in Semistructured Databases.” VLDB, 1997

### ❖ Given an XML data graph $G$ , a DataGuide is an index graph $I$ with the following properties

- Every label path in  $G$  also occurs in  $I$ 
  - Complete coverage
- Every label path in  $I$  also occurs in  $G$ 
  - Accurate coverage
- Every label path in  $I$  (starting from a particular object) is unique (i.e.,  $I$  is a DFA)
  - Efficient search: a label path of length  $n$  traverses  $n$  edges and ends at one node
- Each index node in  $I$  points to its extant: a set of data nodes in  $G$ 
  - Label path query on  $G \rightarrow$  label path query on  $I$

## Strong DataGuide

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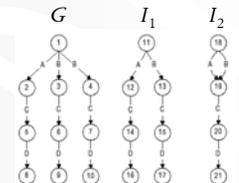
### ❖ Let $p, p'$ be two label path expressions and $G$ a graph; define $p \equiv_G p'$ if $p(G) = p'(G)$

- That is,  $p$  and  $p'$  are indistinguishable on  $G$

### ❖ $I$ is a strong DataGuide for a database $G$ if the equivalence relations $\equiv_I$ and $\equiv_G$ are the same

### ❖ Example

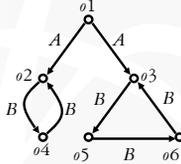
- $I_1$  is strong;  $I_2$  is not
- $A.C(G) = \{ 5 \}$ ,  $B.C(G) = \{ 6, 7 \}$ 
  - Not equal
- $A.C(I_2) = \{ 20 \}$ ,  $B.C(I_2) = \{ 20 \}$ 
  - Equal



## Size of DataGuides

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- ❖ If  $G$  is a tree, then  $|I| \leq |G|$ 
  - Linear construction time
- ❖ In the worst case, the size of a strong DataGuide may be exponential in  $|G|$  because of the DFA requirement



☞ Relax the DFA requirement?

## NFA-based structural indexes

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- ❖ Defined using an equivalence relation (based on the graph structure)
  - Each index node  $v$  corresponds to an equivalence class of data nodes in  $G$  (denoted  $v.extent$ )
  - There is an edge from  $u$  to  $v$  in  $I$  iff there exists an edge from a node in  $u.extent$  to a node in  $v.extent$
- ☞  $|I| \leq |G|$  by definition because extents do not overlap; however, the structure is no longer a DFA

## 1-index

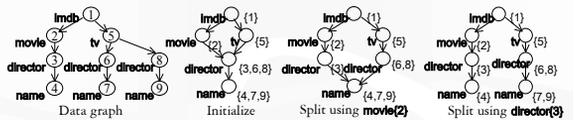
21

Milo & Suciu, "Index Structures for Path Expressions." *ICDT*, 1997

- ❖ "Perfect" equivalence relation: two data nodes are equivalent iff they are not distinguishable by label path expressions
  - That is, the sets of label path expressions that can reach them are the same
  - Too expensive to compute in practice
- ❖ 1-index uses a less perfect equivalent relation, bisimilarity, which is easier to compute
  - If two nodes are bisimilar, then they are not distinguishable by label path expressions
  - The converse is not necessary true
- ☞ May result in larger indexes

## 1-index construction

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- ❖ Initialize the index
  - Data nodes with the same label go into the same index node
- ❖ Pick an index node  $u$  to apply a *split* operation
  - For each index node  $v$ , split it into  $v_1$  and  $v_2$  (if both have non-empty extents)
    - $v_1.extent$  contains data nodes in  $v.extent$  that are children of  $u.extent$
    - $v_2.extent$  contains the rest of  $v.extent$
- ❖ Repeat *split* until there is no more change to the index