#### 1 An Overview

Overview DIS

## 2 Semistructured Data Management

#### 2.1 Horizontal Fragmentation

#### 2.1.1 Relational Databases

- 1. At which phase of the database lifecycle is fragmentation performed?
  - At database design time
  - O During distributed query processing
  - Ouring updates to a distributed database
- 2. The reconstruction property expresses that
  - O In case of a node failure the data can be recovered from a fragment from another node
  - The original data can be fully recovered from the fragments
  - O Every data value of the original data can be found in at least one fragment

#### 2.1.2 Primary Horizontal Fragmentation

- 1. Example: application A1 accesses
  - 1. Fragment F1: with frequency 3
  - 2. Fragment F2: with frequency 1

A1 accesses the whole relation with frequency



- $\bigcirc 13/7$
- $\bigcirc$  4/7
- $\bigcirc 14/7$
- 2. Consider the access frequencies below: How many horizontal fragments would a minimal and complete fragmentation have?



- $\bigcirc$  3
- $\bigcirc$  4
- $\bigcirc$  6
- 3. Which of the following sets of simple predicates is complete?



- O Location = "Munich", Budget ; 200000
- O Location = "Munich", Location = "Bangalore"
- O Location = "Paris", Budget i= 200000
- O None of those
- 4. Which is true for MinFrag algorithm?
  - O The output is independent of the order of the input
  - O It produces a monotonically increasing set of predicates
  - O It always terminates
  - O All of the above statements are true
- 5. When deriving a horizontal fragmentation for relation S from a horizontally fragmented relation R
  - O Some primary key attribute in R must be a foreign key in S
  - O Some primary key attribute in S must be a foreign key in R
  - O Both are required

# 2.2 Graph Databases

## 2.2.1 Semi-structured Data

1. Semi-structured data
○ Is always schema-less
Always embeds schema information into the data
Must always be hierarchically structured
O Can never be indexed
2. Why is XML a document model?
It supports application-specific markup
○ It supports domain-specific schemas
It has a serialized representation
○ It uses HTML tags
2.2.2 Graph Data Model
1. In a graph database
There is a unique root node
Cach node has a unique identifier
O Data values in leaf nodes are unique
○ The labels of edges leaving a node are different
O There is a unique path from the root to each leaf
2. The simulation relationship is a relation
Among nodes in the data and schema graph
Among edges in the data and schema graph
○ Among sets of nodes in the data and schema graph
○ Among sets of edges in the data and schema graph
3. Which is true?
O For each labelled edge in S a corresponding edge in D can be identified
O For each root node in S a corresponding root node D can be identified
O For each leaf node in D a corresponding typed node in S can be identified
O For each node in S a unique path reaching it from a root node can be identified
4. If there exists a uniquely defined simulation relationship among a graph database $\Gamma$ and a schema graph $S$
O The data and schema graph are simulation equivalent
Ambiguous classification cannot occur

	$\bigcirc$	Multiple classification cannot occur
5.	If schema	graph S1 subsumes S2
	$\bigcirc$	Every graph database corresponding to S1 corresponds also to S2
	$\bigcirc$	S2 simulates S1
	$\bigcirc$	S1 has fewer nodes than S2
2.2.	3 Sche	ma Extraction
1.	Which is	wrong? In a dataguide
	$\bigcirc$	Every path in the data graph occurs only once
	$\bigcirc$	Every node in the data graph occurs only in one data guide node
	$\bigcirc$	Every data guide node has a unique set of nodes
	0	A leaf node in the data graph corresponds always to a leaf node in the data guide
2.	In a non-	deterministic schema graph
	$\bigcirc$	Every node of the data graph occurs exactly once
	$\bigcirc$	Every path of the data graph occurs at most once
	$\bigcirc$	Every label of an outgoing edge of a node in the schema graph is unique
3		rmation Retrieval and Data Mining
3.1.	1 Info	rmation Retrieval
1.	A retriev	al model attempts to model
	$\circ$	The interface by which a user is accessing information
	$\bigcirc$	The importance a user gives to a piece of information
	$\bigcirc$	The formal correctness of a query formulation by user
	$\bigcirc$	All of the above
2.	If the top	o 100 documents contain 50 relevant documents
	$\bigcirc$	The precision of the system at 50 is 0.5
	$\bigcirc$	The precision of the system at 100 is 0.5
	$\bigcirc$	The recall of the system is 0.5
	$\bigcirc$	None of the above
3.	If retrieva	al system A has a higher precision than system B
	$\circ$	The top k documents of A will have higher similarity values than the top k documents of B

O	top k documents of A will contain more relevant documents than the
$\circ$	A will recall more documents above a given similarity threshold than B
0	Relevant documents in A will have higher similarity values than in B
3.1.2 Text	-based Information Retrieval
1. Full-text	retrieval means that
$\bigcirc$	The document text is grammatically deeply analyzed for indexing
$\bigcirc$	The complete vocabulary of a language is used to extract index terms
$\bigcirc$	All words of a text are considered as potential index terms
$\circ$	All grammatical variations of a word are indexed
2. The term	n-document matrix indicates
$\bigcirc$	How many relevant terms a document contains
$\bigcirc$	How relevant a term is for a given document
$\bigcirc$	How often a relevant term occurs in a document collection
$\circ$	Which relevant terms are occurring in a document collection
	query be represented by the following vectors: $(1, 0, -1)$ $(0, -1, 1)$ ; the t by the vector $(1, 0, 1)$
$\bigcirc$	Matches the query because it matches the first query vector
$\bigcirc$	Matches the query because it matches the second query vector
$\bigcirc$	Does not match the query because it does not match the first query vector
0	Does not match the query because it does not match the second query vector
4. Which is	right? The term frequency is normalized
$\bigcirc$	By the maximal frequency of a term in the document
$\bigcirc$	By the maximal frequency of a term in the document collection
$\bigcirc$	By the maximal frequency of a term in the vocabulary
$\bigcirc$	By the maximal term frequency of any document in the collection
5. The inve	rse document frequency of a term can increase
$\bigcirc$	By adding the term to a document that contains the term
0	By adding a document to a document collection that does not contain the term
0	By removing a document from the document collection that does not contain the term
$\bigcirc$	By adding a document to a document collection that contains the term

## 3.2 Advanced Retrieval Models

## 3.2.1 Latent Semantic Indexing

<ol> <li>In vector space retrieval each row of the matrix M<sup>t</sup> corresponds to         <ul> <li>A document</li> <li>A concept</li> <li>A query</li> <li>A query result</li> </ul> </li> <li>Applying SVD to a term-document matrix M. Each concept is represented         <ul> <li>As a singular value</li> <li>As a linear combination of terms of the vocabulary</li> <li>As a linear combination of documents in the document collection</li> <li>As a least square approximation of the matrix M</li> </ul> </li> <li>The number of term vectors in the SVD for LSI</li> </ol>	
<ul> <li>○ A concept</li> <li>○ A query</li> <li>○ A query result</li> <li>2. Applying SVD to a term-document matrix M. Each concept is represented</li> <li>○ As a singular value</li> <li>○ As a linear combination of terms of the vocabulary</li> <li>○ As a linear combination of documents in the document collection</li> <li>○ As a least square approximation of the matrix M</li> </ul>	
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$\bigcirc$ As a least square approximation of the matrix ${\bf M}$	
3. The number of term vectors in the SVD for LSI	
or and a second of the second	
$\bigcirc$ Is smaller than the number of rows in the matrix ${f M}$	
$\bigcirc$ Is the same as the number of rows in the matrix ${\bf M}$	
$\bigcirc$ Is larger than the number of rows in the matrix ${f M}$	
4. A query transformed into the concept space for LSI has	
$\bigcirc$ s components (number of singular values)	
$\bigcirc$ m components (size of vocabulary)	
$\bigcirc$ n components (number of documents)	
3.2.2 User Relevance Feedback	
1. Can documents which do not contain any keywords of the original query recognitive similarity coefficient after relevance feedback?	eive a
○ No	
$\bigcirc$ Yes, independent of the values $\beta$ and $\gamma$	
$\bigcirc$ Yes, but only if $\beta > 0$	
$\bigcirc$ Yes, but only if $\gamma > 0$	
2. A positive random jump value for exactly one node implies that	
a random walker can leave the node even without outgoing edges	
() a random walker	
() a random walker	
one of the above	
3. Given the graph below and an initial hub vector of $(1,1,1)$ . The hub-aut	

ranking will result in the following

- $\bigcirc$  authority vector (0,0,1); hub vector (1,1,0)
- $\bigcirc$  authority vector (0,0,2); hub vector (2,2,0)
- $\bigcirc$  authority vector (0,0,1) ; hub vector  $(\frac{1}{2},\frac{1}{2},0)$
- $\bigcirc$  authority vector (0,0,2); hub vector (1,1,0)

#### 3.2.3 Link-based Ranking

- 1.
- 2.

# Credits

Quiz questions were taken from the lecture notes of Prof. Karl Aberer.

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