Advanced Database Systems

1- D Time Series Data Indexing

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I’ll briefly introduce about 1- D time series data indexing. In fact, I cannot get currently used data such as stock data. Therefore I generate artificial data to process and show how to do indexing. I’ll have a look all processes of designed and implemented procedures. (More brief information can find in the book; ADVANCED DATABASE SYSTEMS)
An example of 1-D time series data is stock data. I have designed indexing procedures with computer-generated stock data instead of real data. As you see the histogram on the above, it denotes time series data.

Sample Data (a part)

<table>
<thead>
<tr>
<th>M1</th>
<th>M2</th>
<th>M3 Name</th>
<th>M4</th>
<th>M5</th>
</tr>
</thead>
<tbody>
<tr>
<td>57.709375</td>
<td>59.903125</td>
<td>52.584375</td>
<td>45.6375</td>
<td>38.69375</td>
</tr>
<tr>
<td>69.34375</td>
<td>60.2</td>
<td>61.340625</td>
<td>62.35625</td>
<td>58.6125</td>
</tr>
<tr>
<td>69.859375</td>
<td>77.396875</td>
<td>73.921875</td>
<td>67.8875</td>
<td>75.690625</td>
</tr>
</tbody>
</table>

NAME: ABC News

NAME: MBS TV

NAME: Hallym DB
As I said on the previous page, time series data (Stock data) generated on special program we made. The generating program made with MFC. Stock data has fluctuations about ±10. I generate 100 samples include modeled company names to show it looks like real data. (See on the previous page)
Overall Processing Method

- Preprocessing
  - 1-D Time Series Data -> Spatial Data
    - Using DFT, Clustering (Iterative Method)
    - Indexing (R-tree like method)
- Searching
  - 1-D Time Series Data -> Spatial Data
  - Search nearest data
    - Using Euclidean Distance
  - Refining Process

1-D Time Series Data Indexing has two parts of processing procedures. (Preprocessing/ Searching)

1. Preprocessing
First change 1-D Time Series Data to spatial data on the frequency domain using DFT (Discrete Fourier Transform). Second grouping with similar data using clustering method. Last indexing grouped ID (Index) to get a possibility of searching.

2. Searching
After preprocessing procedure, we can search data which we want. To find data we need to change our data to spatial data using DFT as we used in preprocessing. After get new spatial data which we want to search, calculate Euclidean distance with original spatial data. And then choose a nearest one. In result, we do refining process to get final data.

? see more on next page
Preprocessing (I)

- **DFT (Discrete Fourier Transform)**
  \[ c_k = \frac{1}{n} \sum_{h=0}^{n-1} \omega^{-kh} f_h, \quad \omega = e^{2\pi i/n}, \quad k = 1, 2, \ldots, n \]

- **No fault dismissals (Parseval’s theorem)**
  \[ D_{\text{feature}}(F(x), F(y)) \leq D(x, y) \]

Using DFT, we can change 1-D time series data to spatial data on frequency domain. After change to spatial data, choose three data (f = 1 ~ 3). If the wanted data is included (false alarm), we can find data with the refining process. But if the wanted data is not included (false dismissals), we cannot find data. The reason of changing data to spatial data is that the Euclidean distance of original data is always greater than the Euclidean distance of spatial data which is changed to spatial data using DFT. It denotes no fault dismissals. [2][3]

**DFT Pseudo Code**

```plaintext```
for k=0 to sizeofdata - 1
  real_tmp = 0  // real data
  imag_tmp = 0  // image data
  for n=0 to sizeofdata - 1
    real_tmp = real_tmp + data(n)*cos(2*PI/12*k+n)
    imag_tmp = imag_tmp + -1*data(n)*sin(2*PI/12*k+n)
  next
  power(k) = sqr(real_tmp*real_tmp + imag_tmp*imag_tmp)  // calculate power of data on frequency domain
next
```plaintext```
I use the iterative method to cluster data. Especially I used K-means algorithm to cluster. K-means algorithm is to make K numbers of cluster center using squared Euclidean distance. [4]

--- GROUP LIST(Clustering) ---

<table>
<thead>
<tr>
<th>Cluster Number</th>
<th>{Elements}</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>{0}</td>
</tr>
<tr>
<td>1</td>
<td>{28 44 87}</td>
</tr>
<tr>
<td>2</td>
<td>{19 30 67 84}</td>
</tr>
<tr>
<td>3</td>
<td>{3 32 70}</td>
</tr>
<tr>
<td>4</td>
<td>{4 39 81 91 95}</td>
</tr>
<tr>
<td>5</td>
<td>{5 37 69 83 86}</td>
</tr>
<tr>
<td>6</td>
<td>{6 46 48 62 80 93}</td>
</tr>
<tr>
<td>7</td>
<td>{7 41 71 74}</td>
</tr>
<tr>
<td>8</td>
<td>{8 36 40 63}</td>
</tr>
<tr>
<td>9</td>
<td>{9 60 61 78 88 89 90 97}</td>
</tr>
<tr>
<td>10</td>
<td>{10 31 54}</td>
</tr>
<tr>
<td>11</td>
<td>{11 72 73 92}</td>
</tr>
<tr>
<td>12</td>
<td>{12 96}</td>
</tr>
<tr>
<td>13</td>
<td>{13 42 49 59 85}</td>
</tr>
<tr>
<td>14</td>
<td>{14 35 47 77}</td>
</tr>
<tr>
<td>15</td>
<td>{15 26 34 43 55 57 58 65 82}</td>
</tr>
<tr>
<td>16</td>
<td>{16 50 64 75}</td>
</tr>
<tr>
<td>17</td>
<td>{17 68 94 98}</td>
</tr>
<tr>
<td>18</td>
<td>{18 38 51 56}</td>
</tr>
<tr>
<td>19</td>
<td>{25 52 66}</td>
</tr>
</tbody>
</table>
After clustering data as you can see on the above, we can search with cluster center values. The total data (100) changed to 25 cluster center.

--- Cluster Center ---

<table>
<thead>
<tr>
<th>DFT1</th>
<th>DFT2</th>
<th>DFT3</th>
</tr>
</thead>
<tbody>
<tr>
<td>23.340117</td>
<td>66.733917</td>
<td>10.157041</td>
</tr>
<tr>
<td>50.151770</td>
<td>25.601297</td>
<td>4.004073</td>
</tr>
<tr>
<td>61.989542</td>
<td>15.195794</td>
<td>22.199620</td>
</tr>
<tr>
<td>7.714025</td>
<td>28.789136</td>
<td>24.238079</td>
</tr>
<tr>
<td>40.608878</td>
<td>13.905752</td>
<td>7.586689</td>
</tr>
<tr>
<td>67.583971</td>
<td>34.087372</td>
<td>15.156011</td>
</tr>
<tr>
<td>45.037389</td>
<td>37.401815</td>
<td>18.088304</td>
</tr>
<tr>
<td>12.462210</td>
<td>29.154218</td>
<td>6.057019</td>
</tr>
<tr>
<td>43.047011</td>
<td>20.739349</td>
<td>31.244777</td>
</tr>
<tr>
<td>16.180685</td>
<td>15.162112</td>
<td>11.154085</td>
</tr>
<tr>
<td>8.316712</td>
<td>9.745249</td>
<td>3.354249</td>
</tr>
<tr>
<td>37.904022</td>
<td>31.389206</td>
<td>11.713221</td>
</tr>
<tr>
<td>26.594263</td>
<td>6.602002</td>
<td>19.067617</td>
</tr>
<tr>
<td>26.666832</td>
<td>18.493810</td>
<td>13.540259</td>
</tr>
<tr>
<td>16.295715</td>
<td>23.074414</td>
<td>19.648849</td>
</tr>
<tr>
<td>93.036619</td>
<td>44.201457</td>
<td>28.903012</td>
</tr>
<tr>
<td>30.942753</td>
<td>10.509283</td>
<td>10.143203</td>
</tr>
<tr>
<td>38.378123</td>
<td>18.404003</td>
<td>17.045466</td>
</tr>
<tr>
<td>25.086263</td>
<td>30.154396</td>
<td>19.368506</td>
</tr>
<tr>
<td>22.222350</td>
<td>16.704500</td>
<td>20.465816</td>
</tr>
</tbody>
</table>
As people input data on the web, the preprocessing procedure change data to spatial data using DFT. With this spatial data, we choose the nearest cluster. It denotes false alarm, we can find a result data after using refining process.
As you see on the above, we can find similar search results and refining result. On the refining process, we determined e value must be smaller than 1.0 ($e<1.0$). Otherwise the exact data which we want to search cannot be found.

**Search**

**Found Similar Search Result**
- Company Name: GVE
- Company Name: Medal co.
- Company Name: YAYAWA

**Refining Result**
- Company Name: Medal co.
Implementation (I)

- Random Generate & insert data
  - Visual C++ 6.0
- User Interface
  - ASP, Msoffice2000 component(Web Application)
- Database
  - SQL Server 7.0

We generate 1- D time series data using random generator in Visual C++. Also we designed user interface using ASP(Active Server Page) to search data on the web. 1- D time series data shows graphical view using graph components in MSOffice200.

In fact we should make database system to test data indexing. Instead of making database system, we use well-known database system(SQLServer 7.0).
Implementation (II)

- Datagram(table descript)

Table consists of cluster data and spatial data. Cluster data and ID are connected with foreign key with GRP(cluster group number).

```sql
CREATE TABLE [dbo].[timeseries] ( 
    [ID] [float] NULL ,  
    [M1] [float] NULL ,  
    [M2] [float] NULL ,  
    [M3] [float] NULL ,  
    [M4] [float] NULL ,  
    [M5] [float] NULL ,  
    [M6] [float] NULL ,  
    [M7] [float] NULL ,  
    [M8] [float] NULL ,  
    [M9] [float] NULL ,  
    [M10] [float] NULL ,  
    [M11] [float] NULL ,  
    [M12] [float] NULL ,  
    [DFT1] [float] NULL ,  
    [DFT2] [float] NULL ,  
    [DFT3] [float] NULL ,  
    [GRP] [int] NULL ,  
    [NAME] [nvarchar] (255) NULL ) ON [PRIMARY]
GO

CREATE TABLE [dbo].[cluster] ( 
    [ID] [int] NOT NULL ,  
    [C1] [float] NULL ,  
    [C2] [float] NULL ,  
    [C3] [float] NULL ) ON [PRIMARY]
GO

ALTER TABLE [dbo].[cluster] WITH NOCHECK ADD 
    CONSTRAINT [PK_cluster] PRIMARY KEY NONCLUSTERED 
    ( [ID] ) ON [PRIMARY]
GO

ALTER TABLE [dbo].[timeseries] ADD 
    CONSTRAINT [FK_timeseries_cluster] FOREIGN KEY 
    ( [GRP] ) REFERENCES [dbo].[cluster] ( [ID] )
GO
```
The image shows web designed 1- D time series data.

MAIN PAGE (overall composition)
| + Search by name
| + Search by data
| | + DFT & Similar Search & Refining
| + Graph Form Display (graphical display)
We designed and implemented 1-D time series data searching on the web. Actually we have a look a possibility of searching stock data on the web in relevantly short time. In fact there is no 1-D times series data searching product on the web.

The problem is that searching 1-D time series data searching on the web is possible but there is no analysis method to measure the performance of 1-D time series data searching.
Reference